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FIFRA SCIENTIFIC ADVISORY PANEL (SAP)

OPEN MEETING

AUGUST 24 - 25, 2004

FUMIGANT BYSTANDER EXPOSURE MODEL REVIEW:  
PROBABILISTIC EXPOSURE AND RISK MODEL FOR FUMIGANTS  
(PERFUM) USING IODOMETHANE AS A CASE STUDY

TUESDAY, AUGUST 24, 2004

VOLUME I OF II

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Reported by: Frances M. Freeman, Stenographer

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C O N T E N T S

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1 DR. ROBERTS: My name is Dr. Stephen Roberts.  
2 I'm here to serve as Chair for today's session.

3 The topic today is Fumigant Bystander Exposure  
4 Model Review: Probabilistic Exposure and Risk Model for  
5 Fumigants (PERFUM) Using Iodomethane as a Case Study.

6 I don't know if the titles of these keep getting  
7 longer or not, but it's certainly a very interesting  
8 topic. And the SAP staff have assembled an outstanding  
9 panel to address this issue.

10 I would like to start first by introducing the  
11 panel members. So let me, starting with Dr. Heeringa on  
12 my left, ask each of the panel members today to briefly  
13 introduce themselves giving their name, their affiliation  
14 and the expertise they bring to today's deliberations.

15 DR. HEERINGA: I'm Steve Heeringa. I'm director  
16 of the Statistical Design Group and a research scientist  
17 at the Institute for Social Research at the University of  
18 Michigan.

19 I'm a biostatistician and a permanent member of  
20 the SAP panel, and I'll be chairing the Thursday and  
21 Friday sessions.

1 DR. PORTIER: I'm Ken Portier, a statistician  
2 with the Institute of Food and Agricultural Sciences at  
3 the University of Florida, and a permanent SAP panel  
4 member.

5 DR. HANNA: I'm Adel Hanna. I'm a research  
6 professor with the University of North Carolina, Chapel  
7 Hill. My area of expertise is air quality and  
8 meteorological modeling and annals.

9 DR. SHOKES: I'm Fred Shokes. I'm a professor  
10 of plant pathology and director of the Tidewater  
11 Agricultural Research and Extension Center in Suffolk,  
12 Virginia, a component of Virginia Tech.

13 DR. SEIBER: I'm James Seiber. I'm director of  
14 the Western Regional Research Center, a USDA Agricultural  
15 Research Service operation in Albany, California.  
16 Formerly with the University of California, Davis.

17 DR. MAXWELL: Good morning. I'm Dave Maxwell  
18 with the National Park Service Air Resources Division in  
19 Denver, Colorado. I'm a meteorologist and manage air  
20 quality monitoring projects to the tune of about 10  
21 million dollars in visibility and air quality monitoring.

1 DR. WANG: I'm Dong Wang. I'm associate  
2 professor of environmental biophysics with the University  
3 of Minnesota in the department of Soil, Water and Climate.  
4 Formerly with USDA ARS, Soil Salinity Lab, Riverside,  
5 working with fumigants about 10 years modeling, mostly;  
6 some field experiments.

7 DR. WINEGAR: My name is Eric Winegar. I'm the  
8 principal of Applied Measurement Science. My background  
9 is in monitoring and measurement, analytical chemistry and  
10 exposure assessment.

11 DR. OU: I'm Li-Tse Ou. I'm a scientist with  
12 the University of Florida. My area of expertise is the  
13 fate of the pesticides in the soil.

14 DR. SMALL: I'm Mitchell Small. I'm a faculty  
15 member at Carnegie Mellon University in Pittsburgh. I'm  
16 in the departments of civil and environmental engineering  
17 and engineering in public policy. My areas of expertise  
18 include environmental modeling and statistics.

19 DR. MAJEWSKI: I'm Michael Majewski. I'm a  
20 research chemist with the U.S. Geological Survey. My  
21 background is in the atmospheric environmental fate of

1 contaminants.

2 DR. BAKER: Hi. I'm Dan Baker with Shell Global  
3 Solutions in Houston. I have been there 23 years. The  
4 latter half I have been working on emissions and air  
5 quality issues.

6 DR. BARTLETT: I'm Paul Bartlett, Queens  
7 College, City University of New York. My area of  
8 expertise is air transport, modeling and mission factors,  
9 semi volatiles and other organic contaminants.

10 DR. SPICER: I'm Tom Spicer, professor and head  
11 of chemical engineering at the University of Arkansas. My  
12 field of expertise is atmospheric dispersions,  
13 specifically for episodic sorts of accidental releases of  
14 contaminants in the atmosphere.

15 DR. YATES: I'm Scott Yates. I'm currently  
16 acting research leader of the Soil Physics and Pesticides  
17 Research Unit at a USDA ARS research facility in  
18 Riverside, California.

19 My area of expertise is fate and transport of  
20 pesticides and fumigants in general in soils and  
21 volatilization into the atmosphere.

1 DR. ROBERTS: Thank you. And I'm Steve Roberts.  
2 I'm a toxicologist at the University of Florida.

3 Our designated federal official today is Myrta  
4 Christian. Let me turn the meeting over to her for some  
5 announcements.

6 MS. CHRISTIAN: Thank you, Dr. Roberts.

7 I'm Myrta Christian, and I will be serving as  
8 the designated federal official to the FIFRA Scientific  
9 Advisory Panel for this meeting.

10 I want to thank Dr. Roberts for agreeing to  
11 serve as chair to the FIFRA Scientific Advisory Panel for  
12 this meeting.

13 I also want to thank both the members of the  
14 panel and the public for participating and attending this  
15 important meeting of the FIFRA SAP to review the  
16 Probabilistic Exposure and Risk Model for Fumigants  
17 (PERFUM) using Iodomethane as a case study.

18 We appreciate the time and effort of the panel  
19 members in preparing for this meeting taking into account  
20 their busy schedules.

21 By way of background, the FIFRA SAP is a federal



1 advisory committee that provides independent, scientific  
2 peer review and advice to the agency on pesticides and  
3 pesticides related issues regarding the impact of proposed  
4 regulatory actions on human health and the environment.

5 The FIFRA SAP only provides advice and  
6 recommendations to EPA. Decisionmaking and implementation  
7 authority remains with the agency.

8 As the DFO for this meeting, I serve as a  
9 liaison between the panel and the agency. I am also  
10 responsible for ensuring provisions of Federal Advisory  
11 Committee Act are met.

12 As the designated federal official for this  
13 meeting, a critical responsibility is to work with  
14 appropriate agency officials to ensure that all  
15 appropriate ethical relations are satisfied.

16 In that capacity, panel members are briefed with  
17 provisions of the federal conflict of interest laws.

18 In addition, each participant has filed a  
19 standard governmental financial disclosure report. I,  
20 along with our deputy ethic officer for the office of  
21 prevention, pesticides and toxic substances, am in

1 consultation with the office of general counsel have  
2 reviewed these reports to ensure all ethic requirements  
3 are met.

4 For members of the public requesting time to  
5 make a public comment, please limit your comments to five  
6 minutes unless prior arrangement have been made.

7 For those that have not preregistered, please  
8 notify either myself or another member of the SAP staff if  
9 you are interested in making a comment.

10 There is a public docket for this meeting. And  
11 all background materials, questions posed to the panel by  
12 the agency and other documents related to this SAP  
13 meeting are available in the docket. Overheads will be  
14 available in a few days.

15 Background documents are also available on the  
16 EPA web site. The agenda lists contact information for  
17 such documents.

18 At the conclusion of the meeting, the SAP will  
19 prepare a report as response to questions posed by the  
20 agency, background materials, presentations, and public  
21 comments.

1           The reports serve as meeting minutes. We  
2   anticipate the meeting minutes will be completed in  
3   approximately eight weeks.

4           Again, I wish to thank the panel for their  
5   participation. I am looking forward to both a challenging  
6   and interesting discussion over the next two days. Thank  
7   you.

8           DR. ROBERTS: Thank you, Myrta.

9           I'm pleased to see that we have with us this  
10   morning the director of the office of pesticide programs,  
11   Mr. Jim Jones. Good morning, Jim, welcome.

12          MR. JONES: Thanks, Dr. Roberts.

13          I want to thank all of you for willing to serve  
14   on this panel today and tomorrow. And for those of you  
15   who will be here on Thursday and Friday, I want to thank  
16   you as well in advance.

17          Two of the hallmarks of the agency's work in the  
18   office of pesticide program are transparency and sound  
19   science. I mention that because two of those  
20   characteristics are part of the reason why we're here  
21   today.

1           Transparency, in the sense that we try to do our  
2   best to do our business in front of the public. And we  
3   think that that's very important that the business of the  
4   agency be conducted in a way that the public can watch  
5   what we're doing and participate in that.

6           Sound science, in the sense that we don't feel  
7   at EPA that we have all the answers we often feel. It is  
8   very important for us to reach out to individuals with  
9   expertise specific to the issues that we're dealing with  
10   and get independent peer review of our work.

11           And those two are the very important reasons why  
12   we're here today and tomorrow and, for some of you, for a  
13   little longer than that.

14           I'm going to give a little bit of the regulatory  
15   context within which we are operating. I realize that you  
16   are here to provide scientific expertise, but I think it  
17   is very important for you to understand the general  
18   context within which we're operating around the chemicals  
19   that this work is going to apply to.

20           The agency has an old chemical program, and we  
21   have a new chemical program. There are a number of

1 fumigants that are in our old chemical program. And  
2 within the next year or so, we're going to be making  
3 regulatory decisions, decisions around their safety and  
4 the ultimate regulatory disposition of those chemicals in  
5 the next 18 months or so.

6 We also have before us a new chemical. Actually,  
7 today's analysis uses that new chemical as somewhat of a  
8 test case that's a fumigant. And that chemical is also  
9 before us for a licensing decision.

10 The agency has made a choice to look at these  
11 fumigants all at the same time so that we're not just  
12 trading off potential risks from one to another. We're  
13 looking at all of them at the same time so we can make a  
14 comprehensive logical risk management choice around these  
15 fumigants.

16 We have convened a number of SAPs following the  
17 Food Quality Protection Act to help us deal with some of  
18 the very difficult hazard and exposure issues associated  
19 with pesticide risk assessment.

20 Today's issues aren't going to be about hazard,  
21 obviously. They are going to be about exposure. Some

1 previous SAPs have really helped us as an agency, as an  
2 office, to sort through some very complex exposure issues  
3 associated with how to estimate exposure of pesticides in  
4 food when you have 12 or 13 chemicals that may share a  
5 common mechanism of toxicity. Those are some of the  
6 probabilistic SAPs we had around the organophosphates.

7 The SAPs also helped us sort through how to  
8 measure pesticide exposures through the source of  
9 drinking water when there is huge degree of variability in  
10 pesticide use in the United States and how pesticides may  
11 enter drinking water.

12 Today, we're talking about today and for the  
13 next four days, two for many of you, two more for some  
14 others, and in another meeting we're going to be having in  
15 about two weeks, we're going to be talking about a  
16 completely different source of exposure.

17 That is the exposure to what we refer to as  
18 bystanders. Bystander I'll describe simply as an  
19 individual who is near or around a treated field, that  
20 treated field being a field treated with one of the  
21 fumigants that we have in front of us for regulatory

1 decisionmaking.

2           We certainly do have measured estimates or  
3 measurements, actual, of these chemicals. But those  
4 measurements tend to be somewhat limited in that we don't  
5 have as many data sets, not as robust as you optimally  
6 would like it to be.

7           So what we're exploring are potential models to  
8 help us better characterize the exposure to bystanders  
9 from these fumigants. And better characterization through  
10 this route of exposure will certainly enhance the agency's  
11 decisionmaking as it relates to these chemicals.

12           And so basically, that's what we are here for  
13 the next couple of days is to help get some advice about  
14 some of the models that are available to the agency in its  
15 efforts to estimate exposure to bystanders, individuals  
16 who are near or around treated fields, fields being  
17 treated with these fumigants.

18           I'm confident that the advice that we get, not  
19 only from this panel, but from the two subsequent panels  
20 that are going to be looking at some of these models, will  
21 be instrumental in the agency's ultimate determination as

1 to how to estimate exposure to bystanders from these  
2 compounds.

3           So I thank you all for your service. I know  
4 that you all have very busy and active professional lives  
5 and this is no small endeavor for you to come and, not  
6 only sit for two days, but to invest the time and energy  
7 it takes to review the scientific documentation before  
8 these meetings and the work that goes on after these  
9 meetings to ultimately write the reports. I want to thank  
10 you all for your service.

11           And, thank you, Dr. Roberts, for your  
12 chairmanship.

13           DR. ROBERTS: Thank you. I think those remarks  
14 help the panel place our discussions today in the right  
15 perspective, so we understand that perspective.

16           We also have with us the director of the health  
17 effects division of the office of pesticide programs. I  
18 would add a veteran of many SAP meetings, Margaret  
19 Stasikowski. Welcome.

20           MS. STASIKOWSKI: Thank you. And actually, this  
21 maybe the last SAP that I'm at. I'm leaving, some say



1 retiring, but I'm actually changing careers. So I will be  
2 leaving the agency on October 1st.

3 And one of the things that I will really miss is  
4 the Science Advisory Panel and the advice that you have  
5 provided us during my tenure, which has been eight years  
6 as the director of the health effects division.

7 And you have done a wonderful job helping us  
8 addressing some of the cutting edge risk assessment issues  
9 that for the last eight years as we implement FIFRA and  
10 FQPA.

11 Today, over the next three meetings, this week  
12 and early in September, we are asking you to look at three  
13 different modeling approaches for assessing exposure from  
14 soil fumigants.

15 We are asking you to review this in a similar  
16 way, independently, this is not a comparative model  
17 assessment, in a similar way that you looked at the CARES,  
18 Calendex and Lifeline probabilistic dietary exposure  
19 models.

20 You have done a wonderful job. We are using the  
21 models and have improved them significantly as the result

1 of the input from the Science Advisory Panel.

2 So today, we are asking you to review PERFUM,  
3 Probabilistic Exposure and Risk Model for Fumigants.  
4 That's today and tomorrow.

5 That presentation will be made by Dr. Rick  
6 Reiss, who is making a presentation on behalf of Arvesta  
7 Corporation. He was a consultant to Arvesta.

8 Thursday and Friday we are asking you to review  
9 Fumigant Emissions Modeling System. The presentation will  
10 be made by Mr. David Sullivan from Sullivan Environmental  
11 Consulting.

12 And early in September we'll be asking you to  
13 review SOFEA, Soil Fumigant Exposure Assessment System. I  
14 think that's a very interesting acronym.

15 Without further ado, I would like to introduce  
16 Jeff Dawson, one of our most seasoned exposure assessors  
17 in health effects division who will give some introductory  
18 remarks.

19 But before that, I would like to introduce two  
20 people sitting here to my left who are here from  
21 California Department of Pesticides Regulation. And we

1 really are very happy with the fact that we are working  
2 very closely in this issue of fumigants risk assessment  
3 with California.

4 California has so much more experience in  
5 developing the approaches to fumigant risk assessment that  
6 we really couldn't do this without them.

7 So I would like to introduce Dr. Terri Barry and  
8 Dr. Randy Segawa from California Department of Pesticide  
9 Regulation. Thank you.

10 DR. ROBERTS: Dr. Dawson, I think you are next  
11 up.

12 MR. DAWSON: Thank you.

13 Thanks, Margaret for the introduction.

14 What I like to do today is give a 15-minute  
15 primer or so to help set the stage for the scientific  
16 aspects of this discussion. So if everyone will look up  
17 there at the screen.

18 Basically, what we're going to do is talk about  
19 these four topics very quickly. I'm going to give you a  
20 little bit of information about the background for the  
21 science that we're going to be talking about.

1 I'm going to talk about our current  
2 methodologies for looking at fumigants. So that will be a  
3 good basis for you to compare how this model may differ  
4 and in what aspects.

5 And then a very brief summary of the PERFUM  
6 model. Because you are going to hear a lot about that in  
7 much more detail from Dr. Reiss right after my  
8 presentation.

9 Then we'll talk a little bit just about the  
10 general theme of the charge questions. As we move later  
11 into the meeting, we'll read the specific charge  
12 questions.

13 As far as the background information, we'll  
14 touch very quickly on the modeling approaches, the source  
15 of the case study that we're looking at today, the purpose  
16 of the model and our ultimate goal with this meeting.

17 So as you just heard, we're looking at these  
18 three different models over the course of the next three  
19 meetings up until the middle of September. And again,  
20 we're focusing on the PERFUM approach today.

21 The PERFUM approach is -- what we're going to be

1     doing is looking at a case study based on iodomethane,  
2     which is a chemical, as Jim Jones said, we're considering  
3     for registration and licensing.

4             And the specifics of this case study we're  
5     looking at are based on a variety of field monitoring  
6     data. And the last bullets, few bullets there basically  
7     summarize the types of data that we have available to us  
8     for the basis of this modeling.

9             Basically, what they are is they are emissions  
10    data that are representative of different types of  
11    application methods shank injection with a flat fume,  
12    shank injection with a raised bed type of application, and  
13    then a drip irrigation method with a raised bed. For  
14    example, like you would have with growing strawberries.

15            So the purpose of this model, and Jim did an  
16    excellent job of really explaining why we're interested in  
17    these models, and that is to have a better understanding  
18    of the distributions of bystander exposure after an  
19    application of fumigants, soil fumigants such as this.  
20    And we're really interested in using these to help us  
21    characterize the higher end exposures.

1           And also we would really want to use these  
2   tools, I think, to look at how uncertainties and  
3   variability affect the exposure levels.

4           And so our goal here is manyfold, but,  
5   basically, we're interested in the scientific validity of  
6   the model, how transparent is it to go from the inputs to  
7   the outputs and understanding the whole process,  
8   understanding what types of data are required to operate  
9   the system, looking at how systems such as this might be  
10   used for evaluating exposures across the country in  
11   different growing regions, different kinds of crops, and  
12   also understanding how portable this system might be for  
13   looking at -- using the methodologies here for looking at  
14   other chemicals. For example, the six soil fumigants  
15   we're looking at over the next year or so.

16           So our current approach is very similar,  
17   basically, more or less identical to what California  
18   Department of Pesticide Regulation is using. And it's  
19   based on an agency model called the ISCST model.

20           And we'll talk a little bit about the inputs we  
21   use in our current approach and then the outputs that we

1 get from that.

2           So many of you on the panel are familiar with  
3 this model, but the ISCST model or the industrial source  
4 complex model was developed by the Office of Air and it is  
5 routinely used for their permitting programs and routinely  
6 used in regulatory decisionmaking.

7           It is a steady state Gaussian plume approach.  
8 And it can look at all different types of sources. For  
9 example, it can look at point sources, which are things  
10 like smoke stacks, linear sources, so you might look at  
11 pollution from a roadway where there is a lot of traffic,  
12 and area sources.

13           And in this case, we're using treated farm  
14 fields as our example. And we're using it to deal with  
15 farm fields as an area source.

16           As I just said, DPR uses this model as well. And  
17 for those of you who are not aware, this model is publicly  
18 available. You can go to that web site right there and  
19 download the system and all the documentation associated  
20 with it.

21           So the inputs that we routinely use, they

1 basically fall into five categories. And the first three  
2 are listed here. For example, for field size and  
3 geometry, and geometry is just the shape of the field that  
4 we're using, we use a range from 1 to 40 acres. Again,  
5 that's similar to what DPR is doing.

6 The shape of the field we're using is a square.

7 And we look at varied atmospheric conditions. We  
8 basically go to the lowest wind speed that's allowable in  
9 the model up to about 10 miles an hour. And we look at  
10 varied environmental stability. That's just a measure of  
11 turbulence in the atmosphere. So we look at a range of  
12 inputs that go from a calm day up to a reasonably  
13 turbulent type of day.

14 DPR uses, on that one bullet there, that's the  
15 inputs that they have used. For example, for their  
16 methylbromide permitting, to use a specific set.

17 The third major category of input is the  
18 different application equipment and what are called  
19 control technologies. That's just basically categorizing  
20 the types of data we have available for field monitoring.

21 For example, in this case, we have data for drip and



1     irrigation and shank injection methods. And then  
2     different -- in addition to that, different methods for  
3     reducing or trying to control emissions. For example,  
4     tarping and the use of raised beds as an approach.

5             Based on the data we're looking at in the case  
6     study, the next category is field emissions. This is the  
7     actual emission data that we have used for the case study.

8             And from the data we had available, we  
9     calculated constant, what are called, flux rates, which is  
10    an emission from a treated field in this case. And the  
11    actual numbers run from 66 to 107. And the units are  
12    microgram per square meter per second. That's just the  
13    surface area on the field and the amount coming out per  
14    second.

15            For the categories of application equipment,  
16    sorry, I just showed you, we calculated flux rates for  
17    each different combination of data that we had available.

18            And what we saw in this case is that drip  
19    irrigation was the lowest emitter, and that the highest  
20    emitter was the shank injection flat fume approach.

21            And then the first four inputs are really more

1 chemical specific type of inputs and more scenario  
2 specific for this. But there is also other settings and  
3 parameters within the model that you would routinely set  
4 for whatever kind of analysis you are doing.

5           These are just a few examples of what we have  
6 done in our assessment. We use rural conditions. We  
7 treated it as an area source. And because it's a treated  
8 farm, we're using a release height of zero meters. So the  
9 emissions are released right at the surface of the field.

10           This slide just shows the kind of outputs that  
11 we're getting from the model. You can see our treated  
12 field there on the left, a square. And then basically  
13 what we're doing is we're modeling the wind direction  
14 going downwind 100 percent of the time.

15           So we're making the assumption that the wind  
16 direction is not changing. And then we're calculating air  
17 concentrations on that receptor grid, I'll call it, on  
18 the right.

19           So at those different locations we would get  
20 some sort of a calculated air concentration at various  
21 distances downwind that we use for our risk assessment.

1           But the key to take away from this is that one  
2   of the parameters we're using is that that wind direction  
3   is 100 percent downwind all the time.

4           And this is just what a table of results might  
5   actually look like in an assessment. And this is just  
6   something I extracted from the charge document that's  
7   available on the web site.

8           And what you have here are different distances  
9   downwind in the second column. For example, I presented  
10  from 25 meters down to 1,000 meters downwind. And then as  
11  you go across the columns there, you see the air  
12  concentrations.

13          And the reason the columns are different, as you  
14  go from left to right on the columns, you go from a calmer  
15  day to a much more turbulent type of day. So the more  
16  turbulent the atmosphere, the lower the concentrations  
17  get.

18          And you can see that. For example, at 25 meters  
19  on a calm day you are at 2,116 micrograms per cubic meter  
20  in the air. If you go to a situation where there is 10  
21  mile an hour winds in a less stable atmosphere, you go

1 down to 214.

2 Obviously, the concentrations go down with this.

3 You can see.

4 What we do with these concentrations is we  
5 calculate a measure of risk called an MOE or margin of  
6 exposure. It is shown in the equation there at the  
7 bottom. Basically, what we do is divide these  
8 concentrations into some sort of regulatory threshold  
9 called the HEC or human equivalent concentration.

10 Now what I will do, just so you can kind of  
11 compare and contrast what we're doing and with what PERFUM  
12 potentially could offer us, we'll talk about PERFUM in the  
13 next few slides. And again, PERFUM is, and you will hear  
14 much more from Dr. Reiss in a minute, but PERFUM is based  
15 on the use of the industrial source complex model, just  
16 like we're doing.

17 A key difference here is it uses five years of  
18 historical meteorological data from different stations. I  
19 think in the case study today there are four different  
20 stations that were looked at in Florida and California.

21 And then it also can allow you to look at

1     variability and the emission terms, which we basically  
2     assume a 24 hour average. Again, it is important to us  
3     because it is potentially a tool we could use to look at  
4     uncertainty and variability.

5                 So just for comparing with what we're doing and  
6     then with the case study Dr. Reiss is going to present, he  
7     also used 1 to 40 acre fields and, as I just said, five  
8     years of meteorological data from two stations in  
9     California, Bakersfield and Ventura, and two stations in  
10    Florida.

11                He also used varied emission -- he also used the  
12    variety of application methods and emission control type  
13    of data that we have available to us. And as part of the  
14    system, he also was able to integrate in the actual flux  
15    rate changes over time, where we're using an average.

16                This is just an illustration of flux data from  
17    two sites. This is a broadcast flat fume applications.  
18    In our case study, we used the data from the pink line,  
19    which is Manteca. That's just the site in California.  
20    And the other one is Watsonville in California.

21                Basically, the same kind of application method.

1     You see how -- it is a little bit different over time.  
2     Essentially, what this is is a graph of the percent of  
3     flux rate on the Y axis or a measure of how much is coming  
4     off the field as a percent of application rate versus the  
5     time after application on the X axis.

6             Another key input for the PERFUM model is the  
7     use of actual meteorological data over the five-year  
8     period. This is what is called a wind gross (ph) plot for  
9     Bakersfield, California, that shows basically how the wind  
10    speed and direction changed over time. The size of the  
11    bars represents the amplitude of the wind speed and  
12    obviously the direction shown.

13            This is the kind of output you get from PERFUM.  
14    Basically, what you do is you calculate the red contour  
15    line first. That's just a measure around the perimeter of  
16    the treated field to -- it is a measure of the distance at  
17    which you get to a certain threshold concentration that  
18    you designate.

19            The black line is a measure of the distance at a  
20    selected percentile of exposure. So I think in this case  
21    it was 95th percentile of exposure off the red line. And

1     then the shaded area there is -- you can use that to  
2     define the exposure exceedances over time. That's what  
3     that represents.

4             Very different from our approach where we're  
5     looking at a single receptor line downwind.

6             And basically right now I would like to quickly  
7     wrap up. This slide really represents the theme of the  
8     charge questions that we'll be talking about later. We're  
9     very interested in understanding if you believe that the  
10    documentation of the system is adequate and reflective of  
11    what it does.

12            We're also interested in your evaluation of the  
13    overall system design and the required inputs that you  
14    need to operate the system.

15            And then, finally, we're interested in how the  
16    results were presented, are they clear, can you follow it  
17    through from the beginning to the end and have a clear  
18    understanding of what they represent.

19            DR. ROBERTS: Thank you.

20            Are there any questions from the panel for Mr.  
21    Dawson based on his presentation?

1 Dr. Small.

2 DR. SMALL: I have a question. I'm not sure if  
3 this is -- it is kind of a background issue. I'm not sure  
4 if we're going to get to it in our charge questions. I  
5 don't have a lot of experience with fumigants.

6 One of the basic assumptions in the use of the  
7 ISC model is that everything is in a vapor phase. And I'm  
8 just wondering for those who have had experience should I  
9 worry about particulate phase association. Are there ever  
10 issues with wind blown soil or dust particularly during  
11 high wind periods on which there could be some  
12 particulate association as well.

13 Perhaps those who have worked with it can talk  
14 about the reasons for not considering that sort of  
15 mechanism or pathway.

16 DR. BARRY: We have really only worried about  
17 the vapor phase. I think really that's probably for  
18 fumigants all we need to worry about. Because they are  
19 not really applied in high wind situations. And we  
20 haven't worried about the soil, dust blowing from the  
21 site.



1 I don't really know if that's an issue. Maybe  
2 one of the panel members would have some expert answer.

3 DR. ROBERTS: For the record, the last response  
4 was from Dr. Barry, just so we can sort of keep it all  
5 straight, if folks could identify themselves before they  
6 speak. I think you are about to respond, Dr. Reiss. Is  
7 that correct?

8 DR. REISS: Yes. One point to consider is that  
9 they are injected into the ground when they are applied.  
10 So just by gravity you wouldn't expect a particulate to  
11 escape from the field.

12 DR. ROBERTS: Dr. Yates and then Dr. Baker.

13 DR. YATES: Another thing with fumigants is that  
14 they tend to not absorb the soil particles nearly as  
15 strongly as other pesticides. The absorption is quite  
16 low.

17 So in general, absorption effects really aren't  
18 probably that significant.

19 DR. SMALL: Because they are VOCs (ph) rather  
20 than SVOCs (ph). DR. YATES: Yes, they have very high  
21 vapor pressures and low absorption.

1 DR. ROBERTS: Dr. Baker, did you want to add  
2 something to that?

3 DR. BAKER: Is there the potential for any  
4 vehicle traffic on the field after application which would  
5 enhance the flux of the fumigant?

6 DR. SEGAWA: Randy Segawa with Department of  
7 Pesticide Regulation. Normally, in fact, in all cases at  
8 least in California, there are prohibitions to reentering  
9 that field once it has been treated for at least a week.

10 DR. SHOKES: Fred Shokes from Virginia Tech. I  
11 noticed the peaks on the diagram you showed us, the graph.  
12 Do you know what the factors are that cause these peaks?

13 DR. BARRY: Terri Barry with DPR. Those are  
14 often diurnal changes in the flux. We tend to see in our  
15 field studies higher flux values at night. So you are  
16 seeing a diurnal pattern, just like you are going to see  
17 in Dr. Reiss' presentation.

18 And it is atmospheric conditions that lead to  
19 the higher flux at night.

20 DR. ROBERTS: Dr. Yates.

21 DR. YATES: For the flux studies that we have

1 conducted, we tend to see higher fluxes during the day,  
2 but lower air concentrations because there is a lot more  
3 air movement.

4 At night, generally, we see lower fluxes, which  
5 is usually due to stable conditions occurring, which tend  
6 to kind of repress the ability -- basically, it changes  
7 the concentration gradients at the soil surface so you  
8 tend to have less flux at nights.

9 But basically, Terri is correct about the  
10 diurnal fluctuations do cause the emissions to vary  
11 through the day where you have generally, at least from my  
12 experience, it is high values of flux during the day and  
13 low values at night.

14 DR. ROBERTS: Dr. Wang.

15 DR. WANG: To follow up on Scott's comments on  
16 the high emissions, it's a diurnal variations that is  
17 showing. That's why it's also attributed to the effects  
18 of the diffusion coefficients and also possibly the  
19 permeability. If there is a tarp cover during the day,  
20 especially later afternoon, so that's when you will likely  
21 see a higher partitioning to the vapor phase and higher

1 movement diffusion rate and higher permeability. So  
2 that's probably why you will see a (inaudible) emission  
3 flux.

4 DR. ROBERTS: We're going to have a presentation  
5 now from Dr. Reiss about the model. Let me just ask if  
6 there are any questions regarding clarifications from Dr.  
7 Dawson, Mr. Dawson's presentation before we move on to  
8 Reiss'.

9 Seeing none, then let's go ahead and get to Dr.  
10 Reiss' presentation on the model.

11 DR. REISS: I prefer to stand. Is that all  
12 right? Good morning, everyone.

13 I just want to start with an overview of the  
14 presentation. Give you a general guide as to where I'm  
15 going and to when I'll approximately be done. I'll start  
16 with an introduction and try to lay out the issues.

17 Then I'll give a detailed description of the  
18 PERFUM model that we have built. And I think the major  
19 inputs of the models are the flux rate and a  
20 characterization of meteorological conditions in the  
21 growing areas following these applications. So the

1 following two sections will deal with those issues.

2 We'll give some results of the case study  
3 analysis using iodomethane. And then we'll describe some  
4 of the uncertainty analysis that we have done, and,  
5 finally, some conclusions.

6 This shows a picture for one of the field  
7 studies that Arvesta has conducted to characterize the  
8 flux rates of iodomethane. This is a raised bed  
9 application. You can see the tractor moving across the  
10 field. And as it is moving across the field, it is laying  
11 shanks that injects the fumigant into the ground, and then  
12 a tarp immediately is laid over the field to limit the  
13 emissions and also increase the efficacy of the product by  
14 trapping it into the soil for a long period of time.

15 As you can see in this particular circumstance,  
16 there are houses nearby, which is one of the motivations  
17 for looking at bystander exposures. This is probably  
18 closer than in most circumstances, but it happens.

19 So setting this up, after application, there is  
20 a potential for escape from the field surface of these  
21 fumigants causing potential downwind exposures.

1           In California, buffer zones have been  
2   established that restrict entry around the field after  
3   application. And the purpose of that is to mitigate these  
4   inhalation exposures to these bystanders.

5           So what sort of questions do risk managers need  
6   to know about bystander exposure?

7           First, and the most obvious thing is what the  
8   distribution of concentrations of iodomethane or any other  
9   fumigant downwind of these applications. And then if we  
10   have a distribution of concentrations, what sort of  
11   criteria can we use to establish a protective buffer zone.

12          Then finally, once we establish a protective  
13   buffer zone, we want to specifically look at what are the  
14   exposures and the risks at the perimeter of that buffer  
15   zone in an issue of risk management.

16          As I said, the PERFUM model was developed with  
17   funding from Arvesta Corporation, which is currently  
18   seeking a registration for iodomethane. We have used  
19   iodomethane in this analysis as a case study.

20          Iodomethane is a preplant soil biocide. It has  
21   activity against insects, plant parasitic nematodes, soil

1 borne pathogens and weed seeds.

2 In the marketplace, it is going to be labeled  
3 MIDAS as a trade name. That's going to be a combination  
4 of iodomethane and chloropicrin in various formulations  
5 that are listed there, the four different ratios of the  
6 products.

7 May be used for growing strawberries, fresh  
8 market tomatoes, peppers and some other types of plants.  
9 But the focus here is on field crops like strawberries,  
10 tomatoes and peppers.

11 It is a liquid at ambient temperatures with a  
12 moderate vapor pressure. When these shanks or it's  
13 injected by drip irrigation, the compound turns into a gas  
14 mostly.

15 These plastic tarps as I showed are laid down  
16 over the field surface immediately after the application.

17 These are used to mitigate exposures and increase  
18 efficacy.

19 As Jeff just described, there are three  
20 potential application methods that iodomethane can be used  
21 for, a shallow shank, broadcast flat fume, which we'll

1 just refer to as flat fume, two raised bed methods, one  
2 with injection, shallow shank injection, we'll just refer  
3 to that as a raised bed, and then the raised bed drip  
4 irrigation, which we'll just refer to drip irrigation.

5 And I have pictures of all these which we'll  
6 show in a moment.

7 The current toxicity threshold that EPA is  
8 considering for iodomethane is 120 micrograms per meter  
9 cubed. That's one of the inputs into the model. We're  
10 interested in buffer distances up to about 120 micrograms  
11 per meter cubed. That's averaged over 24 hours. So we're  
12 looking at 24 hour exposures for iodomethane.

13 This will likely change with some new data that  
14 are being developed. But for the purposes of this field  
15 case study, we're using 120 micrograms per meter cubed.

16 Typically, per field you would expect one  
17 application per year.

18 So let's move on to a description of the model.

19 I don't want to get into all of the mechanics of an air  
20 dispersion model. Some of you, looking at your bios, are  
21 very familiar with air dispersion modeling.



1           For some others, it may be a newer topic. I  
2    just want to give the basis for air dispersion modeling.  
3    In a steady state mode, it is what is called a Gaussian  
4    dispersion.

5           You see here kind of a classic graph that's been  
6    used to describe what is going on. You have the stack  
7    source here. It is emitting a plume out into the  
8    atmosphere.

9           In both the cross wind direction, you are  
10   assuming a Gaussian description of concentrations. And in  
11   the vertical direction, you are assuming a Gaussian  
12   distribution. And there are coefficients in the model  
13   called dispersion coefficients that characterize the  
14   shapes of those distributions depending on certain  
15   circumstances.

16           So what does a dispersion model do? Quite  
17   simple. For a source emitting some gas or particle into  
18   the atmosphere, it estimates the concentrations at any  
19   location around the source.

20           If I have a source out here, I can predict with  
21   my model, I can get some estimate of the concentration in

1 all directions around that source.

2 The model, the EPA model that we're using as a  
3 basis for this spits out hourly concentrations. Those can  
4 be averaged over longer periods such as we'll do in this  
5 study for 24 hours.

6 The input requirements particularly for a  
7 fumigant application we need the source dimensions. So if  
8 it's a one acre field, we need to define coordinates that  
9 define a one acre field. In this case for the case study  
10 that we're showing we're assuming all square fields,  
11 although you can do all different dimensions if you would  
12 like.

13 You need a flux rate or sometimes called an  
14 emission rate. For an area source, we prefer the term  
15 flux rate because it is not just mass per time, but it's  
16 mass per area per time. So it could be pounds per acre  
17 per day, for example, would be a unit of the flux rate.

18 We need a characterization of the meteorology  
19 following the application. That includes the wind speed,  
20 wind direction and atmospheric stability on an hourly  
21 basis.

1           I just want to talk a little bit about  
2   atmospheric stability because it's a key parameter in the  
3   model. And some of the different data sources that we  
4   have used to characterize the meteorology we needed to use  
5   different methods to characterize the stability.

6           It is a measure of atmospheric turbulence and  
7   it's expressed on an ordinal scale from A through F with  
8   increasing stability, so F would be a most unstable  
9   conditions and A would be the least unstable conditions.

10          If you were to emit a gas into the air, all  
11   things being equal, it would disperse more rapidly during  
12   A stability than it would during F.

13          During the daytime, you can have anywhere from A  
14   to D stability, and at nighttime from D to F. So you  
15   generally see more stable conditions at the nighttime  
16   periods.

17          So I want to give a little bit of background on  
18   fumigant buffer zones before we get to the guts of the  
19   PERFUM model. The most work to date has been done by the  
20   California Department of Pesticide Regulation, who is here  
21   today, primarily for methylbromide. So a lot of the work

1 we're presenting here today builds on that.

2 DPR has established buffer zones for  
3 methylbromide that vary by application method and  
4 application rate. So they have basically a table for  
5 methylbromide where you would have the application rate on  
6 one or actually a flux rate on one axis and a field size  
7 on the other axis and you would sort of look up what  
8 buffer zone you would need for that field size and flux  
9 rate.

10 So a brief description of the DPR approach. They  
11 back calculate a flux rate from a field study or sometimes  
12 use a direct calculation. I will describe how that is  
13 done in Section 3. Then they run EPA's ISCST 3 dispersion  
14 model, which is a commonly used dispersion model that's  
15 been in various forms around at EPA for 20 or more years.

16 And they use at when they are using a 24 hour  
17 exposure, they use a wind speed of 1.4 meters per second,  
18 which is about 3.1 miles per hour, a C class stability,  
19 which is moderately unstable, and a constant wind  
20 direction for 24 hours.

21 Then they estimate the distance required for

1 that concentration to decline to a level of concern.

2 Here is just a brief picture that shows what  
3 that might look like. We have a source here to the left.

4 And this contour line shows -- we're basically inside of  
5 the contour line. The concentration is above the  
6 threshold concentration that we're interested. And  
7 outside the line, the concentration is below that  
8 threshold.

9 So when we want to set a buffer zone, one way to  
10 do it is to say what is the maximum distance from that  
11 source before the concentration declines to a level below,  
12 a level of concern. And so that would be this distance  
13 here.

14 Now, one of the key assumptions in this  
15 approach is that the wind direction is constant for 24  
16 hours. And as we have shown, that's generally not the  
17 case. But there is others -- like C class stability, you  
18 could have more stable conditions than a C class  
19 stability.

20 So the goal in PERFUM is to try to look at all  
21 those parameters put together using historical

1 meteorological data and see where it comes out.

2           So the PERFUM approach, we want to run the  
3 ISCST3 model with five years of historical meteorological  
4 data. And this is not hard to get. If you are not  
5 familiar with air modeling, there is a plethora of data  
6 sources out there of historical meteorological data on an  
7 hourly basis that you can use to generate these data sets.

8           One distinction I want to make is that it is a  
9 little bit different than people when they are running  
10 like an air model for permitting conditions where they  
11 have a source that's mostly constantly emitting, like a  
12 stack source or an industrial source.

13           In that case, they are running the model for  
14 five years. And each day, that is a concentration of it  
15 would have been that day for those meteorological  
16 conditions.

17           In this case, the fumigant's only emitted maybe  
18 once per year. So what we're really doing in this case is  
19 developing a time series or a set of potential 24 hour  
20 average concentrations that could have existed if the  
21 fumigant was applied on that particular day.

1           If we knew that the fumigant was applied on a  
2   given day, then we could assign it that. But we don't  
3   know what day the fumigant is going to be applied. So we  
4   have this five year set of 24 hour concentrations, which  
5   are all possibilities for what might be the actual case.

6           We're going to estimate concentrations in all  
7   directions around the field. So this gives us sort of an  
8   exposure probability. So if you were talking about  
9   somebody who is at the perimeter of the field, by  
10   estimating the concentrations in all directions, you would  
11   be able to give some sort of exposure probability of them  
12   being exposed to a given concentration.

13           We're also going to use the actual measured  
14   diurnal flux rates whereas in the DPR and EPA current  
15   approach they are using 24 hour average flux rate. And  
16   we're going to show later in the presentation in the case  
17   study that that can be a critical factor.

18           Particularly, you are best off having your  
19   applications at the beginning of the day so that more of  
20   the emissions, relatively speaking, are during the  
21   unstable, higher wind speed conditions during the daytime

1 period as opposed to the very stable conditions at night.

2 So this is a graph that Jeff also showed, an  
3 example of probability of exposure. This is just the  
4 PERFUM output for one day for a field. And the contour  
5 line shows the distance to the level of concern, 120  
6 micrograms per meter cubed in this case.

7 One way to set the buffer zone is to say we want  
8 to make sure that 95 percent of the perimeter of the  
9 buffer zone is below the level of concern. And that's  
10 this line here. This line basically sets this length at  
11 five percent of the total circumference of the field.

12 This is just one way to do it. I'm not  
13 advocating any particular way in this presentation. But  
14 really it is a risk management decision in the end, is to  
15 how to actually set the buffer zones. The PERFUM model is  
16 going to output concentration data in a variety of formats  
17 that leaves a variety of options for how to set those  
18 buffer zones.

19 So what sort of distributions are output from  
20 PERFUM that you could use to set a buffer zone. The first  
21 one we call the whole field. These are the distances to



1 threshold in all directions for the field.

2 We divide the field into sort of slivers or  
3 spokes. And we calculate the distances that it takes to  
4 get to the threshold concentration that we're concerned  
5 about in all those directions.

6 So the number of buffer lengths I might get for  
7 a five year data set, I would have the number of spokes I  
8 have divided the field into or the number of pie slices,  
9 if you will, times the number of days. So I might get  
10 100,000 different numbers for a five year data set.

11 The other distribution that it outputs is the  
12 maximum daily concentration. And this distribution is  
13 simply the maximum distance to the threshold in one  
14 direction. So you are basically taking -- this value here  
15 is the maximum distance to get below the threshold from  
16 the field. So that distribution would just include that  
17 one value for that day. So you would get -- the number of  
18 buffer lengths you would get would be equal to the number  
19 of days that you modeled.

20 So options for setting the buffer zone. You  
21 could look at this whole field approach and you could set

1 the buffer zone based on an upper percentile of the whole  
2 field concentration distribution. Or you could take the  
3 maximum daily concentration approach and you could set the  
4 buffer zone based on an upper percentile of maximum  
5 concentration distribution. Or you could do some  
6 combination of the two.

7 We have an additional program that's part of  
8 PERFUM, which I'll describe in a moment, which you can use  
9 to analyze any buffer length that you were interested in.

10 So how is PERFUM structured? It is a Fortran  
11 model built with the Lahey 95 compiler. And most of EPA's  
12 mathematical models are built with the same for of  
13 platform and are in Fortran.

14 This provided a good opportunity for us. We  
15 found that -- the first version of PERFUM that we  
16 developed, we basically ran the ISCST 3 model, took that  
17 output and then PERFUM was a post processing program that  
18 analyzed that output.

19 There was a lot of disadvantages to that. One,  
20 the output was rather large. And there were a few other  
21 sort of technical details that we couldn't accomplish that

1 way.

2               So what we did is we went to the EPA web site  
3 and we downloaded the ISCST 3 source code, which EPA  
4 provides, and we were able to compile that on our own  
5 compiler. And we basically built the ISCST 3 model into  
6 PERFUM.

7               So we converted the model into a subroutine  
8 that's called by PERFUM. So the way PERFUM basically  
9 works is that you run the main model, it sets up all the  
10 input information that you need, and then it calls ISCST 3  
11 as a subroutine, and it runs through five years of  
12 meteorological data.

13              And after each day or each hour in some cases,  
14 it calls PERFUM subroutines that tabulate those results in  
15 a way that we need them to output the data. And then when  
16 it is finished with the five years of meteorological data,  
17 it returns to the main model and outputs the results in  
18 the format we need.

19              One of the other features in the model is that  
20 we're able to treat the flux rate as a probabilistic  
21 variable. We're going out in these field studies and we

1 have a method to measure these flux rates. But, of  
2 course, there is an uncertainty in those measurements,  
3 like all measurements.

4 In the model, we, basically, treat that  
5 uncertainty by each time we call a flux rate. So for each  
6 period in the model where we need to call a flux rate, we  
7 perturb that flux rate based on the standard error of the  
8 measurement of the flux rate, which I'll describe how we  
9 get in a moment and multiply that by a Z score.

10 So we're assuming a normal distribution. So  
11 we're basically perturbing on a normal distribution basis.

12 That's slightly different from the version I  
13 provided in the original submission. We had a T  
14 distribution with 11 degrees of freedom. I found it  
15 easier with some changes I'll describe in a moment to  
16 convert that to a Z distribution, a normal Z score  
17 distribution, the typical normal distribution.

18 Had very little difference in the model. But as  
19 I said, I couldn't help but continue to tinker with it a  
20 little bit.

21 One of the things I think I did to improve the

1 model is we have added a random number generator. Whereas  
2 before we just fed the model 2000 random T values that we  
3 got that I derived in Excel, I have added now so you have  
4 a random number generator using the Lahey Fortran  
5 compiler. It generates a random number between zero and  
6 one. Then I have a module that calculates the Z score, a  
7 normal distribution Z score from that random number.

8 And I used that to perturb the flux rates. One  
9 of the issues I'm interested in getting the panel's views  
10 on is, as a regulatory model, I have chosen a specific  
11 random number seed in my program. So basically it is  
12 using the same set of random numbers every time it calls  
13 the model.

14 The advantage of that I think for a regulatory  
15 purpose is that you would get the same result every time  
16 you ran the model. Whereas if you were to randomly pick a  
17 seed each time, which you can also do in the compiler we  
18 have, you might get slightly different results each time  
19 you ran the model, because it would pick a different  
20 distribution of random numbers.

21 There could be a temptation there to run the

1 model over and over until you got the lowest result you  
2 could find.

3 I took a look at this, and looked at the  
4 stability of the estimates using 10 different random  
5 seeds. And we found that it is really a very stable  
6 result when you are looking at the 95th or 99th  
7 percentile.

8 If you are looking at the whole field  
9 distribution, the coefficient of variance was below 1  
10 percent, .6 percent. So you're getting less than a one  
11 percent difference in the model results.

12 If you go beyond the 99 percentile, you are up  
13 at 99.9, 99.99, you can start to get upwards of maybe  
14 about a five percent difference in model results for  
15 different runs.

16 But if you assume you are going to regulate  
17 somewhere in this region, I think you are getting  
18 repeatable results that are usable.

19 One of the inputs to the model that we need to  
20 give it is a receptor grid, meaning we need to tell it at  
21 what points around this field do we need to calculate the

1 concentrations.

2           So this is an example receptor grid for the  
3 model. I think it is for a five acre field. I'm not  
4 sure. But they all look the same. You see the field here  
5 in the middle in green. And there is all these different  
6 rings around the field of receptor points.

7           Each of these points is a receptor where we're  
8 going to calculate the concentration in the model. There  
9 are actually 28 rings in all the scenarios. And there is  
10 a number of spokes. I think there is more than 100 or 200  
11 in this particular example.

12           So in addition to these rings, we defined a  
13 spoke with this blue line as an example of a spoke where  
14 it is just a set of numbers, one number in each ring  
15 around the field.

16           And the advantage of assigning these spokes is  
17 that we can do an interpolation of the concentration  
18 results and we can actually calculate the buffer zone even  
19 if it is in between two of these different rings.

20           This is just a blow-up of the northwest corner  
21 of one of the plots. And you see the rings in a little

1 bit closer proximity. This might be 10 meters away and  
2 that might be 20 meters away and 30 meters away. So we're  
3 calculating the concentrations at every point along these  
4 arcs.

5 We wanted to get the model to run as quickly as  
6 possible and get accurate results. When we first ran the  
7 model, we had what we called the fine grid, what we now  
8 call the fine grid coordinates. These included 96 to 232  
9 spokes around the field or pie slivers around the field  
10 where we're calculating concentrations, depending on the  
11 field size.

12 And what we found is that -- we changed that  
13 number, we reduced that number by a factor of four to  
14 develop a course grid scenario. And the model ran more --  
15 ran, obviously, a quicker, about fourfold quicker. And it  
16 got results at the 95th percentile and 99th percentile  
17 that were accurate -- as accurate, nearly as accurate as  
18 the fine grid.

19 So as a matter of computational efficiency, the  
20 course grid could be used for most circumstances.

21 So let me just review some of the features of



1 the model. The model outputs the full percentile  
2 distributions of buffer lengths for what we define as the  
3 whole field and maximum concentration percentiles.

4 It gives the percentiles from the first to 99th  
5 percentile. It also gives the 99.9 and 99.99 percentiles.

6 The model repeats these calculations for up to  
7 10 user supplied application rates. The advantage of this  
8 is that the buffer zones, the buffer zone tables, for  
9 example, for methylbromide are established as a function  
10 of application rate.

11 And it takes quite a while for the model to run  
12 through a whole and other flux rate. If you were to just  
13 run the model over, it would take as long as it took to  
14 get the answers for the first flux rate.

15 But, fortunately, the model is linear between  
16 application rate and emission rate -- emission rate and  
17 concentration. So we can take advantage of that linearity  
18 and just do a simple ratio to adjust the concentrations  
19 from one run with a particular flux rate to another run  
20 with another flux rate.

21 So the model will do that and it will output

1 the results for up to 10 different user supplied  
2 application rates.

3           Also, a lot of these fumigants are applied in  
4 particular seasons. Maybe it is applied -- the growing  
5 season is from April to June or something. So what we  
6 have done is we have outputted the buffer lengths on a  
7 monthly basis so you can see the -- and I will show a  
8 distribution later of how they tend to vary by season.

9           We didn't employ this for the case study  
10 analysis, because when you look at national buffer zones  
11 -- establishing a national buffer zone, it is a little  
12 hard to generalize about what particular months the  
13 application might occur. But it is something that the  
14 model outputs then could be used for seasonal analysis.

15           We have several field sizes, 1, 5, 10 and 40  
16 acres. This could be expanded to other field sizes and  
17 dimensions either by us by providing a receptor grid or  
18 the user can even input their own receptor grid with a  
19 different field size or even a different field geometry.

20           Just like if you were to run ISC, it outputs the  
21 ISCST 3 output file. You need to check that for errors.

1 You can also get a summary of a lot of the output  
2 information from that file.

3 And we have included about 60 error and warning  
4 messages to help the user debug potential problems or  
5 inconsistencies in the input data or diagnose potential  
6 model problems.

7 So one of the key interesting things about the  
8 model is how long it takes to run. We ran just -- I'll  
9 give an example, on my system, which is a 2.4 gigahertz  
10 processor with 512 megabytes of RAM. We used the course  
11 grid option. And the model runs take anywhere from 5 to  
12 23 minutes depending on the field size. So it is  
13 relatively fast.

14 And we provided with the model a series of DOS  
15 batch files which you can use to do multiple runs in the  
16 same session or just let it go in the evening like we have  
17 done. And I think to run the 120 scenarios that we ran  
18 for the case study, you can do that in generally two  
19 evenings. So it is not too onerous.

20 Recent refinements to the program. As I  
21 described, we did some modifications to the random number

1 generator. There is some additional error checking. We  
2 improved the interpolation algorithm in a very minor way.

3 And in particular, I want to note that we have  
4 developed some scenarios to investigate the impact of  
5 multiple fields emitting at the same time. I'll describe  
6 those in the uncertainty analysis.

7 Finally, we have added a second program into the  
8 package called PERFUM MOE. We call it a risk management  
9 tool.

10 If you are not familiar with how EPA often does  
11 risk assessments, they define a margin of exposure as the  
12 human equivalent NOEL. So they might get a NOEL from an  
13 animal study and they may convert that to a human  
14 equivalent concentration, and they divide that by the  
15 exposure.

16 You can see by the form of this equation that --  
17 well, they can translate into safety factors or something  
18 like that.

19 So generally, EPA, for many their of  
20 applications, as a policy decision seeks to have an MOE of  
21 100. And what that would translate into is an exposure

1     that's 100 fold less than the concentration known to --  
2     the lowest concentration known not to cause an effect.

3             Other MOEs besides 100 are used in various  
4     circumstances in the matter of policy and uncertainty in  
5     the database. But for the purposes of this example, we  
6     have looked at a 100 fold MOE.

7             So this program, what it does, is that for a  
8     given buffer zone distance, and no matter how you get it,  
9     whether you get it from PERFUM or it is just a buffer zone  
10    that you are interested in looking at, say I just want to  
11    know what the exposures would be for a 300 foot or a 500  
12    foot buffer zone, this program will calculate the  
13    distribution of margins of exposure.

14            So like from the first to 99 or 99.9 percentile  
15    at the perimeter of the buffer zone.

16            So for a person that's at the perimeter of the  
17    buffer zone or a location at the perimeter of the buffer  
18    zone, more accurately, this program will calculate that  
19    distribution of margins of exposure. It is a way for risk  
20    managers to look at what the potential risks are for a  
21    given buffer zone.

1           Also, it is a way if you were to define a buffer  
2    zone that doesn't assure that in 100 percent of the  
3    circumstances you had a 100 fold margin of safety or  
4    margin of exposure, you would be able to estimate how high  
5    -- or how low in this case, how much lower below 100 it  
6    could get.

7           So really, the severity of any exceedance above  
8    100 or below 100.

9           There are some key conservative assumptions to  
10   bear in mind when you are looking at the results of this.

11    Our calculations assume that a bystander spends 24 hours  
12    following the application at the perimeter of the buffer  
13    zone.

14           If it was the residents, that could very well be  
15    the case. But for many bystanders, they may spend less  
16    than 24 hours at the perimeter of the buffer zone.

17           But as a conservative assumption, for one, and,  
18    secondly, because it is difficult to define the  
19    probability of someone not spending 24 hours there, we  
20    have assumed a 24 hour exposure at the perimeter.

21           The calculation also assumes that the bystander

1 is either outdoors for 24 hours or that the indoor  
2 exposure is the same as the outdoor exposure. We don't yet  
3 have chemical specific data for iodomethane, but we know  
4 it is like methylbromide, a relatively sticky compound.

5 So it is quite possible that the indoor exposure  
6 is less than the outdoor exposure. But for the purposes  
7 of this case study, we have assumed that the indoor and  
8 outdoor exposure is the same.

9 Part 3, I want to talk about how we have  
10 estimated the flux rates for our case study for  
11 iodomethane and generally how -- some methods to estimate  
12 those flux rates from field studies.

13 For iodomethane, Arvesta, the sponsor, has  
14 currently conducted seven field studies. Those include  
15 two with flat fume, two with the drip irrigation  
16 application method and three with the raised bed  
17 application methods. And there may be more field studies  
18 conducted subsequently.

19 We have calculated the flux rates using a  
20 methodology developed by California Department of  
21 Pesticide Regulation that uses the field study results and

1 the ISC model to essentially back calculate the flux rate.

2 It is sometimes called the indirect method.

3 So just to give you an example as to how a field  
4 study is designed and analyzed, let's go through this one  
5 example. You have a square field here. And we have  
6 established 12 monitors to measure the iodomethane  
7 concentration following the application. We have put eight  
8 monitors at about 30 feet from the field in the four  
9 directions around the field. And then we have additional  
10 monitors at the corners about 140 feet from the corners.  
11 We have a total of 12 monitors to characterize the  
12 iodomethane concentration following the application.

13 So that example showed 12. There were others  
14 where we had only had eight. These are generally charcoal  
15 air samplers, and they are also put at about 1 to 1.5  
16 meters above the ground.

17 The measurements are collected in periods  
18 generally encompassing the daytime and nighttime period.  
19 So we get separate estimates of the flux during the  
20 daytime and then during the nighttime. And I'll show an  
21 example of how we divided those periods in a moment. And



1 that's used to capture the diurnal variability in the flux  
2 that we have already discussed.

3 The samples are collected for a minimum of about  
4 10 days. But the peak was always in the first 24 hours  
5 with iodomethane. That may not be true for all fumigants.

6 But for iodomethane, in the seven field studies we have  
7 had, the peak emissions always occurred in the first 24  
8 hours. So that's the focus of our analysis. And they  
9 generally declined to negligible emissions in five to  
10 seven or maybe a few more days.

11 This is an example for the Manteca study how we  
12 divided the flux periods.

13 This basically shows the 24 hours following the  
14 application. We established monitors to capture an  
15 average concentration over the first three hours, the next  
16 three hours, the seven and eight, the next two hours after  
17 that. So we had three separate measurements of the flux  
18 for that first eight hour period.

19 Then there was a long nighttime sample. I think  
20 that's about 13 hours. Then what is really the next day,  
21 we had an additional sample that starts in the next

1 daytime period. So that's generally how most of the  
2 studies were designed.

3 So how do we get a flux rate out of the ISC  
4 model? We know the ISC model will predict concentrations  
5 downwind of a field following an application. But we  
6 don't know what flux rate to put in.

7 But we know that there is some flux rate that  
8 must statistically best predict the concentration profile  
9 that we observe. So we use a method developed by DPR to  
10 best -- to determine the flux rate that best explains the  
11 data that we observe in the study.

12 So what we do is we run the ISCST model with a  
13 nominal flux rate. And then we use a statistical method  
14 to calculate the best fit flux rate. That method takes  
15 advantage of the fact that the concentration and flux rate  
16 are linearly related.

17 So we don't need to keep running the model over  
18 and over with different flux rates to know what the  
19 concentrations would be for different flux rates.

20 Once we run the model once for one flux rate,  
21 just by a ratio we can determine what the concentrations

1 would be for another flux rate.

2           What is done is we do a linear regression of the  
3 model, the measured concentrations. We have maybe our  
4 eight to 12 measured concentrations here on the Y axis and  
5 our modeled concentrations at each of those exact receptor  
6 points on the X axis, and then we calculate a slope and an  
7 intercept.

8           And then what we do is we can't adjust the flux  
9 rate with both the slope and the intercept. So we're  
10 looking at just a slope using -- multiplying that slope  
11 from the regression and multiplying it by that nominal  
12 flux rate that we first ran the model, and that gives us  
13 the flux rate that best explains our observed data.

14           Sometimes there is a problem when you first do  
15 this. Sometimes the fit is poor or sometimes the  
16 intercept term is statistically significant. And I'll  
17 show an example in a moment. So you need to consider the  
18 following options. And DPR suggests the following.

19           First, you sort the data independently and rerun  
20 the regression. The theoretical basis for this is that  
21 the model predicts -- is known -- when they validate the

1 ISC model and other dispersion models, they find that it  
2 predicts the maximum concentration quite well, but it  
3 doesn't necessarily predict the location of the maximum  
4 concentration. So sorting the data sort of removes that  
5 spatial element.

6 You could also sort the data and constrain the  
7 intercept to zero if you are still having a problem  
8 getting a large intercept.

9 So here is an example where we have our measured  
10 concentrations on the Y axis and our modeled  
11 concentrations at those same receptor points at the X  
12 axis. You see there is a relatively good agreement  
13 between the two.

14 And so let's try to fit a linear regression to  
15 it. So we get the following result. We get an R squared  
16 of .97, which is excellent, a slope 4.8 and an intercept  
17 of .0028.

18 The problem was the intercept was statistically  
19 significant. So our concern here is that some of the  
20 explanation -- some of what we're trying to explain in the  
21 measured model data is incorporated in that intercept,

1     which is unusable to us in calculating a flux rate.

2             So let's sort the data independently. We see  
3     these red dots, and then rerun the regression. Again, we  
4     get an excellent fit. Even a little bit better than  
5     before. Almost .99. However, this intercept term was  
6     still statistically significant.

7             Finally, we go to our last option. We just  
8     calculate a regression with the intercept constraint  
9     through zero. So we get a slope of .5126. We just  
10    multiply that by the flux rate that we ran the ISC model.

11    And that gives us the sort of best fit flux rate or the  
12    most -- the flux rate that best explains our observed  
13    data.

14            This just shows another example where the  
15    intercept was low when we first ran the regression. Again,  
16    you have a very high R squared, about .95.

17            If you look at the data, you generally get some  
18    very good correlations for the first 24 hours. But there  
19    are occasionally samples where you don't. And then the  
20    correlations generally could decline if you go farther  
21    away from the first day.

1                   And the reason there is, I think, first, the  
2   measurement variability when you get to lower  
3   concentrations is going to be more substantial.

4                   So we need an estimate of the uncertainty on  
5   that slope estimate to use in our model so we can perturb  
6   the emission rates.

7                   And the way we do that is basically estimate the  
8   standard error on that slope. And I convert it just for  
9   convenience to a coefficient of variance that I can use in  
10  the model.

11                  There are some other options. There has been a  
12  lot of talk about how to go about estimating these flux  
13  rates. I just want to go through a few other options that  
14  you can possibly consider.

15                  You could do a linear regression with the  
16  intercept constrained through zero just from the start.  
17  Not as one of the options. But just use that from the  
18  start.

19                  The advantage of that is that it would minimize  
20  the mean square error between the predicted estimates with  
21  the adjusted flux rate and the measured concentrations.

1           It may not be the best fit slope, because, if  
2   you remove that intercept, you may miss some of the data.

3           You could also use log transform data in a  
4   linear regression. The advantage here is that that would  
5   normalize the data. One of the assumptions of a linear  
6   regression is that the data are normally distributed.

7           But because these data vary over many more  
8   orders of magnitude, they generally aren't normally  
9   distributed. Although, we found that the residuals --  
10   when you look at the residuals of the regressions, there  
11   doesn't appear to be a bias. So that that may mitigate  
12   that concern.

13           One of the problems here is that it minimizes  
14   the mean square error of the logged values instead of the  
15   raw values. The effect that has is that it places more  
16   emphasis on the lower concentration values that may have  
17   higher uncertainty.

18           Another idea would only include values above the  
19   limit of detection or maybe even above some higher  
20   concentration. I generally found that that normalizes the  
21   data and you get somewhat larger standard errors, which

1 maybe are more reflective of the standard error at the  
2 maximum concentrations that you observed. So these are  
3 some other options to consider.

4 I should mention that whatever option is used,  
5 the model and all the models I think you are going to  
6 hear, could accommodate that. It is just a matter of how  
7 you derive that input data.

8 I want to go through some of the field study  
9 data that has been conducted. This shows a flat fume  
10 application. You see, basically, a flat field, and the  
11 tractor is laying down shanks to inject the material. And  
12 this tarp is immediately being rolled over the field to  
13 trap the emissions.

14 Arvesta has done two flat fume studies to date,  
15 one in Manteca, California, and the other in Watsonville,  
16 California.

17 Let's look at some of the results that we got.  
18 This is also a graph that Jeff showed a few moments ago.  
19 I'm showing the flux rate as a percent of application,  
20 which is a convenient way to present it.

21 So this tells you that, say, for this data point



1 here, maybe that's at about 18 percent, that means that in  
2 the first period for the Manteca study 18 percent of the  
3 applied mass was emitted from the field. So these are --  
4 as a function of the application rates, some healthy  
5 amounts come off the field.

6 And this just shows the profile as you go  
7 through the five to seven days after the application.

8 These spikes you see here are generally  
9 diurnally related. And as we already talked about, there  
10 could be a variety of reasons for that, soil temperature,  
11 the soil permeability of the tarp, the amount of the  
12 material in the vapor phase.

13 But it is something that for methylbromide was  
14 the case and is something that has been repeatable in the  
15 iodomethane studies in almost all the data we see. We see  
16 this diurnal profile show up quite often -- I should  
17 mention, lower flux rates during the nighttime period and  
18 higher flux rates during the day.

19 This shows a raised bed application, the same  
20 chart I showed at the beginning. The beds are made prior  
21 to the material being injected. The tractor goes off, the

1 shanks are injected, and the tarp is laid immediately  
2 afterwards to trap the emissions.

3           So far, Arvesta has conducted raised bed studies  
4 in Oxnard, California, in Plant City, Florida, which is a  
5 popular growing area in Florida. And the most recent  
6 study was done in Guadalupe, California. That was just  
7 done in May. So it wasn't included in the submission  
8 package that you all received, but we just recently  
9 analyzed that data.

10           This shows you the raised bed flux rate as a  
11 percent of application and the mean time since  
12 application. You see really a remarkably similar profile  
13 between the three application methods.

14           There is obviously some experimental variability  
15 for whatever reason. You have this peak a few days out  
16 for the Plant City study, but you really see a very  
17 similar profile among these three studies all with the  
18 same application method.

19           This shows the drip irrigation. Basically, a  
20 sprinkler system. The fumigant is injected with  
21 irrigation water into the sub surface strip lines and the

1 plastic tarps are already covering the treated beds.

2           So far, drip irrigation studies have been  
3 conducted in La Selva Beach, California, that's near  
4 Watsonville, and in Camarillo, California. That's  
5 relatively close to Santa Barbara.

6           And this shows, again, a comparison of the  
7 decline profile for the drip irrigation studies. Again,  
8 just like the raised bed, they look very similar, decline  
9 profile, for the two drip irrigation studies.

10           Let's summarize the data that we have from our  
11 seven field studies so far. Again, I'm putting in this  
12 terms of percent of application emitted in the first 24  
13 hours.

14           You see for the flat fume we have at Manteca and  
15 Watsonville 47 percent and 35 percent. That was actually  
16 the largest difference we saw between measurance with the  
17 same application method.

18           For raised bed, we had three different methods,  
19 and they were 55 to 61 percent. They were very tight.

20           And for drip irrigation, we had 42 to 50  
21 percent between La Selva Beach and Camarillo.

1           One of the observations you can draw from these  
2   data is that during the first 24 hours as much as half or  
3   more of the application rate is emitted during that  
4   period. That's a significant amount.

5           And it explains why these bystander exposures  
6   are a concern. It also reduces the sort of variability or  
7   uncertainty that we might be worried about with bystander  
8   exposures. Because we're already assuming a pretty  
9   substantial amount of the mass is emitted during that  
10   first 24 hour period.

11           One of the things we wanted to look at is what  
12   are some of the factors that are causing the variability  
13   we observe.

14           One of the obvious things is temperature. We  
15   have done studies in the winter or at least the fall, the  
16   summer, the spring. So we have a variety of temperatures  
17   that we have observed during our studies.

18           This shows a graph of the flux rate as a percent  
19   of the application emitted versus temperature, the average  
20   daily temperature or the average 24 hour temperature for  
21   that first daytime period or that first 24 hour period.

1           There is no apparent correlation between those.  
2       That's not to say that temperature has no effect on the  
3       emissions that we observed. We do see this diurnal  
4       profile, but it may suggest that we're looking at maybe a  
5       diffusion limited process rather than a temperature  
6       limited process.

7           I just want to underscore the importance of the  
8       first 24 hours. A majority of the iodomethane emissions  
9       occur during this first 24 hour period after application.

10           Therefore, the first 24 hours of emissions  
11       produce the peak exposures. And the second day of  
12       emissions were generally about half of what we observed  
13       during the first day.

14           Let's take a look at a chart of the second day  
15       emissions. This just shows a plot, the blue bars showing  
16       the flux rate for the first 24 hours, and the red bar for  
17       the second 24 hours.

18           In all cases, the flux rate during the first 24  
19       hours was half or more than twice the flux rate during the  
20       second 24 hours. That shows us that from a risk  
21       standpoint we're most concerned about the first 24 hours.

1           There is also -- in addition to this direct flux  
2   calculation where we look at measured concentrations  
3   downwind to the field and try to infer a flux rate from  
4   the model, the ISC model, there is also a direct flux  
5   method.

6           In Manteca, the one study, we actually did both  
7   the direct and indirect flux methods. Let's look at what  
8   we got for that. For the direct flux rate method,  
9   sometimes called the aerodynamic method, there is monitors  
10  placed at varying heights at the center of the field,  
11  typically.

12           And there is a fluid dynamic calculation that is  
13  used to calculate the flux rate based on those observed  
14  concentrations along these mass that are vertically  
15  situated.

16           So it provides for us an independent  
17  verification of the flux rate. And you will see that for  
18  Manteca, the one where we have a comparison of the two  
19  methods, they were very comparable.

20           The blue line shows the estimates we got for the  
21  direct flux method. And the dashed red line shows the

1 estimates for the indirect method. They track pretty  
2 well, I think, as you can see.

3           There is no apparent bias. In some cases, the  
4 direct flux method was higher. In other cases, the  
5 indirect method was higher. If you look at the overall  
6 flux rate for the first 24 hours, I think there was about  
7 a five percent difference between these two methods.

8           That could be fortuitously close, maybe, but it  
9 gave some good reassurance that the direct flux method is  
10 giving comparable results.

11           DR. ROBERTS: I think before Dr. Reiss goes on  
12 to Part 4, it would be a good idea to take about a 15  
13 minute break. And then we'll resume his presentation, and  
14 after that, give the panel the opportunity to ask him  
15 questions.

16           Let's reconvene in 15 minutes.

17           (Thereupon, a brief recess was taken.)

18           DR. ROBERTS: Let's go ahead and get started  
19 with Dr. Reiss' presentation on Part 4.

20           DR. REISS: We talked just before the break  
21 about characterizing the flux rates of iodomethane

1 following the applications. The other major input we need  
2 to the PERFUM model is a characterization of the  
3 meteorological conditions in the growing regions following  
4 the applications.

5 So how can we derive suitable meteorological  
6 data? We have a problem. Most of the historical  
7 meteorological data that have been used for dispersion  
8 modeling are built for stack sources, large point sources.

9 Most of these data have come from urban  
10 airports. If you look on EPA's dispersion modeling web  
11 site, there is a variety of data sources from the National  
12 Weather Service. And they are primarily from large urban  
13 airports.

14 So there is a concern that these may not be  
15 representative of the growing regions. So the solution  
16 is we looked at other sources of meteorological data that  
17 have historically not been used for dispersion modeling.  
18 And we did some comparisons afterwards to determine  
19 whether using these data actually alter the results in any  
20 way.

21 So what are the potential sources of



1 meteorological data that we can consider? First, as I  
2 already mentioned, in National Weather Service. This is  
3 an historical data set available on EPA's modeling web  
4 site and most commonly used in dispersion modeling.

5 It is an observer collected system, meaning it's  
6 collected by meteorologists. And as I said, it's commonly  
7 used in meteorological applications.

8 That observer collected system has now been  
9 replaced by the ASOS system, the Automated Surface  
10 Observing System. It is maintained by the Federal  
11 Aviation Administration. And it is basically, as I said,  
12 an automated system where there is instrumentation that  
13 automatically measures these parameters and stores them in  
14 a data set requiring relatively infrequent maintenance.

15 In California, we also have the California  
16 Irrigation Management information System or CIMIS. That's  
17 an automated system run by the state of California, used  
18 for irrigation management planning. Obviously, since it is  
19 for agriculture, there are stations that are close to some  
20 of these growing regions.

21 And then somewhat similar in Florida we have the

1 Florida Automated Weather Network or FAWN. That's an  
2 automated system run by the state of Florida, used for  
3 agricultural management. Obviously, there are also  
4 stations in the growing regions.

5 Let's look at some advantages and disadvantages  
6 from these four data sets.

7 The NWS, it is widely used. It's high quality  
8 control. The data on the web site from EPA has already  
9 been quality controlled beyond what was originally  
10 collected. However, there are a few stations and there  
11 are not many in the growing regions that we're most  
12 concerned about for these fumigant applications.

13 ASOS, there is a much larger number of stations  
14 in the ASOS system. There are many stations -- they are  
15 all collected at airports, generally, but there are many  
16 other stations, many collected at small airports.

17 For example, Watsonville in California is a  
18 common growing region. And there is a small airport  
19 there. There is an ASOS station there. It's not all that  
20 uncommon that a lot of these small airports that are  
21 generally in rural areas may have an ASOS station.

1           There is quality control, but it's done on an  
2   automated basis. It's not done by a meteorologist. And  
3   there is some issues with collecting cloud cover, you can  
4   imagine with an automated system.

5           Cloud cover, I should point out, is one of the  
6   variables that's used to calculate the stability class  
7   with some methods. You can imagine it is somewhat  
8   difficult to collect a measure of the cloud cover from an  
9   automated system that basically is sending some sort of  
10   signal up to look at the opacity.

11           There has been some analysis done by EPA or by a  
12   contractor that have found that it doesn't always  
13   correlate as well with the observer collected data.

14           CIMIS, there is a large number of stations in  
15   the growing areas of California. It has quality control,  
16   but, again, it's automated. And it's also collected at a  
17   two meter height, which is not standard for dispersion  
18   modeling.

19           Most data for meteorological circumstances are  
20   collected at 10 meters. Some of the NWS data is collected  
21   at six meters. So there was a concern that the data

1 collected at two meters may not be representative.

2 The FAWN data, there are stations, obviously.

3 It is an agricultural network. So there are stations in  
4 the key growing regions. However, there is very little  
5 quality control in the FAWN network.

6 When I analyzed the data, I found many  
7 inconsistencies in the data, circumstances where you had a  
8 two meter per second wind speed on one hour and 100  
9 meters per second the next hour and then back down to two  
10 meters per second.

11 So I ended up calling the people who run this  
12 FAWN network. And they were pretty honest and said the  
13 quality control is, I think, "rudimentary." So there is  
14 not a lot of quality control in that system. But when you  
15 look at the data set in Florida, there aren't many other  
16 stations in the growing regions.

17 What we did is we decided we would take a look  
18 at all of the networks, stations from all of those four  
19 networks. And we would do some analysis of that data, and  
20 we would use the model with data from all those sources.

21 Also, we focussed in California and Florida.

1 Those are the two primary areas where this product could  
2 be used. It could be used in many other areas, but for  
3 the purposes of this case study we decided to focus on  
4 these two areas.

5 A model can be run with any meteorological data  
6 set you create. But for the purposes of the case study,  
7 we decided to focus on California and Florida.

8 This is a map of California. You may not be  
9 able to tell, but the Bay is up here, San Francisco and  
10 Los Angeles is down below the map. You see we have some  
11 coastal stations, a CIMIS station in Monterey, an NWS in  
12 Santa Barbara, a CIMIS station in Ventura.

13 Then there are also stations in the inland  
14 regions, which are also key growing areas. We're lucky to  
15 have both, a CIMIS and an ASOS station in Merced and then  
16 a Fresno -- and a CIMIS station that are very close to one  
17 another in Fresno, CIMIS and NWS station. And a  
18 Bakersfield station, ASOS station.

19 This is a map of Florida, obviously. We have  
20 stations here from the FAWN network. We have stations in  
21 Tavares, Dover and Bradenton. We asked some agriculture

1 extension people what the best FAWN stations would be to  
2 represent these sort of applications. And they thought  
3 those three were good.

4 Winter Haven is an ASOS station, also in a  
5 growing region. Fort Myers, I'm not sure that there is  
6 much growing down there, but it is another -- it is one of  
7 the only NWS stations that is really close to something  
8 we're interested in. And also Tallahassee, there is some  
9 growing up there.

10 I'll talk briefly about how we process this  
11 data. The data come in all sorts of formats. And you  
12 need to put it into an ISC compatible format. You also  
13 need to estimate what is called the stability class.

14 The ISC input file includes the temperature,  
15 ambient temperature, the wind speed and the wind  
16 direction. All of these stations provide that. You just  
17 need to reformat the data. But you do need to calculate  
18 the stability class using the data that you have.

19 For the National Weather Service station, we use  
20 what is called Turner's method. Basically, we used an EPA  
21 program called PCRAMMET. Turner's method is the most

1 commonly used method.

2           It's basically the stability class is a function  
3 of the wind speed whether it is daytime or nighttime, the  
4 solar angle and the cloud cover.

5           We also used Turner's method, we had all the  
6 available data to do that for the ASOS network. The  
7 PCRAMMET program wasn't designed to look at the ASOS data  
8 or to use it in that format. So we just took the PCRAMMET  
9 code and converted it into our old program to reprocess  
10 the ASOS data.

11           For the FAWN, there is no cloud cover data. So  
12 we used a method called solar radiation delta T. It is  
13 basically a method where you are looking at the difference  
14 in the temperature between two and ten meters, which is  
15 collected at the FAWN stations to get a measure of  
16 stability. That is also an EPA recommended method in  
17 their meteorological guidance.

18           For CIMIS, California DPR has already processed  
19 that data. And we requested it, and they gave it to us.  
20 That uses a method called sigma theta, which is the  
21 standard deviation of the wind direction.

1           That is also an EPA recommended method to  
2   calculate stability classes. And it gives a good measure  
3   of turbulence.

4           So how did we select stations to use for our  
5   case study?

6           It was prohibitive to run the model we thought  
7   and analyze the results for all the possible stations that  
8   are out there. So what we did is we picked 15 stations  
9   among these four data networks and used flux data from  
10   Oxnard and Manteca, two of our studies that were analyzed  
11   in an earlier date, and used the five acre field.

12          We just ran the PERFUM model for those 15  
13   stations in two different flux studies. And we used five  
14   years of data for each station, except Santa Barbara where  
15   there was only three years of data available.

16          And we chose a representative set of four  
17   stations from those 15 as I'll show in a moment.

18          Some observations about the results with the  
19   different stations. Generally, the NWS stations gave the  
20   lowest buffer zone estimates followed by ASOS, CIMIS and  
21   FAWN.



1           Although, when you look at NWS, ASOS and CIMIS,  
2   the difference wasn't that large. The FAWN stations are  
3   an outlier on the high end. And I haven't completely  
4   figured out why. One of the reasons is it has relatively  
5   low wind speeds at least at some of the stations.

6           There is not a predictable difference by  
7   location if you look at inland versus coastal or  
8   agricultural versus urban area. Although with 15 data  
9   points, it's a relatively small data set. So we can't  
10   draw any broad conclusions.

11           These are the results where we calculated the  
12   buffer zones with each of the 15 stations. And I just  
13   took the average of Manteca and Oxnard and plotted it from  
14   highest to lowest.

15           When it says percentile, the highest buffer zone  
16   result was for Tavares. With 15 stations, that's about  
17   like the 91st or 2nd percentile. You see here the three  
18   highest buffer zone results were with the FAWN network.  
19   Dover, Bradenton and Tavares.

20           Followed by that, you see a mix of ASOS, CIMIS  
21   and the National Weather Service. When we had comparable

1 stations like a station in Merced with CIMIS and ASOS, we  
2 got very similar buffer zone results. When we had CIMIS  
3 and NWS close together in Fresno, again, we had very  
4 similar results.

5 So we were relatively confident that we were  
6 getting repeatable -- relatively similar results between  
7 these different meteorological networks except for FAWN.

8 But to be conservative, we decided to consider  
9 -- we would include FAWN at least for this case study  
10 analysis. What we did is we divided this profile up into  
11 core tiles and picked one station shown in blue for each  
12 of the core tiles. So for the Bradenton FAWN station, the  
13 Ventura CIMIS station, the Tallahassee NWS station and the  
14 Bakersfield ASOS station, we chose those four stations for  
15 the complete analysis that we'll present.

16 And the buffer zone estimates you got -- the  
17 average of those four stations was very similar to the  
18 average for all 15 stations. So we feel confident those  
19 are representative of the 15 stations that we have. And  
20 we have two stations in Florida, Bradenton, Tallahassee  
21 and two in California, Ventura and Bakersfield and a mix

1 of inland and coastal sites.

2 After all that prep work, let's talk about some  
3 of the results of the case study analysis that we did.  
4 How did we go about this? We did model runs. At the time  
5 that we did the submission, we had six field studies  
6 available, two raised bed, Oxnard and Plant City, two flat  
7 fume, Manteca and Watsonville, and two drip irrigation  
8 studies, Camarillo and La Selva Beach.

9 We chose four meteorological stations, I just  
10 showed a couple slides ago, Ventura, Tallahassee,  
11 Bakersfield and Bradenton. And we have five field sizes  
12 that we're interested in looking at that range from the  
13 possible field sizes that could be out there in  
14 agriculture, 1, 5, 10, 20 and 40.

15 So that equals total number of runs of six times  
16 four times five for 120 different model runs. And like I  
17 said, we were able to do this in a couple evenings of  
18 computer work.

19 We weren't there during the evening. It  
20 actually ran while we were sleeping.

21 This shows a distribution of buffer lengths for

1 a five acre field using the Manteca flux rate and the  
2 Tallahassee meteorological data.

3 And I have given the distributions for what we  
4 have defined earlier as the whole field and the maximum  
5 concentration shown in red.

6 You see for the whole field distribution up like  
7 that. There is this discontinuity here because it is very  
8 difficult to estimate the buffer zones when you are in  
9 very small distances from the field. So what the model  
10 does, when it is very close, it makes a decision as to  
11 whether it is zero or whether it is 20 meters. And it  
12 doesn't calculate anything in between.

13 But since we're looking at upper percentiles,  
14 that's not going to affect things. So if you look at the  
15 '95th percentile here, you are up around a little more  
16 than 500 or 600 feet buffer zone.

17 So if I were to choose the 95th percentile of  
18 the whole field distribution, that's what I would get for  
19 a buffer zone.

20 If I were to choose the maximum concentration  
21 distribution and look up at the 95th percentile, I would

1 get a buffer zone of about 1300 feet or more than twice as  
2 much.

3 So what sort of factors influence these buffer  
4 length estimates? Obviously, we're going to show the flux  
5 rate or the type of application which influences the flux  
6 rate is going to affect the result you get.

7 The meteorological data, we want to look at  
8 whether we're getting different results with different  
9 meteorological stations.

10 Then one of the things we found in this analysis  
11 is that the diurnal profile was very critical in  
12 determining what buffer zone estimate you would get, in  
13 particular, when the application started. Of course, the  
14 field size is important.

15 Now, as I said earlier, the model outputs this  
16 maximum concentration distribution and the whole field  
17 distribution, the whole distribution. But for the  
18 purposes of discussions, I need to choose something to  
19 present some results.

20 So what I did just for discussion purposes is I  
21 defined the buffer zone as the 95th percentile of the

1 whole field distribution for the slides that will follow.

2           So this first slide shows some of the ranges and  
3 buffer lengths for different meteorological stations. So  
4 I have my six field studies here on the X axis, and the  
5 buffer length I got with a five acre field using those  
6 different meteorological stations.

7           That shows the mean, and these bars show the  
8 range. So if you look at the coefficient of variation, it  
9 was about 13 percent among different meteorological  
10 stations.

11           So if I am looking at the same flux profile and  
12 I'm interested in what the variability is across different  
13 meteorological stations, it came out to about 13 percent.

14           So there are some differences, but it is not  
15 that large. I probably should have put this from zero to  
16 900. It may over-exaggerate the length of those bars.

17           This is somewhat obvious. But if you have a  
18 larger field study or a larger field size, you are going  
19 to have a larger buffer zone. So you see with the one  
20 acre field, the buffer zones are down 250, 200 feet. For  
21 a five acre field, they are rising. Then all the way to a

1     40 acre field for this case study you are looking at  
2     buffer zones around 2000 feet or more.

3             As I said a moment ago, the diurnal profile  
4     turned out to be a key factor in determining the buffer  
5     zones when we used the PERFUM model, and we're actually  
6     accounting for that diurnal profile.

7             Let's look at an example with the raised bed.  
8     We had two raised bed application. One in Oxnard and one  
9     in Plant City, Florida.

10            The Plant City study is shown in the red. That  
11     study, if you look at the X axis on the top, started at --  
12     well, it actually started at about 7:30 a.m. and finished  
13     at about 9 a.m. Because there were some other activities  
14     going on, the measurements began after the application was  
15     complete.

16            You see some of the emissions. Started kind of  
17     low, then it went up around 1 to 4 p.m. And then after  
18     about 4 p.m., they dropped and pretty precipitously. So  
19     you saw a strong diurnal profile there with the emissions  
20     after about 4 p.m. dropping to a significantly lower level  
21     than they were earlier for this early start application.

1           If you look at Oxnard, you have the blue dot  
2   showing the start of the application at 12. And it  
3   finished at about 8:30 p.m. This is unusual for typical  
4   field practice. But because of all the other associated  
5   activities that are required to do these measurements, it  
6   just ended up for this day the application started late,  
7   later than expected, and it finished later than expected,  
8   much later than expected.

9           So what was the impact of that? It is kind of  
10   interesting. You had some emissions going up here shown  
11   on this green line. But then at 8 p.m. when the  
12   application ended, you still have a lot of the mass in the  
13   field, and you still had some high emissions from this 8  
14   p.m. all the way through 5 a.m.

15           So for the Oxnard situation, you had a lot of  
16   the emissions occurring during that more stable nighttime  
17   period where you have higher stability and lower wind  
18   speed and it is less conducive to dispersion.

19           So what impact did that have on the buffer  
20   estimates that we would get?

21           This shows the buffer lengths for the whole



1 field distribution for Oxnard and Plant City.

2 Let me go back and mention. The mass that came  
3 off of the Oxnard study and the mass that came off the  
4 Plant City study over the first 24 hours was virtually  
5 identical. About the same amount of mass on a percentage  
6 basis. The only real difference is this diurnal profile.

7 Now, if we were to look at the 95th percentile  
8 of the whole field distribution, with Oxnard, we would get  
9 a buffer zone of 860 feet. With Plant City, we would get  
10 a buffer zone of 485 feet, almost a twofold difference in  
11 buffer zones just as a result of that diurnal profile for  
12 the same mass emissions.

13 This shows the drip irrigation applications. It  
14 is not quite as dramatic, but you still saw the diurnal  
15 profile explain some of the differences in the buffer  
16 length estimates that we got.

17 This is the Camarillo study. It started at 8  
18 a.m., finished around noon. And the red bar shows the  
19 flux peaked at around noon to 3 p.m., and then dropped  
20 pretty precipitously after 3 p.m.

21 At La Selva Beach, it was a little bit

1 different. The application started at 12 p.m., finished  
2 at 6 p.m. And you had a peak emission up here that's  
3 between 6 p.m. and 8 p.m. So you had some very high  
4 emissions still in that early evening hour, even up to  
5 about 9 p.m.

6 So for La Selva Beach, you had higher emissions  
7 during the early evening period than Camarillo.

8 And you saw some difference in the buffer length  
9 estimates. Even though Camarillo had higher emissions, 50  
10 percent of the mass came off in the first 24 hours for  
11 Camarillo whereas La Selva Beach it was only 42 percent.

12 So even though it had higher mass emissions  
13 because of the diurnal profile, the buffer zone at  
14 Camarillo was 480 feet and the buffer zone at La Selva  
15 Beach was 650 feet. So despite that difference in  
16 emissions, that diurnal profile made up for that and a  
17 little more.

18 As I said, we also output the monthly variation  
19 of the buffer zone in the PERFUM model for any user who  
20 wants to do a seasonal analysis.

21 So looking at the buffer length for four

1 different meteorological stations in the Manteca data set  
2 for a five acre field, we see -- it is not always  
3 consistent, but a general profile where you see this dip  
4 during the kind of summer period, maybe April through  
5 August.

6 And you have the highest buffer zones in the  
7 January, February, November, December area.

8 From a meteorological standpoint, that is what  
9 you would expect. You have generally more stable  
10 conditions during the winter, lower wind speeds. You have  
11 a shorter daytime period during the wintertime. So you  
12 have a longer nighttime period where the conditions are  
13 more stable.

14 So that wasn't surprising. And it shows that  
15 these buffer zones can be -- let's look at Tallahassee  
16 where you had more than an 800 foot buffer zone during  
17 January and then about a 500 buffer zone for -- that would  
18 be July. You get a pretty significant difference in some  
19 cases if you just look at the seasonal variation.

20 This just shows another plot showing more or  
21 less another profile, but for the Camarillo study.

1           As I described earlier, we have two programs,  
2   really, PERFUM and PERFUM MOE. The PERFUM MOE program is  
3   what we call a risk management tool. It is used -- once a  
4   user establishes a buffer zone that they are interested in  
5   or that they may have set from PERFUM or they just may  
6   have a general interest in knowing, you can use the PERFUM  
7   MOE program to calculate the distribution of margins of  
8   exposure for the locations around that perimeter.

9           This shows a profile for Oxnard for a five acre  
10   field as an example.

11           For the blue line, I have chosen the buffer zone  
12   that I got from the 95th percentile from a PERFUM run  
13   using the whole field distribution. So that turned out to  
14   be 860 feet.

15           I ran that through the PERFUM model. And as  
16   expected, if I look at the 95th percentile, it is about  
17   100. Because of the geometry, the calculation, it could  
18   be a little different than that. But the goal here is to  
19   have -- at the 95th percentile, to have a margin of  
20   exposure of 100.

21           Then you can see at the lower percentiles, 90,

100

1 85, 80, what the margins of exposure there are, and then  
2 you could also see above the 95th percentile what sort of  
3 margins of exposure you are potentially having.

4 So at the 97th percentile, you have a 74 fold  
5 margin of exposure. And all the way up to the 99.9 you  
6 have a 24 fold margin of exposure. That is showing that  
7 even at the 99.9 percentile your exposure is still 24 fold  
8 below the human effect, the human equivalent no effect  
9 level.

10 The red bar shows what it would be if I chose a  
11 buffer zone from the 90th percentile, and you see  
12 proportionately lower numbers for that.

13 I have one other example of a margin of exposure  
14 curve here. This shows for Manteca a 10 acre field. The  
15 blue line shows what it would be for a buffer of 545 feet.

16 The red line shows for 436 feet buffer. That's from the  
17 90th percentile.

18 Again, it is about 99 -- margin of exposure at  
19 the 95th percentile as we have determined from the  
20 calculation. And then upwards to the 99th percentile it  
21 was about a 36 margin of -- margin of exposure, about 36.

1           I would submit that this is a flatter margin of  
2 exposure curve than you see in most environmental  
3 circumstances. Say if you are looking at a worker  
4 exposure to a mixer loader or something like that where I  
5 think that upper tail in my experience has been much  
6 higher, where the difference between the 95th percentile  
7 and the 99th percentile could be very large. You see that  
8 with looking at like drinking water concentrations or  
9 something like that.

10           So I think the relative flatness or how ever you  
11 want to interpret this margin of exposure curve, I think  
12 it is very useful information that could be used by risk  
13 managers to determine a safe buffer zone.

14           Part 6, the uncertainty analysis, we wanted to  
15 look at some various sources of uncertainty in the model.

16       Obviously, every model has uncertainty. Every  
17 measurement has uncertainty. And we're using various  
18 measures of flux rate and meteorology and whatnot in our  
19 model as inputs.

20           This just lists some of the key areas of  
21 uncertainty. The estimation of flux rates. And I'll talk

1 about that a little more in a moment in detail.

2 The characterization of meteorology in the  
3 growing areas. I think we have uncounted for that  
4 uncertainty relatively well by looking at different  
5 meteorological sources and also using five years of  
6 historical meteorological data, which is the EPA standard  
7 when you do permitting applications. So we're looking at  
8 a relatively long period of time.

9 The air dispersion estimates. Obviously, there  
10 is uncertainties inherent in the dispersion models. I  
11 talked about those in the report. And I think for this  
12 particular application and area source where we are  
13 looking at concentrations very close to the source,  
14 relatively speaking, compared to a point source where you  
15 might be looking miles away, I think those estimate are  
16 comparatively speaking better.

17 Indoor versus outdoor exposure. We don't know  
18 what the indoor exposure is relative to the outdoor  
19 exposure, but we have some idea that it could be lower.  
20 We just want to look at what the impact of that could be.

21 Time activity assumptions. We're assuming for

1     our calculations that somebody -- that a person is at the  
2     perimeter of the buffer zone for 24 hours. That may not  
3     be true at least for most people. So we want to look at  
4     the impact of that assumption.

5             The potential for exposure from multiple fields.

6     I have some additional analysis which I have done on  
7     that, which I will show.

8             And we have looked at the variation of exposure  
9     and application likelihood by season in some of the graphs  
10    I have shown you earlier with the variation by month.

11            Let's talk about the uncertainty in flux rates  
12    in a little more detail. There are two components to this  
13    uncertainty.

14            There is a measurement uncertainty for  
15    individual studies. So when we do a particular flux  
16    study, we derive an estimate of those flux rates. But  
17    there is an error associated with that estimate. And we  
18    have modeled that using the standard error explicitly in  
19    the PERFUM model. So I think we're explicitly accounting  
20    for that error in the model.

21            But there is also variability across studies



1 based on field conditions, such as temperature, soil type,  
2 organic matter content, et cetera. There is a lot of  
3 different things that could happen out in the  
4 environment.

5 We have seven field studies. We have a pretty  
6 tight range. We feel pretty good about that. But there  
7 could be additional variability that is out there that  
8 would show up if we were to do more studies.

9 So what we did is we wanted to look at that  
10 variability by using the larger database for  
11 methylbromide. For methylbromide, there has been maybe 30  
12 to 50 studies that have been conducted. A lot of those  
13 have been conducted and summarized by DPR.

14 So the first uncertainty scenario, we went ahead  
15 and looked at the methylbromide data summarized by DPR.  
16 We came up with a coefficient of variance of 47 percent  
17 among the different studies.

18 Instead of using the coefficient of variance  
19 from our flux studies, we used the coefficient of variance  
20 of 47 percent in PERFUM to model that variability.

21 And you would expect when you increase the

1     variability around a mean estimate, you would expect  
2     higher buffer lengths at the upper percentiles. And we'll  
3     show that in a moment.

4             Now, even though we're assuming a higher  
5     uncertainty in scenario one, we're still assuming the same  
6     flux rate that we had in our studies.

7             The second scenario we decided we would choose a  
8     higher flux rate than we actually measured. And what we  
9     did is we went to the methylbromide data and we calculated  
10    a 75th percentile flux rate based on the coefficient of  
11    variance here for methylbromide and the mean flux rate  
12    that we have from our own studies. So let's look at the  
13    results for that.

14            The red bar just shows a normal PERFUM run. And  
15    the yellow bar shows the scenario one where we used the  
16    higher coefficient of variance. And the blue or, I don't  
17    know what you would call that, a green bar shows scenario  
18    two.

19            You had a very small difference between the  
20    first, the normal run and the scenario one until you got  
21    to maybe a 40 acre field where you saw some differences.

1 So increasing that coefficient of variance had some  
2 effect on that upper percentile buffer zone, but it wasn't  
3 that dramatic.

4 Now, when we did scenario two where we assumed  
5 actually a higher flux rate than we actually measured, we  
6 saw upwards of, I don't know -- you see here for a 20 acre  
7 field where it is about maybe an 1,800 foot buffer zone  
8 compared to a 1,500 foot buffer zone.

9 So it is expected if you put in a higher flux  
10 rate you get a higher estimate for the buffer length.

11 We feel pretty comfortable with the studies we  
12 have, particularly, as we added the Guadalupe study that  
13 was conducted after the submission.

14 We have a pretty tight range. So we consider  
15 this sort of a bounding analysis on the flux rate as to  
16 what kind of variability you might expect on that.

17 Indoor exposure and time activity. As I said  
18 several times, the buffer zone calculations assume the  
19 person is at the perimeter for 24 hours and that they are  
20 either outside for 24 hours or if they are indoors they  
21 are getting the same exposure as they would when they are

1 outdoors.

2           At least combined together, these likely  
3 represent rare circumstances, but they can certainly  
4 happen.

5           One approach you could have is you could try to  
6 build that explicitly into the model with probabilities.  
7 Maybe you could calculate a distribution of exposures  
8 based on distribution of time activity data.

9           But in starting to think that through, there are  
10 some challenges to doing that. We know that most people  
11 will not spend an entire 24 hours at the perimeter of the  
12 buffer zone.

13           But actually quantifying the amount of time that  
14 they would be away from the perimeter and maybe -- if they  
15 weren't at the perimeter, they could still be exposed by  
16 being somewhere else downwind of the application. It was  
17 a pretty difficult thing to quantify.

18           From the chemical characteristics and general  
19 experience, we know that indoor exposures are generally  
20 lower. But in this case, chemical specific data are  
21 unavailable. So we didn't feel justified in explicitly

1 incorporating that into the case study for iodomethane  
2 either.

3 But we wanted to do some uncertainty analysis to  
4 look at what are the potential effects of these variables.

5 We did four alternative scenarios. For the  
6 first one, we just looked at the indoor and outdoor ratio.

7 We assumed from 8 p.m. to 8 a.m., the nighttime period, a  
8 person is indoors with an indoor to outdoor ratio,  
9 concentration ratio of .7.

10 So if we have 100 micrograms per meter cubed  
11 outside, we're assuming 70 micrograms per meter cubed  
12 inside.

13 For alternative scenario two, we did the same  
14 thing except we assumed a lower I/O ratio of .3.

15 For alternative scenario three, we looked at  
16 time activity. We assumed someone is away from the  
17 influence of the field during a normal workday period,  
18 from 8 a.m. to 5 p.m.

19 So this person may live in a residence that's  
20 near a field, but from 8 a.m. to 5 p.m. they go off-site  
21 and are not exposed.

1           Alternative scenario four, we combined the two.

2       We looked at someone who is away from the influence of  
3       the field from 8 a.m. to 5 p.m., and that they are indoors  
4       from 8 p.m. to 8 a.m. with an I/O ratio of .5.

5           So they are only outdoors from 5 p.m. to 8 p.m.

6       That's not that unusual. Most people spend about 90  
7       percent of their time on average indoors.

8           What are the results. This shows an example.  
9       We used the Manteca data with a five acre field and the  
10       Ventura meteorology.

11           For the maximum scenario, we're calling sort of  
12       the normal scenario, we got a buffer zone of 646.

13           When we assumed an I/O ratio of .7, it dropped  
14       down to 597, a little bit different. When we assumed an  
15       I/O ratio of .3, it went down to 545. So a little bit  
16       different too.

17           You get a more dramatic difference when you  
18       assume that someone is away from the field from 8 a.m. to  
19       5 p.m., about 423, about a third less.

20           And then finally, if you assume both someone is  
21       away from the field for the workday and they have an I/O

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1 ratio of .5, you get a buffer zone that's even less than  
2 half.

3 So this looks at some of the potential  
4 variability associated with making those conservative  
5 assumptions.

6 Multiple applications. In all the scenarios we  
7 looked at so far, we're assuming one field is emitting it  
8 at one time. We're not looking at the impact, you know,  
9 what is the potential impact of different fields emitting  
10 at the same time and those plumes from those fields  
11 overlapping.

12 So the first scenario we looked at, we assumed  
13 that different fields were being applied in the same  
14 vicinity and that those applications all occur at the same  
15 time.

16 That's really, I think, a low probability event  
17 where you would see fields worsening 1,500 feet apart. I  
18 think that's a low probability event that you would  
19 actually see that occurring, two different fields being  
20 applied at the same time and they are 1,500 feet apart,  
21 but it could happen.

1                   So it is a low probability event. But there is  
2   not a clear basis to establish the probabilities of it  
3   happening. So we'll just look at it from a worst case  
4   basis.

5                   The second scenario, which is more common, is  
6   when a grower applies a large field in different sections.

7                   And we have assumed four quadrants. So you have  
8   a 20 acre field and they apply in four different quadrants  
9   on either four consecutive days or with a lag period  
10   between applications.

11                  This is actually a common practice for large  
12   fields in agriculture. And the methylbromide regulations  
13   require a lag period between applications to mitigate  
14   these effects.

15                  So this is our multiple field scenario where all  
16   the fields are emitting at the same time. And these are  
17   generally supposed to be different farms. This is the  
18   principal application we're concerned about, and we  
19   assumed that there were additional applications at these  
20   four corners.

21                  Again, I think that would be a pretty rare



1     circumstance, but for the purposes of this analysis, we  
2     developed a PERFUM scenario just to look at this in a  
3     worst case basis.

4                 We're also assuming one acre fields. The result  
5     we get -- we might want to look at this for larger fields,  
6     we might get a different result.

7                 But for an one acre field, we got a relatively  
8     small impact. The buffer zone was about 270 feet when you  
9     have all the fields emitting at the same time whereas the  
10    buffer zone was just less than 250 feet for the first  
11    source only.

12                And if you assume that the buffer zone was set  
13    modeling something like this, this is the potential  
14    discrepancy that you might have.

15                Now, the reason that this isn't a large impact  
16    is in my view is that say the wind is moving in this  
17    direction. You are going to have the plume from the  
18    original field right here, whereas the only other field in  
19    this case that could impact that significantly would be  
20    this one. And it is relatively far away.

21                So just a matter of the geometry of the

1 situation and the fact that the wind is going to be  
2 blowing in one direction at one time. I mean, you are not  
3 going to have a situation where this field is blowing this  
4 way and this one is blowing that way.

5 We're assuming these are relatively close  
6 together and they are influenced by similar meteorology.

7 So let's look at the more common situation in  
8 agriculture. Where you have multiple fields say a 20 acre  
9 field and you apply the field by quadrants. We developed  
10 another PERFUM scenario that you could look at this issue.

11 Say today I'm doing an application on this  
12 quadrant. A day ago, I applied this one. Two days ago,  
13 that one. Three days ago, that one. And the wind  
14 direction is going this way. So my impact is going to be  
15 right along this way.

16 And all three of these fields, these three other  
17 quadrants could potentially contribute to the  
18 concentrations right there. So let's look at what we get  
19 there. Also, let's look at what we get when we have a one  
20 day lag.

21 In the first scenario I used, I just used the

1 emission profiles that we had for the three days after the  
2 application in the field study.

3 In this second scenario, I chose the profiles  
4 that we had two days ago, four days ago, and six days --  
5 well, that would be two days after the application, four  
6 days after the application and six days after the  
7 application. And those are lower numbers than the first  
8 three days.

9 So if you have no lag -- let me back up. It is  
10 difficult to define the PERFUM whole field distribution  
11 for this example. So what we did is just looked at the  
12 maximum concentration buffer lengths.

13 And when we had no lag between applications, we  
14 had a difference of almost, well, about 200 feet. You  
15 have about 1100 feet and 1300 feet there. The blue bar is  
16 when you had just the first source and the red bar is when  
17 you have all sources. If we incorporated a one day lag,  
18 we saw virtually no difference in the emissions.

19 So this model, this scenario could be used to  
20 look at that issue, look at how much of a lag time you  
21 need before you mitigate the effects between different

1 fields.

2           And the model in this case will output both --  
3 it will output the distribution simultaneously for what  
4 you would get for all sources and what you would get for  
5 only the first source, only the main source you re  
6 interested in. So you could immediately compare the two.

7           Finally, I want to get on to some conclusions.  
8 We have described the PERFUM model and briefly summarizing  
9 its capabilities. We can estimate buffer lengths for a  
10 whole field and maximum concentration distribution, which  
11 we have talked about. We can probabilistically treat  
12 emission rate uncertainty.

13           We have two additional scenarios that can be  
14 part of the PERFUM package where we can look how to model  
15 multiple application rates. We can calculate the buffer  
16 length distribution on a monthly basis to use for seasonal  
17 analysis.

18           We have an additional program PERFUM MOE which  
19 can estimate the distribution of MOEs at any proposed  
20 buffer zone that someone is interested in.

21           We can also account for multiple applications.

1 That just repeats what I said above there.

2 Some of the lessons learned from the case study.

3 The diurnal profile of emissions is an important factor  
4 in estimating the buffer lengths. If you want to have the  
5 lowest 24 hour exposure concentration, you generally want  
6 to have your application start as early as possible, early  
7 in the morning as possible.

8 The margin of exposure curve is relatively flat.

9 I think flatter than most other -- many other  
10 environmental circumstances.

11 Multiple exposures may not have a large impact,  
12 particularly, if there is a one day lag between  
13 applications. I would still consider the multiple  
14 exposure stuff we did as kind of a prototype.

15 We might want to do some more analysis on that.

16 But what we have done so far shows that if you have a one  
17 day lag you are going to significantly mitigate the  
18 impacts.

19 So that concludes my presentation. I look  
20 forward to questions and discussion.

21 DR. ROBERTS: Thank you, Dr. Reiss, for a very

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1 thorough presentation. I suspect that the panel will have  
2 some questions for you.

3 So let me open it up to questions right now.  
4 We'll start with Dr. Hanna.

5 DR. HANNA: My question is about the situation  
6 when you have calm wind conditions. How did you read  
7 that? Did you follow the exact approach to using the  
8 ISCST-3 model or different?

9 DR. REISS: Yes. We used the calms processing  
10 routine that we had in the ISC model. I would also note  
11 that, in several of the field studies we did where we're  
12 estimating the flux rates, there were calm conditions.

13 It was not an infrequent occurrence in the field  
14 studies that we had. So essentially, the flux rates we  
15 have estimated at least in some cases are kind of  
16 calibrated to those sort of calm conditions.

17 DR. HANNA: I have a second question quickly.  
18 Are you able to identify or quantify the uncertainty in a  
19 kind of real conclusion? Are we off of a factor of two or  
20 40 percent or 10 percent? Is this can be concluded? Are  
21 you able to get this measure based on your analysis?

1 DR. REISS: There is a lot of literature about  
2 the uncertainties in dispersion models. And I think it  
3 rarely comes to any definitive conclusion in what I have  
4 read.

5 Some people say dispersion models have a factor  
6 of two variability. I think for the particular  
7 circumstance that we have where you are looking at an  
8 area source, you are looking at estimating concentrations  
9 very close to a field, you essentially have flux estimates  
10 that go into the model that are sort of calibrated to the  
11 model.

12 That may not be the best circumstance from a  
13 purely scientific standpoint in terms of developing a  
14 phenomenologically correct model, but from a regulatory  
15 standpoint I think there are some advantages there.

16 I mean, we're using flux rates that we  
17 essentially got from concentration estimates downwind. And  
18 in some cases, in worst case conditions.

19 I'm reluctant to put a number to the  
20 uncertainty. I think that's just something in dispersion  
21 modeling that's not settled. But I think it is

1 substantially less than a factor of two.

2 DR. HANNA: Thanks.

3 DR. ROBERTS: Dr. Seiber.

4 DR. SEIBER: My questions are more on how this  
5 will be used in practice. Let me see if I understand it  
6 right.

7 My understanding is that a permit would be  
8 applied for if a fumigation was to take place let's say  
9 two or five days subsequent. And in the permit  
10 application, a given acreage would be specified, certain  
11 size, certain application type, rate and so forth.

12 And then PERFUM would be used to calculate a  
13 buffer given that input. This is still predicting, in  
14 other words, what might be a protective zone around that  
15 application that's going to take place a couple days in  
16 the future.

17 I just want to see if I'm on the right track  
18 here.

19 DR. REISS: What PERFUM will do, I mean, I can't  
20 speak for how EPA in California will eventually regulate  
21 these things or how California is currently doing it.



1           But, for example, with methylbromide, the DPR  
2   model is used to develop essentially a table that growers  
3   can look up based on their application method and the  
4   field size. And they can get the buffer zone off that  
5   table.

6           So it is not -- it is used prior to the  
7   application. Those are set probably when -- I expect that  
8   these buffer zones and these buffer zone tables would be  
9   set prior to the chemical being registered. And so the  
10   grower would just go to that to look it up. And then those  
11   would come from PERFUM or some other method.

12           DR. SEIBER: All right. And then kind of  
13   following that reasoning on then, if there were let's say  
14   a residence or a subdivision or whatever within that  
15   calculated protective zone or buffer zone, then the permit  
16   would not be allowed at least under those conditions.

17           Am I still on the right track here?

18           DR. REISS: I might defer to somebody who  
19   actually regulates the chemicals. I mean, I don't think  
20   they are disallowed. I think there are several options  
21   they have.

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1                   Do you want to take that?

2                   DR. SEGAWA: What Dr. Seiber described is very  
3 similar to what we're currently doing in California for  
4 methylbromide, where prior to the application a grower  
5 applicator would go to the local ag commissioner with a  
6 plan, specify a number of acres, application rate and  
7 things like that, and our agricultural commissioner would  
8 specify the size of the buffer zones based on the  
9 information by the DPR.

10                  And if there is a house or school or something  
11 inside the buffer zone, they would have to make changes,  
12 break up the field, for example, into smaller blocks in  
13 order to do the fumigation.

14                  DR. SEIBER: Okay. Again, carrying this kind of  
15 -- I'm just trying to figure out how it would be used and  
16 particularly how it would be validated.

17                  Would there be field data collected for some  
18 number of applications, let's say, 1 out of 10 or  
19 something, and this would probably be up to the state  
20 enforcement agency, I suppose.

21                  And what I'm getting at there is, since this was

1 all a predicted buffer zone based on historical weather  
2 and so forth, would somebody then go back and say here is  
3 what the actual weather was when that application  
4 occurred, and maybe even measure some concentrations out  
5 at the buffer zone and see if, in fact, the buffer zone  
6 had been accurately set.

7 I'm getting to the question of how are you going  
8 to validate and give us some feeling of confidence that  
9 this is going to do the job, so to speak.

10 And I know it is early. And probably this is  
11 down the road. But it would help me understand how you  
12 are going to tell whether it is within 50 percent or 20  
13 percent or whatever level of accuracy you agree is -- what  
14 you are after in a case like this.

15 DR. REISS: I can't speak to what sort of  
16 measurements might be made following the registration of  
17 the products.

18 We did conduct seven field studies where we  
19 measured the concentrations downwind of the field. So the  
20 model is predicting those field studies. This model is  
21 (inaudible) calibrated to predict those field studies.

1                   So we do have measurement data and some  
2   experience with collecting that data and using the model  
3   to accurately predict what happened in that field  
4   circumstance.

5                   DR. SEIBER: But that's a little different. You  
6   are using a field to let's say collect downwind  
7   concentrations. You are back calculating the flux. Same  
8   field. And then you are going to go out and say, here is  
9   what the buffer zone ought to be for that field.

10                  But in reality, here is another field over here  
11   that needs to be fumigated and you don't have that  
12   ability. It is not the same field. It is a different  
13   one. If you follow my logic.

14                  DR. REISS: Sure. And we have done seven  
15   different field studies. So we have some estimate of the  
16   variability that you get between different fields.

17                  I think it was relatively tight distribution.  
18   But we'll account for that variability in the model using  
19   the ability we have to perturb the flux rate.

20                  DR. SEIBER: Anyway, I don't know whether  
21   anyone else can answer questions on this, but maybe Randy

1 would tell us what the actual practice would be.

2 DR. SEGAWA: Randy Segawa in response to Dr.  
3 Seiber's question.

4 What we did for methylbromide in California was  
5 that we have had buffer zones for those types of  
6 fumigations since the early 90s. And when we first put in  
7 the buffer zones, it was based on essentially data from  
8 two fields.

9 As we got more data, and today we have data on  
10 some 30 or 40 fields, we continue to go back, look to see  
11 if the buffer zones that we have prescribed were  
12 protective in those new fields. If not, then we adjusted  
13 the size of the buffer zones accordingly.

14 DR. ROBERTS: I have Dr. Yates, Dr. Wang and  
15 then Dr. Spicer.

16 DR. YATES: First, just to kind of follow up a  
17 little bit on what Jim was saying. There are a number of  
18 methylbromide flux studies that have been reported in the  
19 literature that were obtained using aerodynamic chambers,  
20 but not through this back calculation.

21 If you look at the distribution of, say, the

1 total emissions from those studies, it varies from  
2 somewhere like 30 percent all the way up to about 80  
3 percent.

4 So it seems to me that you could see quite large  
5 variability. And this is for cumulative emissions, which  
6 tends to be a little bit better behaved than, say, a two  
7 hour average.

8 If you look at two hour average type  
9 measurements, they can be all over the place. So I think  
10 that the variability that you are likely to see when you  
11 take data collected on one field and apply it to another  
12 field is going to be very, very high.

13 DR. REISS: Right. That's certainly possible.  
14 And I think when we get down to what actual input we'll  
15 use in the model in terms of a flux rate, we're going to  
16 have to look at the variability that we have had for the  
17 seven iodomethane studies or more when it comes to that  
18 and all the methylbromide data and look at that  
19 variability and decide someone is going to have to make  
20 partly a policy decision as to how high on that  
21 distribution you think you need to be to be protected.

1 DR. ROBERTS: Was there a follow up question,  
2 Dr. Yates?

3 DR. YATES: I have other questions too. I can  
4 get back. It is not related to this one. Maybe there is  
5 some follow up here first that should be addressed.

6 DR. ROBERTS: I know Dr. Wang and Dr. Spicer  
7 both had questions.

8 DR. WANG: This follows along the same lines, so  
9 it is probably a good time to ask.

10 DR. ROBERTS: Go ahead, Dr. Wang.

11 DR. WANG: One of the main conclusions you said  
12 is the timing of application has a significant impact on  
13 the maximum concentration that will occur.

14 I think one of the explanations is that there  
15 are two things occurring. Once you apply the fumigants,  
16 it is usually subsurface. So the physics, the diffusion  
17 process in the soil is that the center mass will move --  
18 has a less -- that process, the time that's going to come  
19 out is less dependent on the ambient atmospheric  
20 conditions than once they reach the air. The same time,  
21 the diurnal pattern is that the temperature change, the

1 pressure, the stability.

2           So these two factors, if they tend to coincide,  
3 you may amplify the concentration. Meaning if you apply,  
4 say, around noon or late afternoon, then just about early  
5 afternoon they may come out. And then that's also the  
6 time now you tend to have either unstable condition or  
7 stable condition.

8           If they come out near the evening, then that's  
9 when you're going to see a very high concentration.

10           But to follow up on that is that you posted  
11 several factors that has uncertainty sources. And that  
12 seems to be the main point. My question is that how do  
13 all these uncertainty sources fit together or how would  
14 you integrate them in your assessment since you cannot  
15 (ph) really just look at one at a time. They tend to  
16 occur all at the same time.

17           So if you do a true risk analysis, you may look  
18 at the most, the worst case scenario. Not just for one  
19 source, but maybe multiple sources could be occurring  
20 simultaneously.

21           Do you see my point?



1 DR. REISS: I think so. Let's take the flux  
2 rate in meteorology, two of the key variables. The model  
3 runs through five years of meteorological data. So it is  
4 for every -- it is basically a time series of -- I think  
5 that comes out to about 1800 days or more.

6 So each day it is choosing a flux rate based on  
7 that variability. So you are occasionally going to have  
8 the worst case meteorological condition happen when you  
9 have the high highest flux rate that you get from your  
10 uncertainty analysis.

11 I mean, we don't just -- as a matter of risk  
12 assessment, we don't want to just simply compare all of  
13 the worst case variables together. We think as a matter  
14 -- it provides more information to do this in a  
15 probabilistic way that provides the actual probabilities  
16 of -- well, an estimate of the actual probabilities of all  
17 these things occurring simultaneously, and then, thus, an  
18 actual probability of observing various concentrations.

19 DR. WANG: This could be all doing a joint  
20 analysis of the different uncertainties. It could be run  
21 --

1 DR. REISS: It is run jointly. The  
2 meteorological variability and the flux rate variability  
3 are run as a joint analysis.

4 I think with multiple applications it would be  
5 desirable to have -- to incorporate that into the model in  
6 a probabilistic way, but I have -- you would have to  
7 choose probabilities for whether the multiple application  
8 occurs -- whether it is occurring on a given day, how  
9 often would a multiple application occur, how close  
10 together would the fields be, in what direction the  
11 impacts would be.

12 I mean, it would matter -- the orientation would  
13 matter relative to the wind speed of whether it would be a  
14 multiple impact. Choosing all those probabilities I think  
15 is very difficult to do.

16 So I don't -- I couldn't explicitly incorporate  
17 it into the model. But we developed these additional  
18 scenarios that people can run to look at the impact of  
19 multiple scenarios.

20 DR. WANG: Currently, PERFUM, the model, does  
21 not join all these uncertainties together in a more

1 integrated manner to assess, say, the risk of maximum  
2 concentration. Does it do that now?

3 DR. REISS: It does. It integrates the  
4 uncertainties in the meteorological inputs and the flux  
5 rates.

6 For the meteorological inputs, it is running  
7 five years of meteorological data. So it is accounting  
8 for the uncertainty or not the uncertainty, but the  
9 variation that you observe in the actual environment by  
10 running five years of data.

11 And by using a statistical approach to  
12 estimating the uncertainty in the flux rate, yes, it is  
13 jointly -- the model jointly accounts for the  
14 uncertainties in both of those variables.

15 DR. ROBERTS: We have Dr. Spicer, then Dr.  
16 Maxwell, then we're going to go back to Dr. Yates and Dr.  
17 Small.

18 DR. SPICER: I was curious if the PERFUM  
19 methodology had been applied to methylbromide. For  
20 example, in the tables for a given application rate at  
21 this point in time you look up an exclusion zone of 1,000

1 meters, then does the PERFUM methodology -- how would that  
2 compare under the same set of inputs for the  
3 methylbromide? Is it longer, shorter?

4 DR. BARRY: You mean for the buffer zones?

5 DR. SPICER: Yes.

6 DR. BARRY: One of our colleagues at DPR did do  
7 a comparison of taking the single maximum distance versus  
8 all the way around the field. And the maximum distance,  
9 of course, would always give you a longer buffer zone.  
10 And also to expand on what Randy Segawa was saying  
11 earlier, we had 34 studies and only two of those studies  
12 had buffer zones shorter than what was modeled when we  
13 measured on site in the field.

14 So we have also done that field assessment, too,  
15 for methylbromide.

16 DR. SPICER: It is quite evident that you have  
17 done extensive work on the methylbromide. That's all I  
18 was asking, trying to ask, was if you applied the PERFUM  
19 methodology to methylbromide, would PERFUM predict longer  
20 distances or shorter distances or roughly the same?

21 DR. BARRY: They would be roughly the same.

1     Because what Rick has done in both -- basically, what we  
2     have done with five years of weather data, after we had  
3     used our standardized weather conditions, see stability in  
4     one point for meters per second (ph), we assembled data  
5     from CIMIS stations in California and did similar analysis  
6     to what Dr. Reiss has done.

7             And the single direction would give similar  
8     measurements to what we would get using PERFUM. And the  
9     multiple direction would give similar results to the  
10    multiple direction. Because he's basically doing the same  
11    process that we did for validating our methylbromide  
12    buffer zones.

13            DR. SPICER: In this process, though, what you  
14    are doing is aren't you using essentially data from remote  
15    locations and applying them to a local dispersion  
16    scenario?

17            DR. BARRY: Yes. That's true.

18            DR. SPICER: So there is no way to account, for  
19    example, for drainage flows or flows that are influenced  
20    by local topology, those sorts of thing?

21            DR. BARRY: No. I would say that we do not

1 account for that, because it is remote from a specific  
2 field. Because you don't know where the applications are  
3 going to occur. But we are using stations that are in  
4 agricultural areas where the applications commonly occur.

5 They are not very far away from -- in fact, some  
6 of them are, actually, in the same geographical area, for  
7 example, Watsonville or Salinas (ph) where much of the  
8 methylbromide is applied or the fumigants are applied.

9 In California, the Central Valley, in some of  
10 the areas where these fumigants are used it is flat and  
11 open. You are going to get pretty consistent  
12 meteorological conditions between areas that could be 20  
13 miles apart.

14 But that, of course, is going to be specific to  
15 a region. Not all parts of the country are going to be  
16 like that, of course.

17 DR. ROBERTS: Dr. Maxwell, then Dr. Yates, then  
18 Dr. Small.

19 DR. MAXWELL: This is Dave Maxwell, National  
20 Park Service. I actually have three questions, but I'll  
21 go one at a time.

1                   How is mixing height addressed? Is it addressed  
2   any differently through ISC or are there multiple sets of  
3   mixing height data used?

4                   DR. REISS: We're talking about concentrations  
5   relatively close to the field, and the plume generally  
6   doesn't rise up to the mixing height within that distance.

7                   What we did is we analyzed -- for the National  
8   Weather Service data, we actually used the mixing heights  
9   that came out of PCRAMMET using upper air data. And then  
10   we did a sensitivity analysis where we changed those  
11   mixing heights, just a nominal value of about 300 meters,  
12   which is what we used for the other data sets, and it  
13   changes the answer by a very, very small amount.

14                  So the mixing height isn't a large factor given  
15   the concentrations are close to the field.

16                  DR. MAXWELL: Thank you. Next question. I  
17   don't know who could answer this. Who are the health  
18   effects of iodomethane?

19                  DR. REISS: What are the health effects. The  
20   endpoint we're looking at is something called late  
21   resorptions in a rabbit study. It is a reproductive

1 outcome, basically.

2 DR. MAXWELL: The last question is how is the  
3 120 micrograms per cubic meter threshold determined? Who  
4 determined it?

5 DR. REISS: It is determined from this rabbit  
6 study. So there was a no effect level. There was a  
7 concentration that the rabbits were exposed where no  
8 effect was found.

9 There are various conversions to take that  
10 concentration and convert it to a human equivalent  
11 concentration. And actually, DPR and EPA kind of  
12 disagree on how to do that. I couldn't explain to you the  
13 mechanics of that, but you basically get a human  
14 equivalent value from that study, from the animal study.

15 That's commonly how -- if you are familiar with  
16 criteria pollutants, there is a lot of human data commonly  
17 with pesticides. You are looking at animal data. That's  
18 why you generally look at 100 fold safety factor, tenfold  
19 to account for inter human variability and tenfold to  
20 account for the extrapolation between animal and humans.

21 DR. ROBERTS: I think Mr. Dawson would like to



1     respond to those last two questions as well.

2                   MR. DAWSON: I want to make it clear that 120  
3     value is really for example purposes. Along with this  
4     work with the exposure models, we're currently in the  
5     process of further evaluating the toxic effects of  
6     iodomethane.

7                   Again, I just remind everybody this is for  
8     example purposes only, and we're still determining the  
9     final numbers.

10                  DR. ROBERTS: Thank you for that clarification,  
11     Mr. Dawson. Dr. Yates I believe is next.

12                  DR. YATES: I have two questions.

13                  The first one, in the studies that you did on  
14     the flux rate, you reported the interval flux as like  
15     percent of the application. Did you go through -- for all  
16     the data that you had, did you sum up the percent losses  
17     to see if they exceeded 100 percent?

18                  DR. REISS: Yes. There were a couple that did,  
19     but not by more than a couple percent. So I mean, within  
20     the experimental variability that you would expect.

21                  DR. YATES: Do you think it is reasonable that

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1     you would have 100 percent loss of the fumigant?

2             DR. REISS: That's what we're observing. In  
3     some of the studies, we didn't see nearly that. But yeah,  
4     I suppose it is possible. Certainly.

5             DR. YATES: And let me move on to the second  
6     question because it may help to answer a little bit on  
7     that, too.

8             You did a direct and indirect comparison. But  
9     you didn't really describe much about how you ran the  
10    study, the direct flux study. Could you kind of just tell  
11    us -- you said it was aerodynamic. But this was a flat  
12    fume with a high density polyethylene tarp?

13            DR. REISS: Yes.

14            DR. YATES: What kind of instruments did you  
15    have out in the field, do you know?

16            DR. REISS: I can't speak to that. I wasn't  
17    present during the study. I really just analyzed the  
18    data. I don't know if there is someone here that can  
19    speak to the instrumentation.

20            MR. GILLIS: My name is Matt Gillis from Trical.  
21    I was involved in setting up the equipment for

1 measurement of direct flux during that study.

2           The equipment used was thermocouple temperature  
3 sensors and anemometers placed at a log gradient above the  
4 surface. Air sampling pumps, air samples were collected  
5 at those same heights. And then that data was entered  
6 into flux calculations for aerodynamic flux.

7           DR. YATES: So the equipment was all put in the  
8 middle of the field. And you had a tarp there. How did  
9 you -- what efforts were taken not to do any damage to the  
10 tarp?

11           MR. GILLIS: A ramp was placed along the glue  
12 seam of the tarp to access the samples. The air sampling  
13 method utilized a wind vane where the air samples were  
14 actually collected upwind of the point of attachment into  
15 the field.

16           So the wind never blew across the mast post. It  
17 was always clean air in the gradient.

18           DR. YATES: Thanks.

19           That's very similar to the kind of studies that  
20 we have done with methylbromide. And under very warm  
21 conditions, you can have fairly high flux rates, but it

1 would still seem that 100 percent is probably -- might be  
2 a little on the high side.

3 So I guess there aren't too many of these  
4 studies that look at the direct and indirect methods for  
5 estimating flux and make a comparison. So they tend to be  
6 of interest to me.

7 But are planning to publish the information at  
8 some point.

9 DR. REISS: We are, yes. One comment I would  
10 make on that. If it is less than 100 percent, say we're  
11 overestimating that, that would be an undesirable  
12 situation. But we still -- our flux rates are calibrated,  
13 essentially, to our model, meaning if that 100 percent was  
14 overestimated, then we overestimated the flux rates.

15 But because we've, essentially, calibrated the  
16 flux rates to the model, there is not a particular bias in  
17 any direction.

18 DR. YATES: Right. I agree about the  
19 calibration for the particular study that you obtain that  
20 information for. But when you try to apply it elsewhere,  
21 then I think where there is mismatches I think can tend to

1 be a bit of a problem in terms of developing buffer zones  
2 at the other place.

3 So it really would be ideal if the indirect  
4 method would give a nonbiased estimate of the true flux  
5 from the field.

6 But anyway, I just wanted more of an explanation  
7 of that study more to interpret your figure. I hope you  
8 do publish it, because, like I say, there aren't very many  
9 studies out there that do that comparison between direct  
10 and indirect method.

11 DR. ROBERTS: Dr. Small, Dr. Bartlett, then Dr.  
12 Baker, then Dr. Spicer.

13 DR. SMALL: I have a very few specific questions  
14 here.

15 In your uncertainty analysis method, you have  
16 sequential calls for the random number generator, which  
17 you then apply, originally to the T distribution, now to a  
18 normal distribution that's described as standard air of  
19 your estimate of your slope, which is used to estimate  
20 your flux.

21 Do you call that random number generator

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1 sequentially, independently? In other words, in one  
2 period you call it, you get a value, and the next one you  
3 have an independent call? And if so, what are those time  
4 steps that you operate over?

5 DR. REISS: We have operated -- we set the  
6 model -- we have periods in the flux study which range  
7 from 2 to 24 hours. And I only choose a random number for  
8 each period.

9 So I perturb the flux for that entire, say, 2 to  
10 12 hour period using one random number call. And the  
11 reason is because if I were to break those into individual  
12 hours, I would lose a lot of --

13 DR. SMALL: But the sequential ones are  
14 independent. If there is one two hour period followed by  
15 a four hour period, those are independent --

16 DR. REISS: Those are independent because they  
17 were independent measurements --

18 DR. SMALL: How many of them are there typically  
19 in one day?

20 DR. REISS: Anywhere from two to four.

21 DR. SMALL: And then when you calculate your

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1 buffer zone, you use a 24 hour average concentration?

2 DR. REISS: That's correct.

3 DR. SMALL: Not a maximum concentration?

4 DR. REISS: Using a 24 hour -- that's based on  
5 the advice we have from the toxicologist that that's the  
6 --

7 DR. SMALL: Using the 24 hour average with  
8 independent calls. Good.

9 Second question. I notice in the report that  
10 you have a couple cases where you miss the early first day  
11 ambient concentrations because you couldn't get set up  
12 quickly enough. So you used the second day values back on  
13 the first day.

14 How many of those cases are being used in your  
15 current model now? Which specific ones have that  
16 correction, second day used for first day?

17 DR. REISS: I can't answer which specific ones  
18 have that correction.

19 I know the Watsonville and Plant City I believe  
20 don't. I think most of the others have at least a few  
21 hours where we had to borrow, as we say, from the second

1 day.

2 DR. SMALL: And back calculate to the first day.

3 DR. REISS: It is still an estimate of the flux  
4 rate over the 24 hours following the application.

5 DR. SMALL: Third question.

6 How did you treat your -- you mentioned level of  
7 detection. When your data for your ambient concentrations  
8 are at or below the detection limit and you use those in  
9 your regression model, do you use the detection limit,  
10 zero, half the detection limit, some other approach?

11 DR. REISS: We use zero, but I have done  
12 sensitivity analysis. Unless you do the log transform  
13 method, it makes no difference at all.

14 DR. SMALL: It is harder to use zero with the  
15 log transform.

16 DR. REISS: That's true. But you have to do  
17 something. That motivated my using the limited detection.

18 But most of the values, I mean, could range up to 10,000  
19 -- I mean, the maximum values generally range up to 10,000  
20 fold more than the limited detection. And they generally  
21 dominate the regression result that you get.



1 DR. SMALL: This is motivating my last sort of  
2 area of questioning. That's on this whether or not to  
3 include the intercept or not. That's on the issue of  
4 background.

5 Any idea for iodomethane or other things whether  
6 or not there is some type of background? Do you have  
7 upwind measurements.

8 DR. REISS: It is virtually zero. There is no  
9 other sources. I mean, right now it is not a regulated  
10 product, not a registered product. So there is not other  
11 fields being applied.

12 DR. SMALL: So you expect it to be zero.

13 DR. REISS: We have in a few studies background  
14 measurements where it is basically zero.

15 DR. SMALL: Thank you.

16 DR. ROBERTS: Dr. Bartlett, then Dr. Baker.

17 DR. BARTLETT: Paul Bartlett.

18 One of one of the questions I have is 24  
19 averaging, the 24 hour averaging. I realize -- relating  
20 to questions about physical chemical, biological  
21 properties of the substance.

1           One is why are we concerned about a 24 hour  
2 exposure? This is the first question. Is this an acute  
3 effect? Does the substance -- is there any bio cumulation  
4 for toxic effects?

5           If it is an acute effect, it seems to me that  
6 what the peaks are in like three, four hours might be more  
7 relevant. We see in the area of particulate exposure  
8 right now it is regulated at 24 hour averaging when what  
9 is relevant is, if it is children or elderly people, if  
10 they are exposed to a peak level for a few hours. So it  
11 seems like in some sense regulation maybe moving to a  
12 shorter time period for exposure.

13           The other question that's related to this  
14 averaging is the physical chemical properties reemission.  
15 You mentioned that methylbromide is sticky indoors. I'm  
16 not sure what this substance is like. That implies a  
17 dermal and other forms of exposure might be relevant or  
18 there might be a lag effect if this does get indoors. It  
19 may be trapped in there for some time.

20           And also, this is an issue for multiple sources,  
21 different time periods. I want a little clarification on

1     that.

2                   DR. ROBERTS:  Not to cut you off, Dr. Reiss, but  
3     I think I understood from Mr. Dawson that the toxicology  
4     of this particular chemical is currently under evaluation.  
5     And presumably, that would include, not only the  
6     endpoints, but the most appropriate dose metric.  But I'll  
7     let Mr. Dawson respond.

8                   MR. DAWSON:  That's correct.  We spend quite,  
9     along with the development of this or consideration of  
10    this exposure model, we're going through quite an  
11    extensive process to define what is the appropriate  
12    duration of exposure that we want to look at.

13                  At this point, all factors are pointing to us  
14    considering a 24 hour interval is the appropriate  
15    averaging time.  But again, we still have some final  
16    decisions to make on that based on the toxicology data  
17    that we have for this particular case.

18                  DR. ROBERTS:  But presumably if another interval  
19    appeared to be more appropriate from your analysis, the  
20    model could value the concentration over a different  
21    interval.

1           MR. DAWSON: Right. And whatever the factors  
2   pointed to is the averaging time that we would want to  
3   look at.

4           DR. BARTLETT: The question with your results is  
5   that diurnal variation is quite high. So if people are  
6   being exposed regularly during the day, for instance, when  
7   it is high and you are using a 24 hour average like of a  
8   sequential application or something like that, I think the  
9   hourly data would be valuable to some extent. Especially  
10   if we don't know how it is going to come out. That that  
11   at some point may be of interest.

12          DR. REISS: Let me comment. One thing on the  
13   diurnal variability. There is a large diurnal variability  
14   in the flux rate. But because the lower flux rates occur  
15   at night, which are more stable conditions, you have a  
16   much lower variability between actual concentrations.  
17   That's one point.

18          The model calculates the hourly data. The only  
19   reason we don't output the hourly data is, one, because we  
20   have been told by the toxicologist that they are  
21   interested in the 24 hours.

1                   If that changes, we can change that. But from a  
2 matter of computational efficiency, it is much more  
3 difficult to get that 24 data summarized. So just as a  
4 matter of efficiency, we have limped it to 24 hours for  
5 now.

6                   DR. ROBERTS: Mr. Dawson.

7                   MR. DAWSON: Just to follow up. Jim Jones  
8 mentioned this morning we're looking at six chemicals in  
9 our analysis of soil fumigants. In some of those other  
10 cases, we're focused again on much shorter durations, for  
11 example, hour type intervals for exposure.

12                   So we're definitely interested in this type of  
13 component with the system looking at the shorter averaging  
14 times.

15                   DR. ROBERTS: Did that answer your question?

16                   DR. BARTLETT: Yes. I just would like to throw  
17 out one other question.

18                   When you screened out temperature as an effect  
19 for some -- you have a lot of unexplained variance between  
20 your different studies. And of course, you have a  
21 history with methylbromide.

1           But there are a lot of meteorological factors  
2   and other known factors and multiple regression or some  
3   other ways to clean it up -- may point to that they may be  
4   significant when you said that they weren't significant.  
5   What are your feelings with that?

6           DR. REISS: That's possible. We're seeing --  
7   actually, there seems to be some dependence with wind  
8   speed. You get a little bit higher flux rate with higher  
9   wind speeds.

10           There is bias in the model, because the model is  
11   less accurate for highly convective conditions, the ISC  
12   model, that is. So we have to consider that when we look  
13   at that conclusion.

14           But we're just getting to the point now where we  
15   have seven data points and we're going to get a few more.

16       So I think it is something that we would want to pursue.

17           In an ideal circumstance, you would be choosing  
18   a flux rate each day that depended on the -- that was a  
19   function of the meteorology. And there is some  
20   explanation for -- the meteorology does explain to some  
21   extent.

1                   But we haven't been able to, and I don't think  
2   they have been able to do it for methylbromide either, to  
3   develop an equation, so to speak, that explains all that,  
4   because it is very difficult to get one of these data  
5   points and there is a lot of variability out there.

6                   DR. BARTLETT: I just throw out one thing from a  
7   recent experience in work on PCBs, is that topography had  
8   a lot more influence than we thought in the difference  
9   between meteorology between the station and the  
10   microclimate of the actual measurement. So that can  
11   account for quite a bit of difference sometimes.

12                  DR. REISS: With the flux studies we did, we had  
13   on site meteorological measurements and very little  
14   topographical differences between the measurements and the  
15   actual field.

16                  But yes, in a real circumstance you would have  
17   that.

18                  DR. ROBERTS: Dr. Baker, followed by Dr. Spicer,  
19   Dr. Portier, Dr. Wang and Dr. Seiber.

20                  DR. BAKER: I need just a little bit of  
21   background on the application. Is the application

1 performed by a trained specialist, some sort of  
2 certification program?

3 DR. REISS: Yes. As you can see from those  
4 pictures that we showed, it is a pretty sophisticated  
5 operation. There are trained companies that do those  
6 applications.

7 DR. BAKER: Is there any potential for  
8 mishandling and subsequent release of the material which  
9 could possibly lead to some lingering background in the  
10 field studies but also might contribute to some exposure  
11 in a real application?

12 DR. REISS: Exposure in the real application,  
13 meaning the exposure to the workers, is also measured  
14 during many of these field studies. And it is not that  
15 we're not concerned about it. It is just a separate issue  
16 that is being dealt with by separate analytical methods  
17 and separate mitigation methods.

18 DR. BAKER: You notice the impact of the time of  
19 the day of the application. And certainly there was a big  
20 difference between several hours of the time difference.

21 In the uncertainty sensitivity analysis, would



1     you comment on including the potential to include maybe a  
2     small one or two hour perturbation as a probabilistic  
3     feature in terms of the start time, because I know  
4     sometimes I intend to start things at 8 in the morning,  
5     but it more likely is 9 or 10, versus, I guess, the  
6     current application.

7             As I say, you would rerun with a new start time  
8     and just have a separate run in those two options of  
9     including that probabilistically versus rerunning the  
10    model.

11            DR. REISS: Right. We have been a little  
12    reluctant to decouple the actual meteorological diurnal  
13    meteorology we observe in the studies and just say that we  
14    would have the same profile if we started several hours  
15    later or several hours earlier.

16            But I think the point you make about at least a  
17    few hours could be something that is worth considering.  
18    And that might help to refine.

19            At some point, from a mitigation standpoint, if  
20    you are going to account for the time of the application  
21    and calculate in the buffer zone, some decisions are going

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1 to have to be made about that. And then that might be a  
2 more appropriate time to consider how to model it --  
3 that's whether it is even feasible to account for that in  
4 mitigation.

5 DR. ROBERTS: Dr. Spicer.

6 DR. SPICER: One of the plots that you presented  
7 was of ambient temperature as a function of the flux. Was  
8 that the air temperature or the soil temperature?

9 DR. REISS: It's the air temperature.

10 DR. SPICER: And so because the soil is covered,  
11 would the soil temperature not tend to be higher after  
12 it's covered with this tarp than the air temperature would  
13 be?

14 DR. REISS: It would be much lower. I have  
15 assumed at the moment that they would correlate, the  
16 ambient and soil temperature would correlate. But I  
17 probably should make that graph for soil temperature as  
18 well.

19 It's a graph we made, I'll admit, the last  
20 couple days. But I think we want to do that with soil  
21 temperature and wind speed as well.

1 DR. SPICER: But it is recorded, though.

2 DR. REISS: It is recorded. We have that data,  
3 yes.

4 DR. SPICER: Although it is not a direct  
5 comparison to this, there is a substantial amount of  
6 literature that has to do with estimating flux from liquid  
7 pools. And that's shown to be proportional to the vapor  
8 pressure, which would be a function of the temperature and  
9 then also the wind speed.

10 DR. REISS: Right. I think if you look at the  
11 physical chemical things that are going on, you could have  
12 a situation where it is volatility limited or a situation  
13 where it is diffusion limited.

14 You can have a rate limiting step. It would  
15 depend on the chemical and the circumstance. You have a  
16 very -- you have a tarp, a plastic tarp that limits the  
17 diffusion.

18 So the likely -- it is quite possible that the  
19 tarp, the diffusions of the tarp is the rate limiting step  
20 in the volatilization.

21 DR. SPICER: So it does actually diffuse through

1 the tarp, then?

2 DR. REISS: That's how it gets out. It diffuses  
3 through the tarp. Actually, some of the diurnal  
4 variability you observe is likely the result of the  
5 diffusivity of iodomethane changing with temperature  
6 during the day and night from the tarp.

7 So the tarp becomes a little more permeable  
8 during the daytime when it is warmer.

9 DR. SPICER: It was discussed earlier with  
10 regard to the flux measurements in the area of the curve  
11 (ph). What was the difference between the smallest  
12 closure of mass, if you will, between the integrated flux  
13 rates and the largest one?

14 DR. REISS: You are talking about the  
15 cumulative.

16 DR. SPICER: Yes.

17 DR. REISS: The highest went up to around 100  
18 percent. I believe we're looking at maybe 70 or 80  
19 percent at the lower end, but I'm not absolutely sure of  
20 that.

21 DR. SPICER: That's certainly encouraging. Let

1 me make sure I understand something else about the flux  
2 measurements. The concentrations were made uniformly at  
3 one to one and a half meter height?

4 DR. REISS: Yes.

5 DR. SPICER: So there was no accounting for any  
6 vertical variation in the concentration, then?

7 DR. REISS: No. We did not account for vertical  
8 variation, except during the direct flux, which is really  
9 -- that's how they did it.

10 DR. SPICER: Sure.

11 Of course, your point is taken that the model's  
12 calibrated for those tests. However, implicit in that you  
13 are looking at the atmospheric dispersion. So the vertical  
14 dispersion coefficients will also depend the  
15 meteorological conditions.

16 You are having to assume by doing that that you  
17 are getting the vertical dispersion coefficient correct.

18 Now, granted, because you have all of these at  
19 one elevation, you are not having to assume that about the  
20 lateral dispersion coefficient. But that's the only issue  
21 that might be of concern to me, is that there is no

1 measure of that vertical concentration distribution.

2 DR. REISS: That was one of the reasons we did  
3 the direct flux calculation, is that we would have an  
4 independent measurement of the flux. And it is only one  
5 data point, but the results were very encouraging. We got  
6 a very similar result.

7 I understand there could be some variability  
8 associated with the vertical dispersion coefficients as  
9 well.

10 DR. SPICER: And that would especially be  
11 important during the calm conditions too.

12 DR. REISS: Right. I would point out, we -- we  
13 had a lot of measurements with calm conditions, a lot of  
14 the field studies we had relatively calm, even during the  
15 day.

16 DR. SPICER: I might not understand that  
17 completely. How did you actually model that, then, with  
18 the ISC since the ISC essentially --

19 DR. REISS: We ran it without calms. We ran it  
20 in the no calm -- we turned off the calms processor to do  
21 that.

1 DR. SPICER: So you've got the calms turned off  
2 when you calculate the flux, but you have data that has  
3 the calms in it.

4 DR. REISS: That's right.

5 DR. SPICER: What importance would that have, do  
6 you think?

7 DR. REISS: Well, if you turn off the calms  
8 processor, you are essentially modeling the concentrations  
9 with a one meter per second wind speed when you have a  
10 wind speed below that.

11 So you are estimating a higher concentration  
12 than you would have, I'm sorry, a lower concentration than  
13 you would have. Let me make sure I get that right.

14 You are assuming a higher wind speed than you  
15 observe when you run it without the calms processing. So  
16 if you assume a higher wind speed, you get a lower  
17 concentration.

18 So I think the way that works is you would  
19 underestimate the flux rate.

20 DR. SPICER: It gets confusing.

21 DR. REISS: It was confusing. Let me think

1 about that. I'll answer it again after the break. How  
2 about that?

3 DR. SPICE: Fair enough.

4 DR. ROBERTS: Dr. Portier is next.

5 Before we go to Dr. Portier, let me fill you in.

6 I think we'll continue to take questions if we have them.

7 But probably no later than 12:30, then we'll break for  
8 lunch.

9 If we still have more questions, at that point  
10 we'll take them up again after lunch.

11 Dr. Reiss, I assume you will be available after  
12 lunch if we continue to have questions?

13 DR. REISS: Yes. I'll be here.

14 DR. ROBERTS: Let's continue with Dr. Portier,  
15 followed by Dr. Wang and then Dr. Seiber.

16 DR. PORTIER: Ken Portier, University of  
17 Florida. Could you bring up slide 11, continuing with the  
18 model?

19 In reading the material -- I understand the  
20 basics behind the model, but can you assure me that this  
21 model has been developed for a height of zero? There was a



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1 lot of discussion and things about the point sources, line  
2 sources, area sources, a lot of what happens at the top.

3 My question is when you move that point source  
4 down to zero, does this model -- has this been calibrated,  
5 validated for the zero height position?

6 DR. REISS: Yes. The model -- actually, EPA has  
7 done some recent validations, because they are developing  
8 a new model AERMOD, to replace ISC. It is not yet  
9 approved. At some point we'll have to grapple with that  
10 issue as to whether to adopt it.

11 But in the course of looking at AERMOD, they  
12 have done comparisons between AERMOD and ISC using the  
13 available tracer study they have. And they have several  
14 area sources that are ground level area sources.

15 And in those comparisons, the predictions are  
16 actually better than four point source. So yes, I think  
17 there -- it has been looked at. There is every reason to  
18 believe it would be better for an area source because,  
19 well, in this case we're measuring concentrations closer  
20 to the field. You are dealing with less meteorological  
21 variability, I think, as you go up in the atmosphere.

161

1 DR. PORTIER: I had a follow up question on  
2 Slide 18. I was trying to understand what the sequence of  
3 the simulation is.

4 Essentially, for each day, for this grid, you  
5 are going to simulate all the concentrations for 24 hours,  
6 average them up. You get an average for each point on  
7 this grid. You can develop this red profile for every  
8 day, 1,825 days, five years. Correct?

9 DR. REISS: Correct.

10 DR. PORTIER: And then to a certain extent you  
11 find the maximum distance to the point. And that's the --

12 DR. REISS: Maximum distribution.

13 DR. PORTIER: -- maximum that you have a  
14 distribution of.

15 DR. REISS: Correct.

16 DR. PORTIER: If I overlaid all 1,825 of these  
17 contours on this same graph, they would be all over the  
18 place. Right?

19 DR. REISS: That's correct.

20 DR. PORTIER: I could figure out for every point  
21 how many of these 1,825 days is inside that contour or

1 outside that contour, figure out a probability of  
2 exceedance --

3 DR. REISS: That's exactly the way we have done  
4 it. So the 95th percentile, when you look at the whole  
5 field distribution, that's exactly what it is. 95 percent  
6 of those receptor points -- not receptor points, but  
7 buffer length, slivers, are below it and five percent are  
8 above it, if you were to define it that way.

9 DR. PORTIER: And then when you cut it back and  
10 did monthly or other areas, you did it the same way, but  
11 just for limited tenfold patterns.

12 DR. REISS: That's correct.

13 DR. PORTIER: That was just clarification, an  
14 easy question.

15 DR. REISS: I like it.

16 DR. ROBERTS: Dr. Wang, then Dr. Seiber, then  
17 Dr. Shokes.

18 DR. WANG: A main contribution or strength of  
19 this PERFUM is inclusion of the different databases in the  
20 meteorology and -- from the ISC models. There are four  
21 sources that you have tapped into, the National Weather

1 Service, then CIMIS and FAWN and ASOS.

2           Seems to be the primary goal using this is to  
3 compute the stability factor so that you can use that  
4 model to backtrack all the dispersion processes.

5           Would you think the other parameters from these  
6 sources we have been discussing like, say, possibly  
7 temperature or other things may also be used in addition  
8 to just the stability factor in order to estimate the  
9 risk?

10           DR. REISS: Actually, the model for the  
11 meteorological data for each site for each hour we get a  
12 wind speed, wind direction and a variability in that wind  
13 direction will go a long way to determining the result.

14           The temperature is included, the ambient  
15 temperature is included in the model. But since this is  
16 not a buoyant source, it actually has no impact on the  
17 dispersion calculations, at least.

18           But we do have temperature. If there is some  
19 way to eventually like model the flux rate as a function  
20 of temperature, which from our limited --

21           That's why we did that plot of the ambient

1 temperature versus the flux rate. Because we have the  
2 ambient temperature for every day of all those data sets,  
3 historical data sets. But there doesn't appear to be any  
4 obvious way to use that.

5 So in addition to stability, the wind speed and  
6 wind direction are part of that data base, and the  
7 temperature is also part of the database, although it  
8 doesn't affect the calculations.

9 But if at some later date that could be  
10 incorporated into the model, then that data are also  
11 available from all of those data sources.

12 DR. WANG: Just to follow up, one small point.  
13 The ambient temperature and the soil temperature, there is  
14 a time lag. So soil temperature tends to follow behind  
15 air temperature from the heat transfer in the porous  
16 media, the theory tells that.

17 To follow up on the sources of meteorological  
18 data, have you looked into the Mer flux (ph) data, the  
19 basis, and the variability of those information? Have you  
20 considered that might be useful?

21 DR. REISS: I'm not familiar with it. Could you

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1 explain what that database is?

2 DR. WANG: I think DOE and some other agencies  
3 -- it is not a particular agency, but it is more on the  
4 collection of different groups. Mostly for others -- gas  
5 emissions at the lan ams (ph) for interface could be  
6 greenhouse gases and other source of information.

7 So that the data usually tends to be fairly  
8 detailed in terms of radiation (ph), wind speed,  
9 temperature and even soil temperature and all those tends  
10 to be available.

11 DR. REISS: It is worth looking into. But I  
12 think we have done a pretty good job of looking for  
13 meteorological data near where these things are -- these  
14 applications occur.

15 The FAWN data, I don't know if that's ever been  
16 used for dispersion modeling. ASOS is just coming on  
17 line, really, in terms of being used in dispersion  
18 modeling. So we went far and broad, I think, to get  
19 meteorological data that characterized the growing  
20 regions.

21 DR. WANG: The other thing is assessment on the

1 buffer zones, the distribution, you did that for every day  
2 of the year. Would you think that's realistic for  
3 fumigation to occur any day of the year? Agronomic  
4 considerations is usually a main constraint.

5 DR. REISS: Absolutely. We can incorporate into  
6 the model if you knew for a particular fumigant or for a  
7 particular fumigant being applied to a particular crop,  
8 you could use either those monthly distributions or we  
9 could incorporate into the model a way to calculate the  
10 exact growing season.

11 The reason we did this case study analysis is  
12 that, with just assuming an equal probability of  
13 application on any given day of the year, is because we're  
14 looking at all different seasons in all different states  
15 and all different crops and potentially even different  
16 fumigants. It was just hard to generalize what those  
17 probabilities would be.

18 I would say that we're conservative in the sense  
19 that we're calculating buffer lengths during those  
20 January, February, November, December months, which are  
21 least likely for an application to occur.

1 DR. WANG: Because if you do this focus study in  
2 the most likely period of time that fumigation may occur,  
3 you likely could focus on some other sources of  
4 uncertainty so that it would reduce your requirements for  
5 computation of other needs. So that may be advantage.

6 My last question is I know we already pretty  
7 much beat that to death on the concentration, 24 hour  
8 concentration topic. But I'm still wondering shouldn't  
9 that be a time average, time weighted average in 24 hours?

10 Meaning that if you measure the concentration on  
11 the fixed interval every hour, now you can just take  
12 arithmetic mean. But usually, that may not be a same  
13 increment. So you may have to do a time weighted. Is  
14 that how you did it?

15 DR. REISS: In the PERFUM model, we calculate a  
16 concentration for every hour. So it is just a simple  
17 arithmetic average to get it. One of the advantages of  
18 the model is that in the ISC model you can't calculate the  
19 concentration from, say, 9 a.m. one day to 9 a.m. the  
20 following day.

21 You could output the hourly data and do it, but



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1 it is computationally difficult. By actually going and  
2 using the ISC code, we were able to get it. So we could  
3 calculate a 24 hour average from the start of the  
4 application to the following 24 hours.

5 DR. ROBERTS: Dr. Seiber.

6 DR. SEIBER: Your model has been run and  
7 calibrated and developed really with using a flat terrain  
8 without obstructions and no complexity, as I understood  
9 it.

10 So it is just a simple question. Do you have  
11 some plans to bring that in in future iterations of the  
12 model? And if so, how do you see that developing?

13 DR. REISS: I think it is going to be, really,  
14 the question of whether the terrain, downwind terrain is  
15 going to be a big factor.

16 The ultimate toxicity level that EPA chooses is  
17 going to be a major factor, because you are less likely to  
18 have influences from terrain if the buffer zones are 100  
19 feet or 300 feet than you would if they are 1,000 or 1,500  
20 feet.

21 Now, the ISC model can be run -- we're running

1 it in a rural mode. We also can run it in an urban mode  
2 which accounts for some of the -- if you had more  
3 obstructions that were causing more turbulence, you would  
4 get lower concentrations. It can also explicitly account  
5 for terrain in topography.

6 Most of these things would generally more likely  
7 than not reduce concentrations instead of elevate them,  
8 although, there could be other circumstances.

9 It is just when you are looking at trying to run  
10 the model for five years and trying to generalize it to  
11 all different circumstances. The most sensible way we  
12 thought was just to run it with a flat terrain. But we  
13 could look at that from a sensitivity standpoint.

14 DR. SEIBER: This is somewhat related to that.  
15 A number of these coastal areas which, of course, is where  
16 the fumigant will be used, not only, but in major amounts,  
17 they have fog at different times of the year. They will  
18 have either the ocean fog or maybe ground fog at some  
19 times.

20 Have you considered that at all in how to deal  
21 with the complexity of fog?

1 DR. REISS: I'm not sure what impact the fog  
2 would have beyond what is already included in the  
3 historical meteorological data which accounts for the  
4 stability.

5 I don't know whether it is less likely that an  
6 application would occur during fog. That's one element we  
7 haven't dealt with about the likelihood of an application  
8 occurring given the meteorological conclusions.

9 We have used the historical meteorological data.  
10 Our assumption is that accounts for the variability that  
11 we would observe.

12 DR. SEIBER: I guess one place where it might be  
13 relevant, going back to a very earlier question, is on  
14 particulate matter because you have actually introduced  
15 another phase into the atmosphere then.

16 That's more a question for the toxicologist,  
17 whether that would change the exposure or the effects.

18 DR. REISS: We really are dealing with a gas.  
19 There is no real particulate exposure. Because, as I  
20 said, it is injected into the ground. Just by gravity, it  
21 is not possible for a particulate to escape from the field

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1 and emit downwind.

2 DR. SEIBER: Then finally, just very quickly.

3 When you talked about having measured a value and no  
4 methyl iodide was being used in the vicinity, and you said  
5 there is probably no natural source of it, but there could  
6 be an analytical interference that is measured as though  
7 it was iodomethane, but it is not, in fact. Has that been  
8 explored?

9 DR. REISS: That has been the case, that there  
10 is some finite amount. It is not affecting our flux  
11 estimates just because the concentrations at the monitors  
12 downwind, as I said, are like 10,000, 100,000 fold higher.  
13 And that really dominates the estimation of the flux  
14 rate.

15 I can't speak expertly on the analytical  
16 methods, though.

17 DR. ROBERTS: Dr. Shokes followed by Winegar,  
18 Yates and lunch.

19 DR. SHOKES: One of the things you mentioned in  
20 your report in your uncertainties was the different  
21 factors in the soil that you really didn't have a good

1    measure of.  Yet, we see a tremendous effect, it looks to  
2    me like, of temperature when you put the fumigant in and  
3    you get a lot of emission.

4               Ideally, with a fumigant, the idea is to put it  
5    into soil and bring the lethal level up to some point so  
6    you can kill nematodes or weeds or pathogens or whatever  
7    for a given period of time.

8               In an ideal world, we would like to be able to  
9    see the efflux of that from the soil over a period of time  
10   in a little lower rate, I would imagine.

11              Has any work been done to determine the effects  
12   of things like, for example, I think soil and moisture  
13   might be a factor with a lot of fumigants. And perhaps not  
14   as much with iodomethane, but certainly with some of the  
15   others.

16              And is very much known about that, and could  
17   something like that be put in so that -- if we get to the  
18   point to where we regulate this, we may want to have a  
19   prescriptive application that determines when  could we  
20   get the conditions to be the safest possible to have as  
21   efficacious as possible.

1 DR. REISS: That would certainly be desirable,  
2 yes, like you said both from an efficacy and an exposure  
3 standpoint.

4 It is very difficult to get, as I said, one data  
5 point, when these flux studies are pretty substantial  
6 operations. We now have seven data points and we probably  
7 should start to do some more analysis on the variability  
8 among the weather conditions to see if we can explain some  
9 of the variation we see in flux rates.

10 But as yet, we haven't found any significant  
11 correlations with any of those variables. It is not to  
12 say that those variables aren't impacting temperature,  
13 soil, moisture. It is just that there is a lot of  
14 different things that are impacting. You may need a much  
15 larger set before you get a real handle on that.

16 DR. BARRY: I was going to comment on detecting  
17 the effect of different factors on the back calculated  
18 flux. We only have one direct flux for methylbromide.  
19 But you are asking about soil and moisture and someone  
20 else asked about tarping and different soil types.

21 And we actually did for our 35 or so studies, 38

1 studies calculated emission ratios. They were 24 hour  
2 averages. And the only differences we could detect were  
3 tarped versus untarped and bed versus broadcast.

4 And I think the reason is that there is so much  
5 variability in the measurements that it's -- these are  
6 finer differences we're talking about. And you might need  
7 to have more studies or -- there is variability on the  
8 measurement.

9 It is difficult to separate out those finer  
10 points, I think. And we had soil type. We didn't have  
11 soil temperature for all the studies. But we could not  
12 detect soil type differences.

13 Like I say, those are the factors, tarp versus  
14 untarp, bed versus broadcast. And that was it because of  
15 the variability.

16 DR. ROBERTS: Dr. Winegar, then Dr. Yates.

17 DR. WINEGAR: I have several questions. First,  
18 chemistry, a little bit. Iodomethane, do you know what  
19 the vapor pressure is relative to, say, methylbromide,  
20 which is what I have more experience with?

21 DR. REISS: It has a lower vapor pressure than

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1 methylbromide.

2 Do you know the exact numbers, Jim? Jim Platt  
3 is a chemist with Arvesta.

4 MR. PLATT: Jim Platt, Arvesta Corporation,  
5 consultant now. Previous project manager for this thing.

6 For methylbromide, the vapor pressure at ambient  
7 temperature 25 degrees is about 1600 millimeters, I  
8 believe, and methyl iodide is about 450 at the same  
9 temperature -- millimeters of mercury. So it is about  
10 one-fourth.

11 DR. ROBERTS: Thank you.

12 DR. WINEGAR: I don't have my calculator with  
13 me. What is 120 micrograms per cubic meter in terms of  
14 parts per billion, which is what I'm used to thinking in?

15 DR. REISS: Do you know it, Jim?

16 MR. PLATT: Jim Platt again. When the  
17 correction factors are put in there for -- I believe it is  
18 .02 parts per million, it translates to the 120 micrograms  
19 per meter cubed.

20 DR. WINEGAR: 20 PPB.

21 I imagine you have done this. But looking at



1 slide 21, your structure of PERFUM versus ISC, I presume  
2 you have run some test cases of running -- you said you  
3 modified the Fortran code that incorporated the ISC code  
4 into PERFUM.

5 DR. REISS: That's correct.

6 DR. WINEGAR: You have run test cases devoid of  
7 any of the additional stuff?

8 DR. REISS: I have. In fact -- yes, I have run  
9 the ISC model and run PERFUM and compared just the output  
10 I get from the ISC model to make sure that they are  
11 identical.

12 We also developed -- because the ISC model is so  
13 large, I couldn't run it in debug mode in my computer.  
14 So I developed basically a shell ISC so I could run the  
15 whole model in debug mode.

16 DR. WINEGAR: On slide 27 you talk about the  
17 linearity between flux and concentration in ISC. How  
18 about the linearity between application rate, the  
19 relationship between application rate and flux?

20 It has to do with absorption and volatility,  
21 those kind of things. Are you dealing with only one

1 application rate? How will that be affected with  
2 different application methods?

3 DR. REISS: We have done tests for multiple  
4 application rates. We -- I believe the product, if it  
5 gets registered, we'll register a lower application rate  
6 than some of the studies that were done.

7 With methylbromide, they found a linearity  
8 between flux and application rate. We don't have enough  
9 variability in application rate to do that test. So we're  
10 relying on that, the analogy between methylbromide and  
11 just general physicochemical considerations -- think that  
12 that's linear.

13 The range we're looking at right now is not  
14 that large. The lowest application rates may be in the  
15 125 pounds per acre range and the maximum might be about  
16 175 pounds per acre.

17 DR. WINEGAR: So you don't think the lower  
18 volatility of iodomethane is going to be an issue with  
19 these flux estimates in the calibration of the indirect  
20 flux that goes into the model?

21 DR. REISS: I can't think of a reason why they

1     should violate that linearity assumption.

2                 DR. WINEGAR:   Last, you have given several plots  
3     of the flux variation over time.   Some of those look  
4     pretty smooth.   There have been some discussion about the  
5     diurnal variation with the peaks, up and downs.   But  
6     others look pretty smooth.   Can you comment on why some  
7     are smooth and some do exhibit that diurnal?

8                 DR. REISS:   I think most of them saw some  
9     diurnal variability.   You are dealing with a lot of  
10    experimental variability, particularly, when you get out  
11    to the three, four, five seven days.

12                I think there is just some experimental  
13    variability that you would expect there.   But we have seen  
14    a relatively consistent diurnal profile.   But it is not on  
15    all days.   I assure you that's a diurnal variability or  
16    some other process that we can't explain.

17                DR. WINEGAR:   Have you tried plotting your  
18    hourly model output compared to these -- I know you are  
19    dealing mostly with the 24 hour and that's what the  
20    toxicology people like.   But just curious about how your  
21    model predicts on an hour to hour basis versus these

1 experimental curves.

2 DR. REISS: The model is essentially calibrated  
3 to those experimental curves. So it predicts it quite  
4 well.

5 DR. WINEGAR: Thank you.

6 DR. ROBERTS: Thank you. Dr. Yates.

7 DR. YATES: First, just a follow up on something  
8 that was said before. I think that there probably are  
9 some background, I mean, some natural sources of methyl  
10 iodide. For example, brassica will produce methyl iodide  
11 if there is iodide in the soil.

12 And I also think that in marine environments I  
13 think it is possible to produce, well, all methyl halides,  
14 actually. I don't know that that would really have much  
15 of an effect on flux measurements, especially if you are  
16 pretty far from marine environments.

17 I think it is a potential source in some places.

18 It might be something to kind of keep in mind.

19 The next thing has to do with your figure of 59,  
20 I guess, the ambient temperature. It seems to me that --  
21 you show a plot here for the average temperature versus

1 the flux. And I was wondering if you could comment on  
2 whether you think it is -- that this temperature dependent  
3 should be a factor when you are looking at application  
4 methods.

5 I could see where maybe temperature dependence  
6 for, say, flat fume, where you have tarp, since we know  
7 that the tarp permeability is strongly affected by  
8 temperature, I could see where you would see some kind of  
9 temperature dependence.

10 But when you are starting to compare flat fume  
11 to drip, would you expect to see the same kind of  
12 dependence?

13 DR. REISS: I think we have too little data to  
14 really know that. For example, with raised bed, I think  
15 the range was 56 to 61 percent between the three studies  
16 we did. That's just not enough range to start to look at  
17 temperature variations.

18 There is a little bit more range between the  
19 other two methods, but there is only two studies. Is that  
20 what you are asking? Is there a temperature dependence if  
21 you look at just one application method?

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1 DR. YATES: Right.

2 DR. REISS: I don't think we have enough data to  
3 really look at that.

4 DR. YATES: Also, you talked about having six or  
5 seven studies. It's my understanding that if you were to  
6 use this model to develop buffer zones, you would probably  
7 pick the two studies that go with the flat fume if you are  
8 trying to predict a buffer zone for a flat fume.

9 And you would use the three for the raised beds  
10 if you were looking at a raised bed study. Right? So  
11 really, you would have two, three and two studies  
12 available. Right?

13 DR. REISS: I think we may have three, three and  
14 three by the time it is all done.

15 DR. YATES: But you are taking the appropriate  
16 emission --

17 DR. REISS: Yes. We will. Like methylbromide,  
18 the buffer zones are a function of the application method.  
19 So it is -- we haven't worked out how that will happen,  
20 but that's quite a good possibility.

21 DR. YATES: Right. So it will become very

1     difficult in a way, then, to use this methodology for new  
2     techniques, say, to reduce emissions, because you would  
3     have to go out and run a number of studies, maybe like --

4             Given that two or three is not enough to capture  
5     maybe all the variability, you would have to run, say, 10  
6     studies, maybe, for -- say that you were going to look at  
7     a virtually permeable film or if you were going to look at  
8     some surface amendment to act as a reactive barrier or  
9     something of that form.

10            DR. REISS: If you had a new method and you  
11     wanted to use the model, you would have to have some basis  
12     to know what the flux rate was, of course.

13            I don't know that you would need 10. It would  
14     depend if you did one or two or three and then look at the  
15     variability that you have before you decide how many you  
16     are actually going to need.

17            DR. YATES: Well, the other alternative, I'm  
18     sure has its own problems, but would be to look at some  
19     kind of a soil based emission model where you are no  
20     longer trying to back calculate the flux, but you actually  
21     have a model which captures the processes that you think

1 are important.

2           And there is actually a scientist, ARS  
3 scientist, in Saint Paul, Minnesota, that has come up with  
4 a boundary condition that couples what is occurring in the  
5 soil to what is occurring in the atmosphere. He has  
6 atmospheric stability terms and resistance terms in the  
7 near surface.

8           And the data that you get from the weather  
9 stations could give you the atmospheric parameters that  
10 would go into this boundary condition. And the soil  
11 information, a lot of it could be determined empirically  
12 without having to go out and collect experimental data.

13           I mean, I don't want to make it sound like  
14 regression type empiricism. The only real difficult  
15 parameter to obtain might be soil degradation. But I  
16 think there are probably ways to correlate that with  
17 organic material or other soil processes. And then you  
18 would be able to simulate the flux into the atmosphere.

19           If you had a new application method or a new  
20 emission control method you would be able to pretty much  
21 do it without having to run a lot of experiments, which we



1 all know are expensive and time consuming.

2 DR. REISS: That would be desirable. If you had  
3 a way you could couple the meteorology and the flux rates,  
4 that would be great, because you could incorporate that  
5 into the model.

6 Obviously, from a regulatory standpoint, people  
7 are worried about uncertainty. We have spent a lot of our  
8 discussion today talking about uncertainty. And the  
9 question would be, with an analytical model like that,  
10 what would the uncertainty be and would it be greater than  
11 the measurement uncertainty that you have out here.

12 So if something comes on line like that that  
13 could be used and had a reasonable low uncertainty, then I  
14 think, yes, that would be great.

15 DR. ROBERTS: Let's take a break for lunch.

16 People probably need an hour or so. Let's try  
17 and get back together here. Let's try and start at 1:45  
18 sharp.

19 (Thereupon, a luncheon recess was taken.)

20 DR. ROBERTS: Let's begin with Dr. Majewski who  
21 had some questions.

1 DR. MAJEWSKI: I'm still trying to get a handle  
2 on how you calculate the actual fluxes from the field  
3 data.

4 In the field data or the field experiments, you  
5 have three different methods, three application depths, at  
6 least three different tarp thicknesses and types, four  
7 hour and 12 hour measurement periods, and you are taking  
8 these data and generating hourly flux values --

9 DR. REISS: Period flux values. By period. By  
10 whatever the measurement period in the study was. So if it  
11 was four hours or 12 hours, we just have a flux estimate  
12 for that period.

13 It is expressed as an hourly value, but it is  
14 the same for all the hours of that period.

15 DR. MAJEWSKI: My understanding from reading the  
16 background data is that the modeled flux values are based  
17 only on the field flux data.

18 DR. REISS: That's right.

19 DR. MAJEWSKI: And that you give a higher weight  
20 to the early morning hours when you expect the highest  
21 flux values and then it decreases throughout the day.

1           But the field flux studies were done different  
2 times during the day. And as you showed, you had maximum  
3 fluxes occurring throughout the day depending on when they  
4 started the application. And it's just unclear to me how  
5 -- I'm no modeler, bear with me, but it's just unclear to  
6 me how you can have confidence in the modeled data or  
7 modeled results from all those variables in the field  
8 experiments.

9           DR. REISS: Well, let me try to answer that.

10           Regardless of the start time of the application,  
11 we saw a relatively similar amount of material coming off  
12 the field. So the actual amount of material didn't seem  
13 to be affected significantly by the start time.

14           Now, what was affected was this so-called  
15 diurnal profile. At what time of day did those emissions  
16 come off. And what you would find is in a study where the  
17 flux occurred, more of the flux occurred during the  
18 evening or nighttime period, you would have higher  
19 concentrations than you would for a study where the  
20 application was earlier in the day.

21           And the model I think successfully accounts for

1     that phenomenon. By modeling explicitly the diurnal  
2     profile, the model can use a different flux value for  
3     every hour. So it explicitly accounts for that diurnal  
4     profile in the model to, I think, successfully simulate  
5     what actually was observed in those field studies.

6             Now, how it gets regulated in terms of choosing  
7     a buffer zone when you have these different buffer zone  
8     estimates depending on the start time of the application,  
9     that's another issue. And I don't presume to have that  
10    answered yet.

11            But that will partly be a policy and partly be a  
12    science decision that we have to choose. To the extent  
13    that you can be assured that if you set a permit condition  
14    that the application occurs by a certain hour, you would  
15    have to be assured that you could enforce that and that's  
16    a reasonable condition.

17            But the model can -- all I put forth is that the  
18    model can account for those sorts of variabilities.

19            DR. MAJEWSKI: Another thing I'm having  
20    troubles with is the -- I have done field studies. And I  
21    know that the meteorology on the field can be very

1 different from the meteorology at these CIMIS stations,  
2 because you are doing a field scale study and you have all  
3 your micrometeorology or micrometeorological parameters  
4 going into effect.

5           It just seems to me taking wind speed at 10  
6 meters and comparing it to what is happening down close to  
7 the surface is -- plus, the data you are using in the  
8 model is many kilometers away in most cases, there just  
9 seems to be a disconnect there.

10           DR. REISS: Well, it's a common challenge in  
11 dispersion modeling. You really have on site data for the  
12 particular application you are looking at. But I think we  
13 have dealt with that by -- we analyzed the total of 15  
14 different stations and looked at the variability and the  
15 results you got from those different stations and chose  
16 stations that represented that overall variability.

17           Of course, if someone was concerned about it,  
18 they could look at more data if they like. The model  
19 isn't specific to these four stations that we have chosen.

20           So I think that -- I think that we do have a  
21 good handle on what the variation of meteorology could be

1 out in the growing areas. Now, that's not to say that for  
2 a particular application and particular location that  
3 CIMIS station X or Y is going to be the best station to  
4 represent that.

5 But we're not trying to represent a particular  
6 application. We're trying to characterize the variability  
7 you would observe across all the applications that are  
8 occurring, because we're going to have to set these buffer  
9 zones from a regulatory standpoint prior to any of the  
10 applications occurring.

11 And let me just also say about the two and 10  
12 meter data. The CIMIS data is at two meters. The NWS  
13 data, six meters. The other data, ASOS and FAWN are at 10  
14 meters.

15 Typically, for dispersion modeling, people have  
16 preferred six to 10 meter data because it is more largely  
17 representative of a greater surrounding area.

18 Now, but we did model with the two meter data as  
19 well. We included two meter data in the analysis. And we  
20 also in two of the field studies we conducted we included  
21 measurements at both meter and 10 meter height.

1           I have some graphs. I don't know if it's  
2   necessary to show them, but we calculated the flux  
3   estimates with both meteorological measurements made at  
4   two and 10 meters. And they are virtually identical. They  
5   are very similar estimates. So I think we have accounted  
6   for that as well.

7           DR. MAJEWSKI: One last question.

8           Correct me if I'm wrong, but you did one  
9   comparison between the aerodynamic and the indirect  
10   method. And DPR has done how many?

11          DR. BARRY: We have done one. The Ross study  
12   that was published.

13          DR. MAJEWSKI: So that's two.

14          DR. REISS: That's right.

15          DR. MAJEWSKI: You are putting an awful lot of  
16   faith in two studies to say that this back calculation is  
17   the way to go.

18          I mean, I know these studies are expensive and  
19   time consuming. But it just seems to me that you are  
20   putting a lot of faith in these comparisons saying that  
21   the aerodynamic method is what you are comparing things to

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1 and using that as the starting point and you have got  
2 basically two data points. It seems a little sparse.

3 DR. REISS: We're not using the aerodynamic  
4 method as a gold standard. It has its uncertainties as  
5 well. It was an independent measurement, is what it is,  
6 is what I would describe it. So it was reassuring,  
7 albeit, from one data point that we got the same result  
8 using two independent measurements.

9 Another advantage of using the indirect flux  
10 method for this type of analysis is we essentially  
11 calibrate our model to the flux estimates that we derive.

12 Another point I would make is this product, if  
13 it gets registered and continues to be used, more data  
14 will be collected to further characterize these sorts of  
15 issues.

16 DR. ROBERTS: Dr. Portier, then Dr. Hanna, then  
17 Dr. Yates.

18 DR. PORTIER: Ken Portier here. On slide 22, we  
19 were talking over here trying to figure out. I think you  
20 have a mistake on the equation here. Probably, the flux,  
21 what is it, flux measurement, the second one doesn't



1 belong in there. Right?

2 DR. REISS: You are right. It should be the  
3 coefficient of variance.

4 DR. PORTIER: The question is the standard error  
5 there, is that the standard error of the flux adjusted  
6 estimate or is that the standard error of the M  
7 coefficient from your regression?

8 DR. REISS: It is the -- I calculated a  
9 coefficient of variance based on the M coefficient. So it  
10 is -- when I apply it in the model, it is a standard error  
11 of that adjusted flux estimate, yes.

12 So if the flux estimate was 100 and the standard  
13 error -- this coefficient variation was 20 percent, then  
14 that would be 80 plus or minus, or 80 to 120.

15 DR. PORTIER: I guess they are both the same  
16 because you are just using this (inaudible) relationship.

17 So we were just trying to figure out if that was the  
18 correct --

19 DR. REISS: You are right.

20 DR. PORTIER: It should be the predicted -- it  
21 should be the standard error associated with the flux

1 adjusted prediction.

2 DR. REISS: It is that. I converted everything  
3 to coefficients of variance just for that reason. So I  
4 kind of nondimensionalized everything so I could apply it  
5 in the model.

6 DR. PORTIER: So if we could go to slide 48. If  
7 you take that equation, what you are essentially saying is  
8 that, in your simulations, at each stage, when you are  
9 simulating, because you really are simulating it to stage,  
10 because that's all you have data for --

11 DR. REISS: That's correct.

12 DR. PORTIER: -- you have a uncertainty on each  
13 one of these points.

14 DR. REISS: That's correct.

15 DR. PORTIER: So on simulating a day, you are  
16 going to go through and select at random some value for  
17 each of these dates. Right? You are going to simulate a  
18 flux for each stage.

19 DR. REISS: For each period, yes.

20 DR. PORTIER: Each period. Now, it is  
21 possible, unlikely, but possible, that you can get high

1 values for each one of these for that whole day which  
2 would produce fairly high percentage flux rates for the  
3 whole day, which means you could get way over 100 percent  
4 at the end of a period of time.

5 So there is no constraint on this that if it is  
6 high flux at one point it has to be low flux at some other  
7 point so that the total mass of gassing for a day is  
8 somehow fixed. Correct?

9 DR. REISS: I don't have a mass constraint on it  
10 right at the moment. I have checked it. It is my  
11 observation that for these particular data it doesn't. It  
12 is not a problem.

13 But it probably would be worthwhile just to  
14 generalize the model to put a mass constraint so you don't  
15 emit more than 100 percent of the mass.

16 DR. PORTIER: One of the saving graces here is  
17 that you stopped the whole thing at 24 hours. So there  
18 should be left over mass.

19 DR. REISS: Right.

20 DR. PORTIER: You are just not worried about  
21 how much I left over for tomorrow because tomorrow never

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1 comes.

2 DR. REISS: It would be a lot more complicated  
3 if you had to extend it plus 24 hours to do that mass  
4 conservation. That's correct.

5 DR. PORTIER: The last question is you mentioned  
6 something about detection limits in the data, the real  
7 data, the measured data. How was that handled in the  
8 regressions?

9 DR. REISS: I used zero values. I also did a  
10 sensitivity analysis. Because the detection limit is on  
11 the order of 10,000 fold less than the highest measured  
12 values, it makes no difference in the regression.

13 If you did a log regression, you would have to  
14 choose some detection limit to do that regression. And the  
15 answer would be heavily dependent on what you chose.

16 But for the linear regression it is not a  
17 factor.

18 DR. PORTIER: A lot of the regression is going  
19 to be heavily dependent on those zeros as well. Because if  
20 I remember what you said, for every simulation, about half  
21 of the values are zero because they are upwind of the

1 plume. Right?

2 DR. REISS: That's correct.

3 DR. PORTIER: So they are really anchoring that  
4 regression for every one of your scenarios.

5 DR. REISS: And they correctly anchor it through  
6 zero zero. That's what you would expect. It may reduce  
7 the variability, which is something I'm concerned about,  
8 you know, just having four pairs of zeros, which is one  
9 possible idea is to only do the regression with the larger  
10 values. You get a little bit higher standard error that  
11 way. And it may be a standard error that is more  
12 representative of the larger concentrations.

13 DR. PORTIER: It doesn't look like it would  
14 change the regression all that much.

15 DR. REISS: No, it doesn't.

16 DR. PORTIER: But it really makes it bothersome  
17 when you do a regular regression and the intercept term  
18 is still significant.

19 The fact that you have so many zeros, the  
20 regression wants to go to zero. And then occasionally you  
21 find the intercept term nonzero, that makes me worry that

1 maybe there is nonlinearities going on in here that are  
2 not being picked up or lack of fit of the ISC model to the  
3 data that you actually measured in the field.

4 DR. REISS: It is hard to say. I have not seen  
5 nonlinearities. I have looked at a lot of these plots.  
6 They certainly look linear. You do occasionally get a  
7 positive intercept. Occasionally get a negative  
8 intercept.

9 I think part of the reason is that -- the ISC  
10 model, it is known to predict well what the maximum  
11 concentration might be, but it doesn't necessarily predict  
12 the location of that concentration as well.

13 So I think that's part of the variability you  
14 see in these results. It may be why that intercept term  
15 is sometimes statistically significant.

16 I don't think it is often really large, I mean,  
17 compared to constraining it through zero. Typically, when  
18 we constrain the regression through zero, it might  
19 increase by -- it usually goes up, but it is maybe another  
20 5 or 10 percent.

21 DR. PORTIER: We would like you to explain again

1     what you meant by ordering. You had the two regressions  
2     on --

3             DR. REISS:   Sorting.

4             DR. PORTIER:  Yes, the sorting.

5             DR. REISS:  With the DPR method, if you don't  
6     get a satisfied fit with just doing it, your first linear  
7     regression, what you do is you just take the data from  
8     highest to lowest independently on the X and the Y axis.  
9     Just order them independently from highest to lowest and  
10    then refit the ordered data.

11            And the theory behind that is, again, what I  
12    just talked about a moment ago, is that the model predicts  
13    the maximum concentration quite well, but it may not  
14    predict the location of that concentration as well.

15            So removing that spatial component is one idea  
16    that you could use to get a better fit.

17            DR. PORTIER:  That's kind of equivalent to  
18    shuffling the locations. I mean, you have the fitted  
19    model and you are shuffling the concentrations at the  
20    different locations until you get the ones that seem to  
21    fit the best.

1 DR. REISS: Yes. And it is generally how the  
2 people validate the model. I mean, they are looking to  
3 see that the model predicts those maximum concentrations  
4 well, but not necessarily the locations of those  
5 concentrations, which is a little too much to ask from the  
6 model.

7 DR. PORTIER: So you assign the maximum observed  
8 location to the maximum model fit.

9 DR. REISS: Maximum measure -- that's correct.

10 DR. PORTIER: So it is no wonder that the R  
11 square goes up.

12 DR. REISS: Yes. It is no wonder that the R  
13 square goes up. That's correct.

14 DR. ROBERTS: Dr. Hanna.

15 DR. HANNA: My question to Dr. Reiss is again  
16 related to the uncertainty. Suppose we have a perfect  
17 input, flux, emissions, everything. We still expect to  
18 find uncertainty in the result, either the concentration  
19 or the buffer zone or the exposure, everything, as the  
20 result of the different methodologies and schemes being  
21 employed in the ICSC model and the consequent models and



1 so on.

2 My question is, the methodology you are  
3 proposing, does it address this in particular? For  
4 example, the horizontal dispersion and the vertical  
5 dispersion are being calculated in the ISCST. So unless  
6 we have measure of the uncertainty or an estimate of the  
7 uncertainty in this parameter, which can be found, and  
8 maybe you can do Monte Carlo, so we will not be able to  
9 get an estimate of the resulting uncertainty into the  
10 concentration.

11 So really my question is really have you looked  
12 at uncertainty within the model parameterization or  
13 algorithms used?

14 DR. REISS: I haven't tried to modify the model  
15 in any way like that. I mean, if uncertainty bounds could  
16 be developed on the dispersion coefficients, that would be  
17 one way I could easily incorporate that into the model,  
18 since I have the code incorporated into the model.

19 And that might be an interesting idea, to put  
20 that into -- to incorporate that as a Monte Carlo element  
21 into the model. And that might get at some of the

1     uncertainty in dispersion coefficients.

2                 We are calculating for five years of data for  
3     over 1800 values. So you are getting some of the  
4     uncertainty that way. But yeah, I think that's a  
5     reasonable idea.

6                 DR. HANNA: There are a number of studies  
7     actually that can give some estimate of the uncertainty in  
8     the dispersion coefficients or parameters like the sigma Z  
9     and Y in the ISCST compared to different kind of  
10    observations or measurement or so.

11                DR. REISS: Thank you.

12                DR. ROBERTS: Dr. Yates.

13                DR. YATES: I'm referring to figure 18. If I  
14    understand this correctly, if you are on the red line,  
15    then you are -- basically, the red line is a stable  
16    result. It is after you do all your simulations, this is  
17    the line where you are outside -- if you are outside that  
18    line, you are outside the exposure area that's been set?

19                DR. REISS: That's correct.

20                DR. YATES: I guess I know one question for  
21    people who fumigate, like farmers who fumigate their

1 fields, is that the size of the buffer zone kind of  
2 impacts them in terms of what kind of profit they can make  
3 on their farms.

4 This kind of suggests that maybe there doesn't  
5 need to be a very large buffer zone on one side of the  
6 field and maybe down in that corner -- so this technology  
7 in principle would be able to identify a noncircular  
8 buffer zone?

9 DR. REISS: Yes, it could. And it does,  
10 essentially, for every day in the five year historical  
11 data set. The question would be really a risk mitigation  
12 issue. If you were confident that the predominant wind  
13 direction was in a particular direction for a particular  
14 site, then I think you could justify having buffer zones  
15 that were different in different directions.

16 The problem you are looking at -- you are  
17 running the model before the application. You don't know  
18 the meteorological conditions. I mean, you could forecast  
19 to some extent, but you don't know for sure what the  
20 meteorological conditions are going to be during the  
21 application. That's one of the challenges in doing that

1 sort of analysis.

2 DR. YATES: Clearly, you would have to have some  
3 base buffer zone like 50 feet or 100 feet or something  
4 like that. But I mean, I guess if the five years worth of  
5 data is sufficient to capture the variability in the  
6 meteorological data, then this should indicate -- if a 95  
7 percent risk threshold or whatever you said is adequate,  
8 then theoretically -- I mean, in a sense, you are overly  
9 safe over on this side.

10 DR. REISS: Absolutely. If you had a  
11 particular site that you were interested in and you were  
12 willing to set a buffer zone for that particular site,  
13 then you could have a situation where you could  
14 essentially calculate the 95th percentile in each  
15 direction around the field and have this sort of oblong  
16 buffer zone.

17 It would just be a question of whether that's a  
18 feasible risk management strategy.

19 DR. YATES: I don't have much experience with  
20 the micrometeorological or the meteorological models. Do  
21 you think that there is sufficient safety in all this that

1     that's possible, I mean, from your expertise?

2                 DR. REISS: It is hard to say universally true.

3     There could be a lot of micrometeorological factors that  
4     can affect particularly the wind direction in a particular  
5     area.

6                 You would really have to evaluate that on a  
7     case-by-case basis to know that you really have a handle  
8     on what the predominant wind direction is at a particular  
9     location.

10                DR. ROBERTS: I think Mr. Dawson, do you want to  
11     respond to that question as well?

12                MR. DAWSON: I was just going to say that one of  
13     the things that we need to be, we, as the agency, need to  
14     be cognizant of as we go through this process is to get a  
15     clear understanding of exactly what this means, what it  
16     represents, and I guess ultimately how we use that and  
17     some kind of risk mitigation action or whatever you want  
18     to call it.

19                There is going to be a lot more process around  
20     that. So I just want to make clear that we have come to  
21     no sort of decision about how we might end up ultimately

1 using these results, especially when you consider we're  
2 looking at things like national level labelling and how  
3 the implementation and all the other issues that go with  
4 that.

5 That's something we need to think about as we go  
6 through. We're still definitely grappling with that.

7 DR. YATES: I wasn't trying to put anybody on  
8 the spot. I'm just curious, because it does sort of make  
9 sense that if you feel that the meteorological data, if  
10 you feel that you know it well enough, and you see certain  
11 patterns, that a buffer zone should reflect those patterns  
12 at least in some way. But the other side of it is as  
13 someone who might be sitting near that buffer zone, I  
14 might have a whole different idea about it than if I'm  
15 quite a far distance away.

16 Anyway, on to another question. In some of the  
17 data I have seen when you look at comparing the model  
18 results, this would be in trying to fit the flux, there  
19 will be places where the model has a lot of zero values  
20 but sometimes you will get a measurement that's not zero.

21 I was wondering if you have any thoughts on what might

1     cause that.

2                 DR. REISS:   The model, the ISC model doesn't  
3     account for diffusion from like a concentration gradient.

4     It is only basically moving the gas in the direction of  
5     the wind.

6                 Perhaps it is possible that you have some  
7     diffusive transport that would result in low  
8     concentrations in an upwind direction.   That's my only  
9     possible guess as to what is going on there.

10                DR. YATES:   Do you think with the meteorological  
11     information, you know, that you get a mean wind direction  
12     and then there is also like a variance or something on it,  
13     is it possible that there might be some -- for a short  
14     period of time there might be a drastically different wind  
15     direction that somehow -- when you look at the mean  
16     direction of the variance, you don't get that component  
17     going in a different direction?

18                DR. REISS:   That's certainly possible.  
19     Particularly, when you have light winds, you have the  
20     so-called light and variable winds.   You have quite a lot  
21     of meteorological variation.

1           The dispersion coefficients are designed to  
2   account for some of that variation by just using one wind  
3   direction for the hour. It disperses it at some angle  
4   around that to account for that variation you would get on  
5   a sub hourly basis. But I think you are right, especially  
6   with calm winds.

7           When you have calm winds, it is very stable and  
8   you can have concentrations built up. You have a highly  
9   variable wind direction. In fact, there is some data to  
10   show that the direction is even more variable than you are  
11   actually measuring just because you are below the sensor  
12   threshold.

13           DR. ROBERTS: Any other questions, Dr. Ou.

14           DR. OU: One of your slides indicated the future  
15   of methyl iodide for that rate including chloropicrin, as  
16   much as 75 percent.

17           If chloropicrin is included, do you think the  
18   buffer zone needs to be adjusted to include in the  
19   chloropicrin?

20           DR. REISS: Yes. What is likely to happen, I  
21   can't again speak for how EPA will ultimately regulate it,



1 but one possibility would be that there would be sort of a  
2 rate limiting buffer zone or you would calculate the  
3 buffer zone for both materials and you would probably use  
4 the larger buffer zone of the two.

5 I don't know specifically, the toxicologist will  
6 have to answer it, but I don't know if there is any  
7 synergistic or additive effects between the two. They have  
8 relatively different toxicology.

9 DR. OU: Also, that the methyl iodide is much,  
10 much more volatile than the chloropicrin. Chloropicrin is  
11 like water. If they are mixed together and applied to  
12 soil, I assume the methylbromide (inaudible) would be  
13 (inaudible) by the chloropicrin.

14 May need to adjust the flux rate, including the  
15 chloropicrin.

16 DR. REISS: You think the mixture of  
17 chloropicrin that you add could affect the mass that's  
18 emitted?

19 DR. OU: Yes.

20 DR. REISS: It is not an issue I have looked  
21 into.

1 DR. OU: You add much as 75 percent of  
2 chloropicrin. You are (inaudible) going outside the flux  
3 rate of the methylbromide -- not methylbromide, methyl  
4 iodide -- methylbromide as well.

5 DR. REISS: We are accounting for the mass. But  
6 we calculate the flux rate based on the mass of the  
7 material that's in the actual formulation, of course. But  
8 whether there is an actual impact on the flux rate,  
9 depending on the chloropicrin you add, that's something we  
10 may need to look into.

11 DR. ROBERTS: Dr. Wang.

12 DR. WANG: In the margin of exposure assessment,  
13 you have like a sub model, a subroutine.

14 Would phytotoxicity being considered in there or  
15 just human exposure? Because in some locations there may  
16 be other plants across growing that can be a big concern.

17 DR. REISS: The model is purely to estimate the  
18 air concentrations. So yes, it is human exposure. If  
19 there are other plants in the area -- I mean, the model  
20 generates concentrations. And also, when you calculate  
21 the margin of exposure, you give it right now a value a

1 toxicity threshold that is applicable to iodomethane.

2 I haven't considered it, but I suppose if you  
3 had an equivalent value for a nearby plant, there would be  
4 no reason why you couldn't use it for that purpose as  
5 well.

6 DR. ROBERTS: If there are no other questions  
7 from the panel, I think we need to move on with the public  
8 comments.

9 Let me take this juncture to thank Dr. Reiss,  
10 not only for his presentation, but for his willingness to  
11 engage in the panel in answering our many questions. It  
12 has really been very helpful for us to understand the  
13 model and the case study. I think that will be very  
14 useful as we discuss it later on.

15 Let's now move to public comments. The SAP  
16 always welcomes comments from the public on the issues  
17 that we are addressing. We have two people that have  
18 indicated an interest in addressing the panel.

19 And the first person on the list is Ms. Shelley  
20 Davis from the Farmworker Justice Fund.

21 Welcome. For the record could you identify

1     yourself, please.

2                   MS. DAVIS:   Good afternoon.   My name is Shelley  
3     Davis, and I'm the co-executive director of the Farmworker  
4     Justice Fund.   And I submitted written comments this  
5     morning on behalf of the Farmworker Justice Fund and the  
6     California Rural Legal Assistance Foundation.

7                   I hope you all have a chance to read the  
8     complete written comments.   They are not very extensive.  
9     I'm just going to briefly highlight a few points that we  
10    made more in greater detail in the written document.

11                  From our perspective, it is important to analyze  
12    this model to determine whether it would yield adequate  
13    buffer zones to protect all bystanders in the area,  
14    especially in the area of maximum concentration.

15                  And to that end, we have a number of concerns.  
16    One of our key concerns, really, is when a model, as this  
17    one does, calculates -- uses as a key assumption the idea  
18    that concentration can be equal in all directions.

19                  We're concerned that the number of low  
20    concentrations will bring down the 95th percentile and so  
21    it would not be as protective as a model that looked at

1 the maximum concentration area.

2           What we're also concerned, that 95th percentile  
3 is just not enough with all due respect to the idea that  
4 the curve is not that steep. There is a significant  
5 difference between 95th percentile and 99.9.

6           And in other areas, for example in the dietary  
7 exposure area, EPA does look at 99.9. And that is -- we  
8 would want you to evaluate the adequacy of this model at  
9 that level of exposure.

10           A couple other points that I want to highlight.

11       In evaluating this model, the sort of key variables  
12 include weather, the flux rate and the consideration of  
13 how conservative it is. So we have a few comments in each  
14 of these areas.

15           Starting with weather. I think in general the  
16 idea of using actual weather data sounds like a good one.

17       But we're concerned that the way that the weather data  
18 has been used is not sufficiently protective.

19           First of all, on the quality of the data. The  
20 National Weather Service data is acknowledged to be of far  
21 greater quality than the other data sets that are used.

1 And so one of our concerns is the use of data other than  
2 the National Weather Service.

3 Another key concern, really, is related to the  
4 necessity of focusing on worst case weather conditions,  
5 which we have a good handle on from California experience,  
6 because in the last five years or so, there have been a  
7 number of mass poisonings in California due to fumigants.

8 And one key component appears to have been worse  
9 case weather, which is generally high stability and low  
10 wind speed. And so one of the concerns we have with the  
11 weather data that was used is how they treat calm hours  
12 since calm weather appears to be a key component in actual  
13 poisoning incidents.

14 And with the ISC model that's incorporated, calm  
15 hours are not included. And that too has been cited in  
16 the past by California agencies as underestimating the  
17 concentrations.

18 So the whole effect of calm hours is something  
19 we also would like you to look at and have a great concern  
20 about.

21 Just in sort of the practical of it, there was a

1 comparison provided of Bakersfield data, which was not  
2 National Weather Service data, and buffer zones produced  
3 from Fresno data, which was the National Weather Service  
4 data. And the Fresno data produced a far larger buffer  
5 zones.

6 A comment on the issue of flux rate. We commend  
7 the registrant for doing a number of studies. But we're  
8 concerned that again they didn't maybe focus on the worst  
9 case scenario.

10 And most of the flux studies that were done were  
11 done in coastal areas where the temperatures are more  
12 moderate. Of the two that were done in the Central Valley  
13 where it's hotter, they weren't done in the hottest time  
14 of the year. For example, the Manteca study was done in  
15 September.

16 So we're concerned that they don't reflect the  
17 flux when it is hottest. And this has been mentioned also  
18 -- is a time when frequently many of these fumigations  
19 actually occur, during the warm months. So that appears  
20 to be another limitation.

21 Now I guess I want to make a few comments about

1     how conservative or not the model is and what it produces.

2       There was the idea, the concept that this model produces  
3     conservative results because it is not all that likely  
4     that someone would be at the perimeter of the buffer zone  
5     for 24 hours.

6               And I guess this is another area where I feel  
7     like the real world provides a different view. The fact  
8     is that people who live at the edge of the buffer zones  
9     could easily be 24 hours in that area. Especially, the  
10    very young, the very old, and the disabled who would tend  
11    to be at home 24 hours a day. And in the summer months,  
12    school children also may well be home 24 hours a day.

13              There was another suggestion that it is  
14    protective because the indoor air is different from the  
15    outdoor air and less concentration inside.

16              We cite a study which shows actually that the  
17    indoor and outdoor air levels are quite comparable.

18              But again, focusing on the people who are likely  
19    to be hurt in such an event, low income, rural residents  
20    frequently people leave their windows open as a form of  
21    insulation, especially in the warm months.



1           So the likelihood is that they would have the  
2   same exposure indoor or outdoor over a 24 hour period. So  
3   we don't view that notion as at all a conservative  
4   assumption.

5           Another point that actually got raised in the  
6   end of the discussion, we also would like to raise in our  
7   discussion, is the idea that the model needs to take  
8   better account of multiple applications and multiple  
9   chemicals.

10           In rural areas, it is very common for fields to  
11   be quite close together, even if owned by different  
12   parties. And so a distance apart of 50 feet or 100 feet  
13   is not at all uncommon. So the possibility of multiple  
14   applications is not at all a rare event.

15           The other thing is in this model where you are  
16   looking at a 24 hour time period for concentration, it is  
17   quite possible that over the course of that time period  
18   the wind will come from one direction for some of the  
19   hours and another for another of the hours to get you a  
20   concentration from fields even if they are not contiguous.

21           So we think that that has to be taken a whole

1 lot more seriously than it appears to have been. And  
2 similarly, multiple chemicals, and as one of the panelists  
3 mentioned, that iodomethane itself is being combined with  
4 chloropicrin, so the combined effects of these different  
5 chemicals would need to be taken into account.

6 And just another point on the time of the day,  
7 because that appeared to have been a variable that had a  
8 big effect on the buffer zone, in the practical, in  
9 California, frequently these applications are made in the  
10 evening when the temperatures are cooler. And so you  
11 could much more easily be in this worst case weather  
12 scenario of high stability and low wind.

13 So these are a number of our concerns. And we  
14 hope that you will carefully consider our comments. Thank  
15 you very much.

16 DR. ROBERTS: Thank you, Ms. Davis. I would  
17 like to give the panel the opportunity to ask any  
18 questions about your comments if they them. Let me see if  
19 anyone has any questions. I don't see any. Thank you  
20 very much for your comments. I appreciate it.

21 The second individual that has requested the

1 opportunity to address the panel is Dr. James Platt from  
2 Arvesta. Welcome, Dr. Platt.

3 DR. PLATT: I'm James Platt, former project  
4 manager for iodomethane with Arvesta, now a consultant.  
5 And I'm here with some other Arvesta people today.

6 I just wanted to comment on a few things that  
7 seemed to be perhaps unanswered during the earlier  
8 discussion.

9 There seemed to be an interest in some of the  
10 chemistry of iodomethane that was not part of the  
11 presentation. So I thought I would just mention briefly  
12 that iodomethane is actually a liquid, boils at 42 degrees  
13 C, 108 (ph) degrees fahrenheit. As we mentioned, the  
14 vapor pressure is about 400 millimeters at ambient  
15 temperature. The specific gravity is 2.3. It is a very  
16 dense material. The water solubility is about one and a  
17 half percent.

18 One of the issues that we addressed early in the  
19 development of this chemical was its sensitivity to  
20 ultraviolet radiation. It breaks down very rapidly.

21 And part of the process here was to get a ruling

1 from EPA on whether it was a threat to ozone, what its  
2 ozone depletion potential was. And that was stated it was  
3 developed at 0.0015 where 1.0 is the worst case that  
4 things are compared to.

5 Soil half lives in our soil dissipation studies  
6 were in the range of three to four days. So that's an  
7 important thing.

8 Just to be sure that -- perhaps I didn't  
9 understand the questions, I wanted to make a comment. In  
10 terms of our sampling for these studies, there were no  
11 data gaps. And I got the impression perhaps someone  
12 thought there was. The sampling started when the  
13 application was being done, and then when we got to  
14 midnight we switched to a 12 hour cycle.

15 So it is possible in the first day when we were  
16 doing three hour intervals, that that would take us to,  
17 say, 19 hours. And then the remaining piece had to be  
18 taken from the next segment.

19 So there were no data gaps. And we always got  
20 the front end where the highest potential flux rates were.

21 And the 100 percent evolution that was material

1 from -- that was mentioned at Manteca, that was entirely  
2 consistent with the soil dissipation studies and with the  
3 laboratory studies on aerobic soil metabolism. That there  
4 was very little retention in soil and it rapidly evolves.

5 Another point was the fact that most of the  
6 sampling data, except for the direct flux, samples are  
7 taken at 1.5 meters. That's a regulatory requirement,  
8 that these sampling points represent the approximate  
9 breathing zone of typical workers.

10 So it wasn't part of a research program to  
11 evaluate vertical heights beyond what we did with modeling  
12 in the direct flux.

13 And then also from a regulatory standpoint, to  
14 be conservative, all of our studies were done at the  
15 maximum application rate. So studies were not done as a  
16 research program to look at flux versus application rate.

17 But again, that was a regulatory requirement and that's  
18 where we focused our efforts.

19 Again, not only do they represent -- in terms of  
20 the areas that we picked from a regulatory standpoint, we  
21 picked the highest potential agricultural uses. We picked

1 the typical growing seasons and the conditions where they  
2 are going to be grown.

3 So just to be clear on what we're doing and how  
4 we got to where we are today. If there are any questions,  
5 I would be happy to answer those.

6 DR. ROBERTS: Dr. Yates.

7 DR. YATES: Two questions, I guess. The first  
8 is I'm a little curious about -- you said that the soil  
9 half life was one and a half days.

10 MR. PLATT: No, I said the water solubility was  
11 1.5 percent. Soil half life is in the three to four day  
12 range.

13 DR. YATES: Was that soil dissipation or is that  
14 the transformation of methyl iodide in soil?

15 MR. PLATT: Those were actual field data, so  
16 they were soil dissipation studies. It is a combination  
17 of degradation by multiple paths plus volatilization.

18 DR. YATES: Because some of the work that we  
19 have done shows that the soil half life as far as just  
20 transformation would be about double that of  
21 methylbromide.

1                   Figure, for at least the soils that we work  
2   with, the soil degradation half life would be about 20 to  
3   30 days where for methylbromide it is 10 to 15 to 20 days.

4    So I just want to be clear. Because that's the lowest --

5                   At first I thought you were saying that was the  
6   transformation half life and it seemed low. But for  
7   dissipation that's fine, yes.

8                   The second question, do you have any kind of  
9   like, say, a transformation half life for photodegradation  
10  in the troposphere?

11                  MR. PLATT: Actually, we do. It is 1.5 to four  
12  days. And that's the atmospheric lifetime. Not a half  
13  life. But that's the one that was -- that's not our data,  
14  but that's reported by the Montreal protocol related  
15  working people.

16                  DR. YATES: Thank you.

17                  DR. ROBERTS: Dr. Winegar, then Dr. Maxwell and  
18  then Dr. Ou.

19                  DR. WINEGAR: My question is actually similar to  
20  the last one in regards to the chemistry of iodomethane in  
21  the soil. It is not clear to me. You say that the half

1 life is two to three days for dissipation.

2 So that's basically the flux, is what you are  
3 talking about. Is that right or --

4 MR. PLATT: No, I'm talking about a soil  
5 dissipation study where we're measuring soil samples and  
6 we're measuring different depths, we're calculating  
7 disappearance over time in a field study.

8 In California and in Florida, we see a half life  
9 of three to four days.

10 DR. WINEGAR: I'm trying to understand what the  
11 mechanism of dissipation is that you are referring to.

12 MR. PLATT: In that case, it is a combination  
13 of everything that's going on, because we know that from  
14 the flux studies there is volatilization going on. We  
15 also know there is hydrolysis. We know there is microbial  
16 breakdown. So that's the net effect we got for those two  
17 sites.

18 DR. WINEGAR: What do you know in terms of the  
19 relative fraction of volatilization versus loss through  
20 biological transformation or hydrolysis or that kind of  
21 thing, non volatilization mechanisms of dissipation?



1           MR. PLATT: I don't have a lot of information on  
2   that with me right now. We have done the -- we have done  
3   laboratory studies on hydrolysis at various pHs and  
4   temperatures. But in terms of the numbers I can give you  
5   today are in the soil half life. Others I would have to  
6   respond in some other way.

7           DR. ROBERTS: Dr. Seiber, then Dr. Maxwell, then  
8   Dr. Ou.

9           DR. SEIBER: You said the chemical breaks down  
10   rapidly. You didn't say initially how fast that was. But  
11   I gather you mean it was 1.5, 4 days is the reported  
12   atmospheric half life.

13          MR. PLATT: Unless I said something other than  
14   what I had written down, I said the water solubility --  
15   oh, was 1.5 percent. But I do have data here from the  
16   Montreal protocol studies that put the atmospheric  
17   lifetime, yes, at 1.5 to four days.

18          DR. SEIBER: One of the things ISCST fails to do  
19   is take into account either deposition or chemical  
20   breakdown in the atmosphere. I'm not sure that it is  
21   impossible to include it, but normally it is not included.

1 Have you looked at that?

2 DR. REISS: The model can take into account both  
3 of those factors. When you have, say, a life time of a  
4 day and a half, what you are actually looking at when you  
5 have these field studies or in a field situation, you are  
6 interested in the concentrations that are 500 to 1,000  
7 feet from the field, it takes seconds to get there.

8 It is not that we're not accounting for it. It  
9 is just that it is a negligible factor with the small  
10 distance between the field and the receptor site we're  
11 interested in.

12 DR. SEIBER: Right. But again, that number we  
13 probably ought to find out more what condition it was run  
14 under. Because if it was high so ozone conditions or  
15 something different about that particular atmosphere  
16 around the field, those numbers tend to slide quite a bit,  
17 those half lives in the atmosphere depending on --

18 DR. REISS: Like I said, the half life would  
19 have to be on the order of minutes to matter. It would  
20 have to be as quick as a few minutes to make a difference  
21 in the model.

1 DR. SEIBER: I hear what you are saying. I kind  
2 of agree with you. But still I don't think we should  
3 just, you know, neglect it completely. Particularly, if  
4 your model is to be used with many fumigants.

5 DR. REISS: The model itself can accommodate  
6 that. It builds in the ISC, which has both the deposition  
7 rate and a first order of decay rate. So a user of  
8 PERFUM could add both of those variables into the model.

9 DR. ROBERTS: Dr. Maxwell.

10 DR. MAXWELL: Dave Maxwell, National Park  
11 Service.

12 Are there any criteria for applying the  
13 chemicals? For example, are there any circumstances when  
14 the iodomethane should not be applied? Any atmospheric  
15 conditions where you say we shouldn't apply this right  
16 now?

17 MR. PLATT: We're now putting together our  
18 labels. And in our draft labels, we have talked about  
19 conditions to avoid would be atmospheric conversion. And  
20 that's kind of standard language for fumigants. But I'm  
21 not aware that we have developed any specific conditions

1 other than that.

2 We have talked about possible areas about  
3 minimum wind speeds or maximum, but we haven't developed  
4 them yet.

5 DR. ROBERTS: Dr. Ou.

6 DR. OU: I have one question. Does intense  
7 degradation occur for methyl iodide when you repeat apply  
8 the methyl iodide to soil?

9 MR. PLATT: I'm sorry. I don't think I quite  
10 understood your question.

11 DR. OU: For some fumigant, not --  
12 some pesticide, including fumigant like one sodium, MITC,  
13 you apply soil one time or more than one time. Then  
14 degradation rate will increase. The degradation just  
15 amend (ph) the biological. My question is does methyl  
16 iodide can cause, enhance degradation when repeat applied  
17 to soil.

18 MR. PLATT: I'm sorry. That's outside my area  
19 of expertise.

20 DR. OU: Because if (ph) enhanced, can be very  
21 severe (ph). How that can be done to only few hours?

1 MR. PLATT: I'm not sure I have gotten the  
2 essence of the question. Maybe someone else could --

3 DR. ROBERTS: Dr. Wang.

4 DR. WANG: I think we completed a study recently  
5 on that topic. It is looking at the accelerated  
6 degradation of fumigants when a field has a history of  
7 fumigation with that particular chemical. Since you  
8 probably preferentially are selecting some of the  
9 microbes, will become more efficient, degrade that  
10 particular compound.

11 I think the question I believe he was trying to  
12 ask if for methyl iodide will there be microbes will also  
13 enhance the degradation if you repeatedly use that  
14 compound at the same location.

15 MR. PLATT: I'm certainly not aware of that.  
16 That seems to be outside the scope of the meeting today.  
17 But we could respond to that if there is someone who would  
18 like to follow up.

19 DR. ROBERTS: Dr. Wang.

20 DR. WANG: Appears the specific gravity is 2.3  
21 and is applied in the soil. That means there might be

1 gravity driven. I wonder if you monitored any soil gas  
2 concentrations during your experiments besides the error  
3 concentrations in the ambient.

4 MR. PLATT: No, we didn't. We measured what we  
5 could collect in soil samples, and then above ground  
6 monitoring. But we didn't do any internal gas monitoring.

7 DR. ROBERTS: Dr. Platt, I would like to thank  
8 you for stepping forward and adding your expertise in  
9 answering some questions that we had about the chemistry  
10 of this material and how it behaves in the environment. I  
11 appreciate that.

12 There were only two people that initially had  
13 expressed an interest in addressing the panel. But I  
14 would like to make that opportunity available to anyone  
15 else in the audience, would like to make comments to the  
16 panel. This would be the only opportunity to do so.

17 After the public comment session closes and we  
18 proceed in the deliberation of questions, there won't be  
19 another opportunity for public comment.

20 Last chance. If anyone in the audience wants to  
21 make a comment. I don't see any. Let's move on then to

1 the questions.

2 MR. DAWSON: The background information  
3 presented to the SAP panel by the PERFUM developers  
4 provides both user guidance and a technical overview of  
5 the system. Please comment on the detail and clarity of  
6 this document.

7 Are the descriptions of the specific model  
8 components scientifically sound? Do the algorithms in the  
9 annotated code perform the functions as defined in this  
10 document?

11 Please discuss any difficulties encountered with  
12 respect to loading the software and evaluating the system,  
13 including the presented case study.

14 DR. ROBERTS: Question 1, which is really two or  
15 three questions, let's ask Dr. Spicer to lead off the  
16 discussion of this one.

17 DR. SPICER: I have briefly had a chance to  
18 review the user guidance document and the technical  
19 overview.

20 Just in general -- I take it I'm to respond at  
21 this point to the questions to my best ability?

1 DR. ROBERTS: Yes, if you would. Then we'll ask  
2 other discussants if they have anything to add to your  
3 comments. I will ask you to try and speak into the  
4 microphone.

5 DR. SPICER: The document, of course, we have  
6 received rather late in the process. But I did have a  
7 chance to review it. One of the things that it did seem  
8 like it was a preliminary document in the sense that there  
9 were additional studies that were alluded to in the  
10 document that were not included.

11 And I think that it's evident from the  
12 discussion today that you are still in the process of  
13 conducting field tests and those sorts of things, which is  
14 understandable.

15 As far as the detail and clarity of the document  
16 are concerned, there were some things that were clear to  
17 me, some things that were not. The things that were not  
18 were things such as the direct method of calculation for  
19 the flux.

20 That may be something that is more familiar to  
21 someone else in the field. But that was discussed, but it



1 was not clear what specifically how that information was  
2 derived from the experimental programs.

3 To be honest, I have not had a chance to look at  
4 the algorithms in the code and to see whether they are  
5 correct. I did try and load the software, and I was able  
6 to get the software off the disk. But when I tried to run  
7 it according to the read me file on the CD, I was unable  
8 to do it with a few minutes' effort.

9 I might have been able to have done that with  
10 additional effort, but I simply was not able to do that.

11 There are specific questions that came up, but I  
12 don't know whether those are apropos to deal with at this  
13 point in time or not just in terms of -- like for example,  
14 the table on atmospheric stability does not mention  
15 difficulties that you have associated with the fact that  
16 the hour before sunrise and sunset and those sorts of  
17 things are automatically destability and detail such as  
18 that. But I don't know whether that's appropriate to be --

19 DR. ROBERTS: I think if there are some areas  
20 that were particularly unclear, if you could go ahead and  
21 sort of highlight those, that would be useful.

1 DR. SPICER: I think the other area that in  
2 general was unclear, and your presentation today did help  
3 in that regard, was this idea of estimating the flux  
4 calculations.

5 For example, in your overheads, there was one  
6 overhead that included where the sensors were located  
7 during the process of the flux calculation. But that was  
8 not included in the report, at least I don't believe it  
9 was. And so that process was clearer in the presentation  
10 today.

11 Those are the main comments that I have at this  
12 point.

13 DR. ROBERTS: Fine. That's great. Dr. Portier,  
14 let's see what you think.

15 DR. PORTIER: This is Ken Portier.

16 The first issue is descriptions of the specific  
17 model components being scientifically sound. By this, I  
18 understand that the description has to be clear enough  
19 that a user could replicate what was done.

20 If so, then I think the answer is yes, the  
21 descriptions are sound. The only problem encountered was

1 with the technical documentation for the ISCST 3 model  
2 from the EPA web site that when I downloaded that data, I  
3 couldn't get some of the fonts to get my PDF file to come  
4 up.

5 It wasn't the PERFUM problem. It was the EPA  
6 side. Sorry about that. But I'm trying to be complete on  
7 this.

8 Some of the description discussed in the  
9 presentation by Dr. Reiss should be incorporated into the  
10 documentation. Specific sections that could use  
11 strengthening include description of flux rate estimation  
12 process with a clear description of the amount of data  
13 actually used in the process.

14 I think that compliments what Dr. Spicer was  
15 saying.

16 Second, we need a clear discussion of how those  
17 exceedance probabilities for each location on the grid are  
18 computed. In my discussion, I had to ask for that  
19 clarification. I think it is because it just wasn't clear  
20 enough in the documentation how those values were  
21 computed.

1           The second question deals with whether the  
2   algorithms in the annotated code performed the functions  
3   as defined. I think the answer there is yes. In fact, it  
4   is difficult to find something to say about this.

5           There are a few lines of codes where tabs are  
6   used to align the code with column 7. I look very  
7   carefully at this stuff. And this could cause problems  
8   with attempts to compile the code in other systems other  
9   than the Lahey compiler.

10          Finally, I should mention that the code actually  
11   looks like it uses Fortran 77 formatting conventions, not  
12   Fortran 95 or anything, Fortran 90.

13          As such, it is going to be quite inefficient as  
14   compared with modern coding standards. With minor effort,  
15   such as changing how the do loops (ph) are coded, it is  
16   quite possible to greatly improve the processing speed, I  
17   think, of this application.

18          I would not be surprised if you could increase  
19   it fivefold, simply bypassing it through an optimizing  
20   compiler. I attempted to do that. Of course, if you can  
21   do this, you can run more simulations. If you can run

1 more simulations, we'll get a better understanding of what  
2 is going on.

3           Were there difficulties encountered in loading  
4 the software and evaluating the system presented. I  
5 didn't load the software and run it, but I did try to  
6 compile the fixed format fortran files found on the CD rom  
7 submitted to the committee.

8           I attempted to use the compact visual Fortran  
9 compiler, which is the one I have available to me. For  
10 the most part, the code compiled with the exception that  
11 the main program, the PERFUM.4 program, it uses a max and  
12 mod functions which are very slightly formatted different  
13 between the compact compiler and the Lahey compiler. It  
14 takes about 30 seconds to fix that.

15           Actually, this thing could be compiled in a  
16 different compiler. The nice thing about the compact  
17 compiler, it has the optimizer function. I could actually  
18 start looking at where you could optimize this code to  
19 really increase the processing speed.

20           The other thing about using a more recent  
21 compiler is that you could add a visual interface. You

1 shouldn't have to open up a DOS command window to run the  
2 application. It probably should pop up its own little  
3 application window.

4           Again, that's an hour's worth of work for a  
5 programmer to create that little interface. And then I  
6 think that would make it easier for the users to use in  
7 the future.

8           Finally, one thing I really liked was your  
9 standard of using one item per line on the input file and  
10 allowing for descriptions in that first part of each of  
11 those lines. I think that's really great. It makes it  
12 easy to document the scenarios that that file is  
13 attempting to run.

14           And for those of you who have done model and  
15 attempted a lot of scenarios, it is very easy to lose  
16 track of what you are doing unless you can document the  
17 meta data that goes with the file. And the format that  
18 you have is real simple and I really like that.

19           DR. ROBERTS: Thank you, Dr. Portier.

20           Dr. Wang.

21           DR. WANG: Yes. The first question on the

1 description of the specific model component scientifically  
2 sound, I will say yes, also. The overall approach I think  
3 (inaudible). It is a design of the framework integrating  
4 the ISC model into the performance. It's a logical  
5 approach to create a more probability assessment.

6 Algorithms in the code performance, again, that  
7 might be expedited by adding functions to that so you can  
8 select the duration that you can do the simulation rather  
9 than using all the possible days and scenarios. That way  
10 will be much more targeted and also the computation time  
11 will be much smaller. But overall, I will say it performs  
12 the function for what it's designed for.

13 As far as loading the program and evaluation, I  
14 work with many models and I will say this is acceptable.  
15 But for those who has no experience of working with  
16 computer codes and models, probably it is very difficult.

17 And it depends on your target, who is going to  
18 be using this. And it is really a question of prior  
19 experience before they can use this code. That prior  
20 experience will be needed.

21 DR. ROBERTS: Dr. Yates.

1 DR. YATES: I also tried to load the model. The  
2 MOE model for some reason I wasn't able to get running. I  
3 think -- I guess there is a control TXT file. Is that  
4 right? Yes. And it took me a while.

5 But it seemed like maybe there was a file name  
6 discrepancy and something down lower. I don't have it in  
7 front of me, so I can't tell what it is. But there was  
8 some file that it was looking for on input that had a  
9 different name than anything in the directory.

10 And then what I intended to do, although I  
11 didn't have time, was to later on go back change that and  
12 try running it. But I didn't have time doing that. So I  
13 assumed that once that was done it would work.

14 The other, the PERFUM model worked fine. I ran  
15 the simulation that was the test case. And it went  
16 through without any flaws. The input file was easy to  
17 read. I agree with Ken, what he was saying about the ease  
18 of reading and having all that information in the input  
19 file.

20 I didn't have to go to the user manual to try to  
21 understand what things were, which I thought was pretty



1 nice. It would have been nice to have a user interface, a  
2 simple one at least. Going to DOS -- I used to like DOS,  
3 but I don't anymore, but it's a small thing.

4 One thing I thought for the documentation --  
5 that you probably should provide a more complete  
6 description of the field sites. There wasn't very much  
7 information on like soil types, organic matter content,  
8 things that a soil scientist would find useful in trying  
9 to interpret the things that you observed at each of the  
10 field sites, average temperature.

11 If you knew the water content or at least, maybe  
12 not in numerical form, but some description of how the  
13 field was prepared that might give an indication of  
14 whether it was really dry or had some moisture to it.

15 I thought that if with methyl iodide photo  
16 degradation is a possibility, it might be worth running an  
17 example where you could show how you could include that.  
18 Not so much that it makes a difference for risk assessment  
19 in determining the buffer zones in this case, but just, if  
20 this model was used for something else, it might help  
21 someone to help see how to do that.

1           And then I thought that as far as a little more  
2   discussion about the causes or some of the uncertainty in  
3   all this could be included in the documentation. I  
4   thought you did a very good job this morning talking about  
5   it.

6           As a matter of fact, your presentation this  
7   morning helped quite a bit in answering some of the  
8   questions that I had from the documentation. So if you  
9   could kind of merge the two, I think you would have a  
10  really good document.

11           DR. ROBERTS: We have heard some suggestions for  
12  areas to increase clarity and perhaps increase the speed  
13  and functioning of the program.

14           Let me ask other members of the panel if they  
15  have anything to add, anything they want to weigh in on  
16  agreeing or disagreeing with? Let's start off with Dr.  
17  Seiber and then go to Dr. Baker.

18           DR. SEIBER: I thought it was a good  
19  description. I enjoyed reading it. It was brought up  
20  earlier this morning, and I think it is just more of a  
21  suggestion for the future. One of the first things you

1 look for is references in the back to see if it had, in  
2 fact, any part of it been peer reviewed.

3 I understand from the comments of Dr. Reiss this  
4 morning that that is being thought about. I strongly  
5 encourage that that be continued.

6 Parts of the model, however, have been  
7 extensively peer reviewed, like the ISC component. That  
8 maybe could be brought up a little stronger. There were  
9 references in the back to that particular model. So that  
10 certainly strengthens the confidence.

11 I would only add to what Dr. Yates said, that in  
12 the description of the fields, I think the terrain maybe  
13 could be specified, not only the terrain, but surrounding  
14 trees or buildings or whatever, structures of one type or  
15 another, that might have been nearby, particularly in  
16 relationship to where the samples were taken.

17 Finally, just a general comment. We read this  
18 document. We begin to think that all fields are square.  
19 But, in fact, they come in all different sizes and shapes.

20 So I thought it might have been good to have  
21 included a description of what might happen with a truly

1 irregular sized field of types that you might encounter in  
2 agriculture and how it might perform or be made to perform  
3 with that type of field.

4 DR. ROBERTS: Thank you. Dr. Baker then Winegar  
5 and then Dr. Hanna.

6 DR. BAKER: I wanted to follow up on comments  
7 made about the file names. I was able to load the  
8 software and run it. I forget where I encountered the  
9 name problems, but I ran PERFUM in the course mode and  
10 PERFUM MOE in the course mode.

11 I think it was PERFUM MOE that was looking for  
12 a file in Bakersfield AS, but I think it was named  
13 Bakersfield 95 or something like that.

14 DR. REISS: Yes, that was an error I made. I  
15 sent an e-mail I think on Friday. It may not have got  
16 out. I apologize for that inconsistency.

17 DR. BAKER: Following the warnings, though, I  
18 was able to find it relatively quickly.

19 Then I went and tried the find mode. I  
20 shouldn't have. For me, it took me I think roughly an  
21 hour per year. So it ran for five hours on the find mode.

1 I don't know if that's -- you were saying you got quicker  
2 times.

3 DR. REISS: Yes, it is going to depend on your  
4 system.

5 DR. BAKER: Then to follow up on comments --  
6 several of the things you mentioned today were useful and  
7 should be included.

8 In particular, the update on how you perturb the  
9 flux. You got rid of the 2000 variables. So that needs  
10 to get incorporated. I'm not sure if I caught on, but  
11 there was updates like that that you had today that would  
12 be useful.

13 DR. ROBERTS: Thank you. Dr. Winegar?

14 DR. WINEGAR: I didn't get far enough in the  
15 loading process to try and run a test case. But I'm  
16 looking through the documentation in regards to the output  
17 and any kind of graphic type of capabilities it has.

18 Unless I missed something in there, but it will  
19 be useful, I think, for users to, if not incorporate it  
20 into the program, at least directions on how to  
21 incorporate the output into a graphical program so that it

1 can be viewed visually.

2 Huge tables of numbers tend to lose their  
3 meaning. So some type of graphical output like you are  
4 suggesting would be useful.

5 DR. REISS: Can I comment on that?

6 DR. ROBERTS: Sure.

7 DR. REISS: I would love to do that. I think  
8 one of the first steps is getting clear that this is a  
9 valid scientific approach before that investment might be  
10 made to make a user friendly interface and some  
11 graphical components.

12 I would comment on the user population. It is  
13 less than a dozen, I would say. It may be less than that.

14 So we're not talking about a large user population for  
15 this model. So that does enter into our thinking in terms  
16 of how user friendly and how much we want to invest in  
17 making it that way. But it is possible that more could be  
18 done.

19 DR. WINEGAR: At least some way so people know  
20 how to get it into surfer or some kind of things so you  
21 can see contour plots, that kind of thing.

1 DR. REISS: Sure.

2 DR. ROBERTS: Dr. Hanna?

3 DR. HANNA: I also like the document. I think  
4 it is well written. It is written to inform rather than  
5 through information results.

6 I agree that maybe if you include the extent of  
7 the uncertainty analysis as you presented today, that  
8 might be helpful.

9 Also, I guess my main comment is the ISCST3  
10 model. Since this is a central component of the whole  
11 project, it might be worth even one flow diagram of the  
12 ISCST3 component, the I/Os.

13 And also mentioning how -- especially the ISCST  
14 with the calm wind condition. It puts the bound at one  
15 meter per second and uses, I guess, the wind direction  
16 from the previous hours or previous meteorological  
17 observation, I guess. Is that what's --

18 DR. REISS: That's what it does when you don't  
19 use the calms processor. If you use the calms processor,  
20 which is the regulatory default mode, it actually skips  
21 and it calculates the average for that 24 hour period. It

1 doesn't ignore the calms. DR. HANNA: Right.

2 That might be good to include because the calm is an  
3 essential part of this. Thank you.

4 DR. REISS: It is, yes. Sure.

5 DR. ROBERTS: Any other comments from panel  
6 members on this question? Before we move on to the next  
7 question, let me ask the agency folks if the panel's  
8 responses or suggestions were clear?

9 MR. DAWSON: Yes, thank you.

10 DR. ROBERTS: Great. Let's go ahead and take --  
11 since we got a little bit of a late start after lunch,  
12 let's go ahead and take question two. Then we'll go to a  
13 break.

14 MR. DAWSON: Question 2: In section 2.3:  
15 Development of the PERFUM Modeling System of the  
16 background document, a series of detailed individual  
17 processes and components included in PERFUM are presented.

18 The key processes include (1) incorporation of  
19 ISCST3 into PERFUM, (2) probabilistic treatment of flux  
20 rates; and (3) development of a receptor grid. Please  
21 comment on these proposed processes, the nature of the



1 components included in PERFUM, and the data needed to  
2 generate an analysis using PERFUM.

3 Are there any other potential critical sources  
4 of data or methodologies that should be considered?

5 DR. ROBERTS: Dr. Majewski, could you lead off  
6 discussion by giving us your thoughts on this question?

7 DR. MAJEWSKI: Sure.

8 Not coming from a modeling background, I read  
9 the section and it made sense. All your arguments about  
10 incorporating the ISCST3 model into the PERFUM seemed to  
11 be the way to go in terms of simplifying the data  
12 processing and outlook time.

13 One question I had is that -- let's move on.  
14 With the different application methods changing, actually  
15 I think we discussed this earlier, but I have it written  
16 here, so I will ask it, the different sealing methods and  
17 the application methods seem to change with time.

18 How does that affect the output, the flux source  
19 term, and how does the model deal with that? And it  
20 appeared to me that it wouldn't have that big of an  
21 effect. Is that a correct assumption?

1 DR. REISS: Yes. We got relatively comparable  
2 results among the different application methods, I think.  
3 The model, I mean as we ran it for this case study,  
4 whatever you get for the flux rate for whatever the  
5 conditions for that study are, that's what goes into the  
6 model.

7 DR. MAJEWSKI: As far as the probabilistic --

8 DR. ROBERTS: Dr. Majewski, I'm sorry, could you  
9 speak up a little bit? I think the mics are not picking  
10 up very well. DR. MAJEWSKI: As far as the  
11 probabilistic treatment of fluxes, your use of the  
12 standard error from the flux studies to estimate the  
13 uncertainty seems to be valid. Again, I'm not a  
14 statistician. So I have to defer to my colleagues to  
15 comment more in depth on that.

16 Then the receptor grid, I like the idea of being  
17 able to run the model in the course mode to get an idea of  
18 the outcome first and then run it in the 99.9 or 99  
19 percentile to fine tune it. It seems appropriate.

20 The only -- my guess is that it is not a  
21 problem, but the only thing I noticed was, what Dr. Seiber

1 mentioned, was that all the examples are for square plots.

2 That's probably good for an introductory document, but I  
3 think maybe for the final thing you might want to put an  
4 odd shaped field.

5 In the data needed to generate an analysis using  
6 the PERFUM model, there are five studies done in  
7 California and one in Florida. Is this compound going to  
8 be used primarily in California or what is the  
9 distribution, the percentage?

10 DR. REISS: I don't know if I can give a  
11 percentage. But I think the predominant use is going to  
12 be in California and Florida would be the next highest  
13 usage area.

14 It is not currently used, so giving a percentage  
15 is hard to say. But I think those are going to be the  
16 predominant use areas. As I said earlier, if this product  
17 gets registered, Arvesta is committed to continuing to try  
18 to characterize some of this variability.

19 DR. MAJEWSKI: So then additional field studies  
20 are in the works if this --

21 DR. REISS: Yes. There are additional ones

1 already being planned. And then there could be additional  
2 studies with different application methods and whatnot as,  
3 you know, if a registration is achieved.

4 DR. MAJEWSKI: Now I have a comment on the  
5 background concentrations. And you mentioned two things.

6 One, that right now there is no background  
7 concentration from the use of iodomethane, but there is  
8 possible other sources. But in either case, these  
9 concentrations would be relatively insignificant compared  
10 to what is coming off the fields.

11 In an area where iodomethane is used -- begins  
12 being used in an area, it is conceivable that this  
13 background concentration will increase with time, which  
14 also brings up the point of residues coming off previously  
15 treated fields or interfering concentrations from other  
16 fields. And I think that may need to be looked into a  
17 little more.

18 And one question I have is that this is a field  
19 based study looking at the emissions from a single field.

20 And generally, the fumigants are used in a wide area, a  
21 large area.

1                   So how are you going to distinguish the  
2       emissions from a single field and know that your boundary  
3       measurements are accurate or do they need to be adjusted  
4       due to influences of other fields and applications in the  
5       area?

6                   DR. REISS: I'll try to answer that. I think,  
7       unlike methylbromide, it has a very short atmosphere half  
8       time. I think Jim just quoted it as one and a half to  
9       four days. So that is one mitigating factor in any kind  
10      of buildup of concentrations.

11                  I suspect, and I haven't done an analysis, we  
12      can use -- not really PERFUM, but we can use the ISC  
13      model, I think, to look at this in a little more detail.

14                  My suspicion is that background concentration is  
15      going to be pretty low compared to the toxicity threshold  
16      that we're worried about.

17                  DR. MAJEWSKI: Moving on to other potential  
18      sources of data or methodologies to consider. It didn't  
19      seem that the CIMIS data locations were all that plentiful  
20      -- or the National Weather sites -- you had seven or eight  
21      of them?

1 DR. REISS: There are many CIMIS stations in  
2 California. I don't know the exact number, but there are  
3 dozens in California. But we chose --

4 DR. MAJEWSKI: In that figure you showed with  
5 the met station locations, they seemed to be almost in a  
6 semi circular form in the area.

7 DR. REISS: We chose four stations, the four  
8 stations that California DPR used for their methylbromide  
9 analysis. The data were already processed.

10 We could certainly look at other CIMIS data.  
11 There is a plethora of different stations that you could  
12 look at and also for the ASOS.

13 For National Weather Service, there is a very  
14 limited number of stations. There is maybe seven or eight  
15 in all of California and some of those aren't in the  
16 growing areas or anywhere near growing areas, like San  
17 Francisco Airport, which wouldn't make any sense to use.

18 And in Florida, the same, there is a limited  
19 number of National Weather Service stations. But there is  
20 a plethora of data out there. We have tried, you know --  
21 it took a monumental effort to get 15 of these files

1 created. It is quite an effort to process the data.

2 But we developed the software at this point to  
3 do so for all the different systems. So it is possible we  
4 could look at other stations, but I think we got a pretty  
5 good handle on the variability that is out there. But  
6 there are other data.

7 DR. MAJEWSKI: I was also thinking of -- the  
8 California Air Resources Board has monitoring sites all  
9 over the place. If you can tap into that data if you need  
10 it --

11 DR. REISS: They do. I think they tend to be  
12 mostly in urban areas. They are developed for ambient air  
13 pollution. But they may have some other rural sites.

14 DR. MAJEWSKI: Actually, that brings up another  
15 point. You ran the IST model in the rural mode where  
16 you're considering it flat terrain.

17 Yet, you showed a picture where there was a  
18 housing development right next to it. And presumably,  
19 these footprints or buffer zones are for people. So why  
20 aren't you using an urban --

21 DR. REISS: The urban -- generally -- say you

1     have -- the buffer zone is at the perimeter of a housing  
2     development. The air is going to travel across a  
3     relatively flat surface before it gets there. And then it  
4     might get to a point where the dispersion is going to  
5     increase because of all that turbulence created by the  
6     houses.

7                 So I think -- we have to choose one or the other  
8     in the model. The conservative choice is to use a rural  
9     mode. That's what we have done in this case. But it is a  
10    conservative element of the model.

11                DR. MAJEWSKI: Then as another potential source  
12    of data, this one would be for validating your model or  
13    testing it. I know the Pesticide Action Network Group has  
14    a drift catcher program where they are giving air samplers  
15    to citizens and they go out and take air samples at their  
16    homes or whatever.

17                That may be a potential source of getting  
18    downwind data or something like that.

19                DR. ROBERTS: Thank you. Dr. Baker is next.  
20    Before we go to Dr. Baker, let me just point out that  
21    we're in an acoustically challenging environment.



1           It is very difficult for them to adjust the gain  
2   on the microphone so that the people in the back can hear.

3    I realize it is getting on in the afternoon and energy  
4   levels are starting to flag, but let me just exhort  
5   everyone around the table to try and speak loudly and  
6   forcefully into the microphone so that the folks in the  
7   audience can hear.

8           Dr. Baker?

9           DR. BAKER: Having used ISC from the DOS prompt  
10   several times, I know the -- the output from ISC is  
11   inflexible and it often doesn't satisfy the needs of the  
12   exposure community. And I've tried to put a bug in the  
13   ear of people on the AERMIC to consider that in AERMOD.

14           I would applaud -- for the first question, I  
15   would applaud use of ISCST3 within PERFUM in extracting  
16   the type of data. Slicing and dicing per hours and  
17   different types of output isn't that easy to obtain. And  
18   I think you did a good job.

19           I think the exposure community could be more  
20   well served if they were included up front in the  
21   development, but that might be water under the bridge.

1           The probabilistic treatment of the flux, I think  
2   it is appropriate. One issue that came up was can  
3   somebody choose an option of not perturbing the flux if  
4   they just wanted to study, say, the meteorological  
5   variability. Is there an option that you can choose?

6           DR. REISS: Yes. You just set the coefficients  
7   of variance to zero and it will run without perturbing the  
8   flux.

9           DR. BAKER: Right. Then, again, the  
10   documentation we had was for the -- with the Excel file  
11   information, and I need to look at the update. That's a  
12   good option to have.

13           Back to number one, I missed it, the ISC within  
14   PERFUM, I read some things about AERMOD. Did I read or I  
15   just thought I read that you did an ISC3 in an AERMOD  
16   model comparison for these area sources?

17           DR. REISS: I didn't do it, but it has been done  
18   by EPA recently as part of their evaluation for AERMOD.

19           DR. BAKER: The development of the receptor  
20   grid, I thought it was well explained and scientifically  
21   sound, performed in the GIS platform and then brought in.

1           I guess that if you move away from flat terrain,  
2   that would complicate the development of the receptor grid  
3   and the flag pole receptors. So it's not easy to  
4   incorporate generically, but there is EPA guidance on how  
5   to do that as long as that is suitably referenced.

6           You just mentioned in the last discussion that  
7   you had a meteorological preprocessor that allowed you to  
8   take information from the stations.

9           Referencing that, you say the user community is  
10   small, maybe smaller than five, but still you might want  
11   to make that available to the user community if they have  
12   other stations.

13          DR. REISS: Yes, it would be a goal. It is just  
14   it would have taken a little more effort and time I didn't  
15   have. But eventually, yes, if people are interested in  
16   using the FAWN or the ASOS data, we could possibly make  
17   that available.

18          DR. BAKER: I guess the emissions preprocessor's  
19   a little bit more complex. And it requires the field data  
20   and some decisionmaking. But again, as long as that's  
21   referenced, that these preprocessors are available and

1 then the community knows about them.

2 DR. REISS: You really just need Microsoft Excel  
3 to apply that sort of method. The trick is the long field  
4 study you have to do.

5 DR. ROBERTS: Thank you. Dr. Bartlett?

6 DR. BARTLETT: I think a lot of this was taken  
7 care of earlier on this section on Number 2, probabilistic  
8 treatment of the flux rates, that the discussion we had  
9 earlier could be in this section and be much more useful.  
10 It was a little confusing to me on the purpose  
11 and why of doing that. And I think what you did explain  
12 is that, if I don't have it wrong, is that when you have  
13 tried different random number generators, it didn't really  
14 make much difference in the actual results.

15 Now, with the indication that you gave to us of  
16 some of the graphs, though, where we have information  
17 where there is noise in it, which I assume some of this  
18 noise is coming from the air term that you are putting in,  
19 which is now I guess a Z as opposed to T, and I wasn't  
20 sure if that affected spatial as well when you had the  
21 illustration, I forget what slide it was, where you had

1 the red.

2 DR. REISS: Is this the buffer lengths?

3 DR. BARTLETT: Yes. With the buffer -- is any  
4 of the spatial also affected by the air term?

5 DR. REISS: Yes, it would. Generally, the air  
6 term won't change the 50th percentile of the buffer length  
7 distribution, but it will change the upper percentiles.

8 So both in terms of just using higher flux rates  
9 and also possibly combining a higher flux rate with a more  
10 worst case meteorological situation, you just have a  
11 better chance of that when you have more higher flux  
12 rates.

13 DR. BARTLETT: And it might affect the  
14 orientation as well of the location?

15 DR. REISS: Yes, it could certainly affect the  
16 orientation as well.

17 DR. BARTLETT: What still puzzles me about it is  
18 to the -- a standard error a lot of times does some random  
19 process that we don't quite understand. But there are  
20 processes that we do understand. And that's somewhat  
21 within soil emission the different factors that we

1 discussed earlier, like soil moisture and other factors  
2 about soil, soil temperature and other factors.

3           So it is not clear to me whether if including  
4 that in the model or in the testing or the evaluation of  
5 the model, whether it makes sense to do that.

6           DR. REISS: At this point with the data set we  
7 have and the analysis we have done, I don't see a way to  
8 incorporate temperature and soil moisture content.

9           We only have seven data points. If start to try  
10 to analyze all these potential variables like organic  
11 matter content, soil temperature, ambient temperature, et  
12 cetera, you really don't have enough data at this point to  
13 develop a model of that process.

14           If that's eventually possible or there are some  
15 -- you could combine that with some more  
16 phenomenologically correct model like Scott was referring  
17 to earlier, Dr. Yates, that would be great. It would be  
18 an improvement on the predictions. But right now I don't  
19 see that that's possible.

20           DR. BARTLETT: What is useful for me was the  
21 discussion of the methylbromide studies, because I'm less

1 familiar with the volatiles, that a lot of these  
2 parameters that we're used to affecting other compounds  
3 may not be so.

4 And the fact that we had the discussion earlier  
5 on physical chemical properties from the audience, it  
6 would be very useful to have that information in the  
7 report so we can understand better in this section why you  
8 are choosing which processes you are choosing.

9 I very much like introducing probabilistic  
10 approaches, because that's closer to the real world even  
11 though it may not be at the right time and place the way  
12 we can do that.

13 One quick thing, I guess on the receptor grid, I  
14 guess it would be problematic with approach to do a  
15 traditional square grid, which people are much -- it is  
16 easy to produce a square grid.

17 But I was wondering -- then you will have  
18 distances that are varying distances in your receptor  
19 grids. But with the interpolation that you are doing,  
20 maybe that would be more traditional as far as looking at  
21 dispersion in a squared grid and easier to generate on a

1 GIS for those of us who take a long time to do simple  
2 things in GIS.

3 DR. REISS: Actually, we have already run it  
4 this past week for an irregular sized grid.

5 There is no reason why the method can't be used.  
6 We just define -- in the GIS program, we can just define  
7 all those receptor points along the rings at the right  
8 distances and the model had no -- we didn't have to modify  
9 the model in any way to run it.

10 It is something we can look into more in the  
11 future. The impact of it would be if you had the wind  
12 going across the smaller distance, you would have a less  
13 of an impact. If you had the wind, the predominant wind  
14 going across the longer distance of the field, you might  
15 get a larger concentration.

16 It is a potential mitigation option even to at  
17 least to give advice to growers as to, you know, what sort  
18 of orientation to apply. But it is something we can  
19 account for in PERFUM, and we'll develop a few more of  
20 those receptor grids as we go forward.

21 DR. BARTLETT: It could be a nested grid too.



1 And then you might get some more efficiency in running it  
2 as well as closer endpoints, closer to, and a larger grid  
3 farther out. Then you might get some computational  
4 efficiencies.

5 DR. REISS: That is how it works, actually. The  
6 grid sizes are about 10 meters apart for the first one  
7 hundred meters then they start to get progressively --

8 DR. BARTLETT: I was thinking efficiencies in  
9 the sense in the corners you have high density that's  
10 redundant in a lot of ways. But it's not a significant  
11 point.

12 As far as the weather data, I think I put that  
13 off to when we talk -- that Dr. Majewski was -- to later  
14 when we talked about the other weather data.

15 So I guess -- the other thing in this section,  
16 it is a section you talk about seasons. And there is a  
17 bias that's introduced here that's not necessarily upward  
18 or downward. It really depends on the region.

19 That our experience in -- we have done a lot of  
20 emission modeling or working with emission modeling with  
21 ORTECH, which is another -- other chemicals, other

1 pesticides have been modeled for emissions.

2 I don't know what is closest to the fumigants.

3 There has been a lot of work recently also on lindane, but  
4 none of this is as volatile as what you are working with.

5 I don't know if there are things that could be learned  
6 from that or mentioned with that.

7 But one thing we have noticed a lot is extreme  
8 variability between season, which also popped up, came out  
9 in your analysis by month. By doing yearly averages, it's  
10 good to know what kind of bias might be coming out of  
11 that.

12 The other thing, I guess, about probabilistic  
13 approaches is when we're talking about impacts and health  
14 is typical impacts on who happens to be where. And I  
15 think the comment from the audience was particularly  
16 important is, if you have worse conditions like an  
17 inversion or something like that, then the 99th percentile  
18 is important.

19 I guess this may go further ahead in a future  
20 section, but the maximum, there is a virtue of having --  
21 focusing on not diminishing the importance of the maximum

1 24 hour length.

2           Because then in those conditions, if the farmer  
3 is aware of this when they are applying it, what kind of  
4 conditions that may result in these maximums and certain  
5 distances like a school that happens to be at the corner  
6 of the farm or something like that, then their  
7 probabilities have changed dramatically.

8           So I think as far as the other types of  
9 scenarios you have --

10           DR. REISS: In terms of what probability to  
11 choose for regulation, it is an issue I don't want to  
12 really comment on.

13           I will try to just deal with the scientific  
14 computation of those probabilities at this point. And,  
15 you know, how it is actually used to calculate buffer  
16 zones is really a policy decision that goes beyond just  
17 science as to how conservative to be.

18           DR. BARTLETT: I have one technical question, I  
19 guess, on crosswind. I believe you said you eliminate  
20 that in interpolation if that situation arises.

21           But sometimes in the real world that's what

1 results in higher concentrations in certain areas. That's  
2 a real effect as well. I don't know -- if it's only  
3 within the interpolated space, it is not that meaningful.

4 DR. REISS: The problem -- a problem occurred  
5 because of the geometry of the calculation. You  
6 occasionally had -- for a given spoke, the concentration  
7 didn't decline for the first period or the first couple  
8 spokes. And it just caused -- basically, the  
9 interpolation algorithm fails if you do that.

10 So I just had to eliminate those points. I'm  
11 certainly not -- we're still going out beyond the field to  
12 correctly calculate the distance until it reaches the  
13 toxicity threshold. That was purely a matter of just  
14 making the interpolation algorithm work correctly.

15 DR. BARTLETT: In regional modeling we overlap  
16 and then combine. You get new centers of high  
17 concentration, which seems to be a real phenomena. I  
18 guess the last comment -- I guess I will stop there.

19 DR. ROBERTS: Dr. Hanna, do you have comments to  
20 add?

21 DR. HANNA: I wonder if we consider, especially

1 regarding the integrity of the meteorological data, a  
2 possibility, if you want to, for a broader use of the  
3 PERFUM is to use meteorological data from another  
4 meteorological model.

5 I examined five or so areas really for areas  
6 where the data does not exist, for example, or very low  
7 quality as we have seen in some of this.

8 DR. REISS: I certainly haven't built it at this  
9 point to incorporate a wind field like MM5. Although I  
10 don't think the spatial resolution of a wind field from  
11 like MM5 would -- you would have the same answer for --  
12 among the space you have for agricultural field, there  
13 wouldn't be any variation in those winds fields in MM5. I  
14 don't think the resolution would get down that far.

15 You can use any data that you can convert to an  
16 ISC format. If it was predictions from MM5, you would  
17 just need to convert that to ISC format.

18 DR. HANNA: Right. And MM5 can go to four  
19 kilometer in some of the nested applications. So that  
20 might be close enough to represent the meteorological  
21 conditions in certain areas.

1 DR. REISS: Sure.

2 DR. ROBERTS: Thank you. Dr. Wang?

3 DR. WANG: In the report, you said that you used  
4 the five year meteorological data. And you used the five  
5 years for a different reason than the EPA's Office of Air  
6 using five year for permitting purposes. But you mostly  
7 tried to provide a more probabilistic approach.

8 So if the reason is different, than why use a  
9 five year? If you use a 10 year or 30 year, likely the  
10 means of variation will be different. You go back to  
11 history.

12 DR. REISS: I used five years in one level for  
13 the same reason that EPA uses five years. They found in  
14 their meteorological analysis and their guidance that  
15 using five years in meteorological data characterizes the  
16 historic variability in meteorology.

17 What I meant to say when I made that statement  
18 is that when people run a permitting application for a  
19 continuously emitting industrial source, and they run it  
20 through five years of meteorological data, they are  
21 interested -- they are actually calculating real estimates

1 for each and every hour because that source is emitting  
2 for all that period.

3 We're using it in a different way just in the --  
4 we're just getting individual estimates of the 24 hour  
5 concentration. We're doing that 825 times to develop that  
6 time series. That's all I meant.

7 But in terms of characterizing variability, the  
8 same argument -- I'm making the same point that EPA is  
9 making in terms of using five years.

10 DR. WANG: Another comment on the very last  
11 question we are supposed to ask. On the  
12 methodologies that you likely may use to improve the flux  
13 estimates, I guess it was brought up earlier, is using  
14 some of the soil space, the emission model's more  
15 deterministic predictions, that will integrate all the  
16 factors we have been debating, the soil moisture, the  
17 temperature, even the bulk densities, soil type, organic  
18 matter, degradation. Those can all become an input.

19 Even though you don't have direct measurements,  
20 likely you may have a very reasonable guess, I suppose, to  
21 put that in the model and provide another prediction of

1 the source trend to fluxes or time.

2 And at least that can be provided as a  
3 comparison to your current approach, probably.

4 DR. REISS: I'm very interested in these soil  
5 models. Like I have said, we have only a limited amount  
6 of data to sort of characterize what the impact of all  
7 these factors are.

8 And in developing a model to be used for  
9 regulatory purposes, I think we're reluctant to rely on  
10 something that may be purely theoretical without field  
11 data, actual field data to back it up.

12 But that's probably the future and I can't  
13 honestly say I know what the state of those models are and  
14 how accurate they are. But if something could be done  
15 that was accurate and would meet the regulatory burden,  
16 then I would be all for incorporating something like that.

17 DR. ROBERTS: Let me ask other panel members if  
18 they have any comments to add on this question.

19 Dr. Heeringa?

20 DR. HEERINGA: I want to be sure that we're  
21 clear in our report on the interpretation of the



1 perturbation of the estimated flux rates.

2           In slide number 43, I don't know if we can bring  
3 that up, slide 22, I think in your presentation, and I  
4 haven't cross-referenced with it the report to see, but I  
5 think we recognize -- I think there is a typo there. You  
6 have standard error instead of CV in the equation. But I  
7 want to be sure.

8           If we go now to slide 43, the sigma error there,  
9 that is the, essentially, the standard error on a  
10 prediction of a future value.

11           As I look at this CV, that CV makes sense as a  
12 standard error on the expected value of the flux rate  
13 conditional, which is a different item. And I wonder if  
14 that sigma includes not only the variance associated with  
15 estimating the regression, but also the residual variance  
16 associated with predicting a future value off of that  
17 regression.

18           In other words, in linear regression where you  
19 have the prediction problem, one of developing confidence  
20 bounds for the expected values, one of them of developing  
21 a projection bounds for a future value, the projection

1 bounds are broader than the confidence bounds on the  
2 expected value.

3 I'm thinking that this is essentially a random  
4 draw from the expected normal distribution for the  
5 expected value, taking into account the variance on the  
6 regression coefficient, but it does not incorporate the  
7 residual variance for predicting a future value from that  
8 regression coefficient. Just mention that here. We'll  
9 try to provide formula, just for clarification.

10 DR. REISS: I think if you -- I may not  
11 completely understand your question. But I think at least  
12 when you constrain the intercept through zero, those would  
13 be the same. Is that correct?

14 DR. HEERINGA: No, they shouldn't be. It's just  
15 a different concept of whether you are trying to develop  
16 error bounds for your predicted regression -- your  
17 expected value regression line or projection bounds for a  
18 future value that you predict from that regression.

19 There is an additional error term, I think, that  
20 has to be added for the latter.

21 DR. SMALL: I think you are right. If you think

1 about the prediction error for a value, though, it would  
2 be for an ambient concentration value predicted as a  
3 function of an aerial emission rate of a flux rate.

4 And that's really not what they want. What they  
5 do want is the uncertainty in the emission rate or the  
6 flux rate, which is the slope.

7 So in a sense, what they have done there is  
8 correct statistically as terms of characterizing the  
9 uncertainty in the flux rate, which is the slope of the  
10 regression.

11 The bigger issue, which we'll get into when we  
12 address question 3, is whether or not that's -- the  
13 coefficient of variation that comes out of that, which as  
14 we see are typically on the order of 10 to 30 percent,  
15 really reflects the site to site, period to period  
16 uncertainties and variabilities that are really out there  
17 or if it just reflects the emission rate at that site for  
18 those set of tests.

19 So that is a much bigger issue.

20 DR. HEERINGA: Thank you for that clarification.

21 But it is the issue of whether in fact you are

1     trying to deal with projecting a value from that  
2     regression to a larger population from which this sample  
3     is drawn here.

4             DR. SMALL:   Correct.

5             DR. ROBERTS:  Dr. Portier, I believe, would like  
6     to follow up.

7             DR. PORTIER:  No.

8             DR. ROBERTS:  Cover something different.

9             DR. PORTIER:  Different question.  I wanted to  
10    talk a little bit about the projection grid.

11            When you stop and think about the computational  
12    cost that goes on, I would say probably 70 percent of your  
13    computation occurs in or immediately around the field.  
14    And yet your interpolation for your boundaries are much  
15    further out.

16            It seems like you have a sparse set of points  
17    where you are trying to interpolate and develop that pink  
18    curve and you have a dense set of points right close to  
19    the field.

20            I'm wondering if some kind of adaptive system,  
21    which would start with a course grid and then allow you to

1 fine tune that grid in a second task to actually get  
2 really nice, if that's your objective, might really speed  
3 up the computational thing.

4 If you are trying to put stuff into a GIS, maybe  
5 -- I tend to agree, if you were trying to actually develop  
6 the whole probability surface, you probably got the right  
7 grid right now.

8 But if all you are trying to do is estimate that  
9 threshold boundary, you are wasting a lot of your  
10 computational area simulating what is happening close in  
11 when what you want to do is what is happening out at that  
12 --

13 DR. REISS: You still get -- there are still  
14 distances or slices of the field where it is close in.

15 Actually, the first run I did of this we didn't  
16 have as dense a grid close into the field for that reason.

17 And I just found that there were too many errors  
18 in the interpolation algorithm. I have a fairly good hold  
19 on these kinds of mathematics, but I'm sure there are  
20 other more computationally efficient things. Maybe a two  
21 dimensional interpolation might even improve the

1 computation.

2 I'm all willing to hear ideas on how you can  
3 improve the computational efficiency.

4 Can I make one other comment about the point you  
5 made in the -- as I said, I'm more than willing to hear  
6 ideas on computational efficiency. But I would mention  
7 that at least 75 percent of the computation time is the  
8 ISC model.

9 If you really want to reduce the computation  
10 efficiency, you need to reduce the number of receptor  
11 grids as opposed to the latter issue about the loops.

12 I think the biggest bang for the buck is getting  
13 a sparser receptor grid to simulate the calculations we  
14 have here.

15 DR. PORTIER: I want to see you tackle both of  
16 them.

17 DR. ROBERTS: Other comments or suggestions from  
18 the panel members? Yes, Dr. Seiber?

19 DR. SEIBER: The documentation provided I think  
20 gives a very good description of how ISC is coupled with  
21 PERFUM. I thought that was very nice. And of course they

1     have -- the back calculation of flux and ISC have been  
2     used separately and now you put them together in very  
3     logical combination. I thought that was very well  
4     described.

5             On the other hand, even though ISC has been used  
6     for many years and successfully in these kinds of, not  
7     only fumigants, but in some cases for other pesticides and  
8     I think for fumigation chambers as well as fields, there  
9     is a longstanding use.

10            But we probably shouldn't forget some of the  
11     differences in an agricultural field where you typically  
12     start spraying over on this side of the field and in a  
13     ribbon manner you may go through and fumigate, and, of  
14     course, you are emitting over here while the tractor is  
15     still moving over there.

16            That only happens during the application. But I  
17     have always wondered myself, and maybe you have an answer  
18     for this, maybe it should be brought up, is that a  
19     complicating factor? It would be nice if the field was  
20     all treated at once and then you go out and put your  
21     samplers up and it all emits at once, but it doesn't

1     happen that way.

2                   DR. REISS: For a couple of the field studies, I  
3     didn't describe it in the report, but for a couple of the  
4     field studies where it took a long time to do the  
5     application, we essentially broke them into slivers in  
6     accordance with the time it took to do the application.

7                   It is a simplification, obviously, when we run  
8     PERFUM to put it altogether or just assume it is all  
9     applied at once. You are certainly capable of doing that  
10    in the model. Whether or not it would be a worst case or  
11    best case situation would just depend on the wind  
12    direction relative to where you were applying.

13                  It is hard to make choices about how to exactly  
14    apply that. You could look at it in terms of sensitivity  
15    analysis, but for calculating the fluxes we did take that  
16    into account.

17                  DR. SEIBER: So a simulation could be run taking  
18    that into account and show that really it doesn't matter  
19    that much in the overall result if it doesn't.

20                  The second comment on are there any other  
21    potential critical sources of data. Again, come back to



1 the fact that the wind and -- particularly, the wind can  
2 vary so dramatically from one field to another when you  
3 get into more of the complex terrain situations, and some  
4 of these fields are in fairly narrow valleys, for example,  
5 where you could have a Ventura effect that's probably not  
6 taken into account by the nearest meteorology.

7 I don't know how to get around that except --  
8 I'll just go ahead and say it. There may be cases where  
9 you need to buy or rent a met station and take it out to  
10 the field. It is just not -- that weather pattern just  
11 isn't simulated by your nearest recording station.

12 DR. REISS: That's possible. I think the key  
13 question is, since we're not predicting the concentrations  
14 for a particular field, is whether the weather stations  
15 that are out there are accurate enough to capture the  
16 overall variability that's out there.

17 You know, the idea of looking at a specific  
18 field if somebody was very interested in doing that, then  
19 that's an interesting idea. And you could certainly  
20 refine the estimates.

21 As Dr. Yates pointed out, maybe you could

1     justify not having a buffer zone in a direction that you  
2     know is not going to have a lot of impact.

3             DR. ROBERTS:   Anything else on question 2?

4             Let me ask the agency then if the responses from  
5     the panel to question two were clear?  Do you need any  
6     clarifications on some comments or suggestions?

7             MR. DAWSON:   No, they are clear.  Thank you.

8             DR. ROBERTS:   Before we move on to question  
9     three, which will be the last one we'll tackle today, I  
10    think the panel could probably benefit from a 10 minute  
11    break.  And I think the audience could benefit from a 10  
12    minute break too.

13            Let's take a short break.  Try and reconvene  
14    about 10 minutes after 4.  We will tackle number three,  
15    which will be the last one today.

16            (Thereupon, a brief break was taken.)

17            DR. ROBERTS:   Pose question three to the panel.

18            MR. DAWSON:   Question 3, The determination of  
19    appropriate flux and emission rates is critical to the  
20    proper use of the PERFUM model as these values define the  
21    source of fumigants in the air that can lead to exposures.

1           Upon its review of how flux rates can be  
2     calculated, the agency has identified a number of  
3     questions it would like the panel to consider.

4           In PERFUM, flux rates were treated as a  
5     probabilistic variable with an uncertainty developed from  
6     the statistical bounds of the flux calculation. For each  
7     measurement period, a standard error is generated that  
8     reflects the measurement uncertainty of the flux rate.

9           PERFUM then perturbs the concentration estimates  
10    within each period by the standard error using Monte Carlo  
11    methods to simulate the uncertainty in the flux estimates.

12    What, if any, refinements are needed for this process  
13    including the manner in which the flux values were  
14    calculated for each monitoring period to generate the  
15    standard error estimates?

16           How appropriate is it to use a flux or emission  
17    factor from a single monitoring study (or small number of  
18    studies) and apply it to different situations such as for  
19    the same crop in a different region of the country?

20           Please comment on PERFUM's capability to  
21    adequately consider multiple, linked application events as

1 well as single source scenarios.

2 Does PERFUM appropriately address situations  
3 where data are missing?

4 In the back calculation approach used for  
5 estimating emissions rates, the regression of measured  
6 versus modeled values can be forced through the origin or  
7 not. Which approach does the panel prefer and what are  
8 the implications of each approach?

9 DR. ROBERTS: Okay. There are lots of questions  
10 in this question.

11 Before we get started, let me remind panel  
12 members, please, bring the microphone in close and speak  
13 directly into it as you make your comments. Let's start  
14 with Dr. Yates.

15 DR. YATES: This one has a lot of -- there is a  
16 lot of substance to it. In some ways it's -- I think  
17 maybe the -- well, I guess where I would start is that I  
18 think that the idea that the PERFUM uses a probabilistic  
19 treatment overall is better than if it would have all been  
20 deterministic, just as kind of a background statement.

21 Because capturing variability, even as you will

1 see in a minute in some of my comments, is probably not  
2 all the variability that's present. It is still better  
3 than just assuming some single average value.

4           There are a number of assumptions that go into  
5 how they determine the flux. And if you can get to the  
6 point where you can accept the assumptions, you know  
7 everything is really fine.

8           However, it seems like there are possibilities  
9 to improve the risk assessment by adding additional  
10 sources of variability, which I have got some slides I'll  
11 be showing in a couple minutes that I hope will bring the  
12 ideas across clearly.

13           One thing I like is that the -- just the  
14 approach as a whole, is the fact that the buffer zones are  
15 affected by meteorology. And that the -- as I think we  
16 talked about earlier, that the shape wouldn't necessarily  
17 have to be circular.

18           So anyway, now to get to the questions that are  
19 at hand, I guess the first one was what refinements --  
20 where is it, flux values. Okay.

21           I can't seem to see the question in here. There

1 are so many of them. What refinements are needed for this  
2 process including the manner in which flux values are  
3 calculated for each monitoring period to generate the  
4 standard error of estimates?

5 It seems to me first off that the standard error  
6 the way it is being currently done captures some of the  
7 variability. In essence, it is the variability that is  
8 expressed as what I would call discrepancies between the  
9 data and the model. And that's good.

10 And I think that in terms of if you are just  
11 looking to having a flux value that would allow you to use  
12 the model and the risk assessment for that particular  
13 field, it seems like it is okay. Because as Dr. Reiss  
14 said, it is sort of calibrated for that particular field  
15 study.

16 Now, the problem I see is when you start trying  
17 to apply that elsewhere. And let me have the first slide.

18 We have looked at a number of studies of  
19 methylbromide. This slide here shows a cumulative  
20 probability versus total emissions for -- I forget how  
21 many methylbromide studies are shown here, but, basically,

1 these are all direct measurements of methylbromide total  
2 emissions.

3 The studies that are included here would be  
4 Yagi's studies, Williams', who also worked with -- Yagi  
5 also worked with Ralph Cicerone. Williams worked with  
6 Ralph Cicerone. Mike Majewski and Jim Seiber's study and  
7 some studies of ours, which are the ones that have the  
8 white bars.

9 Now, if you look at how much variation there is  
10 across studies, you can see that the low end total  
11 emissions are somewhere around, I think that's about 30  
12 percent. And they run all the way up to the high that we  
13 have seen, which is over 80 percent.

14 When I was looking at this, this actually was  
15 part of a study that we were trying to look at what the  
16 effect on global emissions would be if VIFs were used  
17 instead of high density polyethylene. Because these field  
18 studies were all shallow injection, tarped, flat fume  
19 studies.

20 So in the process, my objective for the paper  
21 was I looked at fitting a log normal distribution to this

1 data where -- I didn't want to fit through the data, so I  
2 put kind of bounding distributions on there.

3 So basically, it looks like the mean value would  
4 be 50 percent and there is a lot of spread between these  
5 studies.

6 So if one of these studies were used to obtain  
7 the emissions and then used for risk assessment, I think  
8 you would get quite a different outcome than if you  
9 sampled from one of the let's say a 50 -- a mean  
10 distribution of 50 percent with this kind of spread.

11 If you look at the bars up here, this is the  
12 variability we observed in the experiment. We had a  
13 number of methods for estimating the emissions and so we  
14 were seeing like -- I don't know, that's about 10 or 15  
15 percent variation in one study.

16 But you start looking at over the study, and the  
17 variation is quite a bit larger. And I think this is  
18 typical. The data I see, and it is shown in some of these  
19 reports, it seems to be all over the map.  
20 And as a matter of fact, there are times when you can see  
21 things -- for example, if you look at in the report on



1 page 92, table 6.1, there is emission ratios for  
2 methylbromide from the CDPR analysis that have three  
3 broadcast columns.

4 The broadcast high barrier, they have a very  
5 high barrier and a VIF. If you look at the mean emissions  
6 for these things, you will see that applications with a  
7 permeable film have a mean of .253. If you have a less  
8 permeable film, the mean goes up to .5, which right away  
9 you start saying physically something seems a little wrong  
10 here. VIFs, the mean drops to .3. But it is still higher  
11 than when you have a permeable film.

12 So there is -- the variation that you are seeing  
13 in experiments is tremendous. Even to the point where  
14 when you use a VIF, which is very good barrier, I will  
15 actually show a slide which shows the permeability of some  
16 films, you can have higher emissions from an experiment  
17 than you would when you basically have a barrier that lets  
18 a lot of gas go through.

19 So this kind of variation seems like it might be  
20 something that should be somehow included into the risk  
21 assessment model.

1           Can you get the next slide? Now, all that was  
2   on cumulative emissions. What the model uses are period  
3   emissions. This slide shows -- I guess you can't see the  
4   scale. There is days. That's one day, two day, three day.  
5   It goes out to eight days.

6           Here we have the flux density. These are period  
7   values of two to four hours. Three methods. The same  
8   basic data set was used. This was all done on the field.  
9   We had a sampling mass in the field. We had anemometers.  
10   We got temperature gradients so we could do the  
11   aerodynamic flux method.

12           But that same profile of concentrations on the  
13   field and the anemometer wind speeds that we collected on  
14   the field, we used a method called theoretical profile  
15   shape. And it integrated horizontal flux method.

16           Three separate ways to analyze the data. If you  
17   look at this, there is no correspondence between the flux  
18   at a particular time for each of the three different  
19   methods. And yet the cumulative emissions are almost the  
20   same for each three.           So there is a real problem  
21   when you are trying to get period -- what I like to call

1 instantaneous flux. It is very difficult to get numbers  
2 that have meaning when you start spanning across different  
3 locations, different methods for estimating them,  
4 different times.

5 Can I have the next slide? We had one other --  
6 unfortunately, this has some other information on it as  
7 well, but we also used flux chambers to measure the flux.

8 The points here are the flux chamber  
9 value. And if you compare it to the ones before, the  
10 scale on the previous slides were 300. This is 120. You  
11 will see that there is, again, a drastic difference when  
12 you use different flux methods.

13 So it would be nice if the method that is used  
14 for obtaining the period flux measurements contained  
15 variability that you would be experiencing when you are  
16 using it in a region as opposed to just at a particular  
17 field.

18 Let me just run through the rest of the slides  
19 and then I'll finish up on my comments.

20 But let's skip that one. Basically, all that  
21 was showing was if you have different amounts of organic

1 material in the soil you will get different emissions,  
2 which is fairly obvious.

3 I'm showing this slide for Randy Segawa. He was  
4 asking about what the permeability of high density  
5 polyethylene film would be for methyl iodide. It is shown  
6 here. This is for methylbromide. I think the value at  
7 25, I think is 4.3. And at 25 for methyl iodide it is  
8 1.0.

9 This basically means if you use high density  
10 polyethylene, you can expect to find that the emissions  
11 from flat fume with a high density polyethylene barrier  
12 will be more than what you would experience with  
13 methylbromide, assuming that the film is a controlling  
14 factor, which in our experiments and in our modeling that  
15 we have done, we have found that the film does tend to  
16 control the emissions.

17 You can see also there is a temperature effect.  
18 A very strong one. This is going across a whole number  
19 of chemicals, but even here there is -- I think if you go  
20 up 10 degrees you get about a 1.7 factor in the  
21 permeability.

1           So as you have daily cycles in the temperature,  
2   the temperature to the film is extremely hot. It is  
3   basically like a greenhouse effect.

4           And you get temperatures say on the order of 70  
5   degrees C. And that -- we don't even have permeability  
6   measurements at that high temperature. So the film really  
7   does have a lot to do with the emissions.

8           Can I see the next slide? I don't want to talk  
9   about that yet.

10          So I think that it's kind of important to  
11   include the spatial and temporal variability that occurs  
12   in flux measurements.

13          The difficulty, of course, is how is that going  
14   to be done these experiments are very expensive, very time  
15   consuming. And yet to get those -- I mean, there was like  
16   10 or 11 studies there, full studies. And that might be  
17   cost prohibitive.

18          Another question was whether a single study -- I  
19   think that pretty much has been answered. It is hard to  
20   believe a single study would be appropriate to generalize  
21   to a regional or maybe a state scale.

1                   Most of this I have already covered in talking  
2   about the slides. I think that it's good that you  
3   separate the analysis based on the fumigation type, where  
4   you have flat fume and you get flux values for the flat  
5   fume versus the raised bed and the drip.  
6   Although I did see some things in the data that you  
7   presented that kind of struck me as a little bit odd in  
8   just that -- I think it seemed to me that the drip had  
9   kind of a kind of a high emission rate, which a lot of  
10  people are saying that emissions tend to be lower with  
11  drip, which could be due to the partitioning of the  
12  chemical into the liquid. You put a lot more liquid into  
13  the soil, there is more partitioning to it.

14                  But anyway, that's good in the sense that the  
15  emission values that you are using are more appropriate  
16  for the analysis.

17                  What was the other question? Oh, about linking  
18  it. It seems to me that the approach -- I don't see any  
19  problem with being able to link applications assuming that  
20  the emission data is appropriate for each of the fields  
21  that are put in there.

1           And if you use some kind of a probabilistic  
2   form, well, then, it wouldn't even be an issue there. You  
3   could just sample from that probability distribution and  
4   just assume that the -- if the two fields would fit on  
5   that probability curve.

6           Then as far as with missing data, I'm not sure  
7   how to respond to that. In my reading of it, I didn't see  
8   where missing data -- except for maybe with the  
9   meteorological, which I'm not really the best person to be  
10   answering that question.

11           There was some discussion in the text about when  
12   you are missing data how to go about filling it in. You  
13   know it probably would be better for someone else to talk  
14   about that.

15           Then as far as the back calculation approach for  
16   estimating emission rates, this idea of using the  
17   regression I think is fine, especially when you have very  
18   small intercept.

19           But I wonder if it wouldn't be -- I think you  
20   could bypass this whole thing if you, instead of setting -  
21   - in the model setting a default emission rate and then

1 using the regression to try to figure out a way to scale  
2 it to what you observe in the measurements. It would seem  
3 like you could just at each location that you have a  
4 receptor you could find the emission rate that causes a  
5 match between measured and the model.

6 So what you would end up getting is a series of  
7 emission rates. You could have like -- for every receptor  
8 you would have the field emission rate that gives the  
9 match.

10 So say that you have 10 receptors, you get 10  
11 emission rates, you take the average. That should be the  
12 same as what you would get from your slope. Then you  
13 never really have to worry about an offset because it is  
14 not really pertinent. You also have a range in emission  
15 rates which gives you in a sense a standard error or some  
16 kind of an error measurement.

17 The only difficulty I can see in doing something  
18 like this would be those situations where at a receptor  
19 you have a measured concentration but the model can't give  
20 you anything but a zero.

21 And in a sense, that's -- I mean, that's going



1 to cause you -- that's what is causing the problem with  
2 the offset anyway, probably.

3 And it would seem that if you just looked at  
4 where the zone where the model -- where the plume is, you  
5 could exclude those points that are outside of it.

6 Otherwise, I don't see any real -- I can't think  
7 of a preferred way in terms of whether you should go  
8 through the origin or not. I think depending on what you  
9 think the cause is for that intercept -- If you look at it  
10 as a background concentration, then it seems like all the  
11 data should have that little background amount scaled out  
12 or subtracted out so it goes through zero.

13 If it is from some other factor, then it  
14 probably should be ignored. But who knows what the truth  
15 is.

16 So then I guess the last thing I would like to  
17 just at least show something would be this idea of using a  
18 soil based model. I will try to run through this quickly  
19 as I know we're running out of time.

20 Can I have the slides again?

21 This is some work that actually Dr. Wang has

1 worked with us when he worked in our group a few years  
2 back. What we're doing is using a fairly complex  
3 numerical model that describes water, heat and soil  
4 transport, the chemical transport would be methyl iodide  
5 in this case.

6 And historically, there have been -- actually,  
7 historically, there was one way in which this atmosphere  
8 soil boundary condition was characterized. And that's the  
9 one that says -- it is the boundary condition right here  
10 where the flux at the soil surface is equal to a mass  
11 transfer coefficient times a difference between the gas  
12 phase concentration in the soil and the atmospheric  
13 concentration.

14 This mass transfer coefficient has been  
15 parameterized by setting it equal to the gas phase  
16 diffusion coefficient in air and some boundary layer  
17 thickness.

18 This boundary layer thickness is something that  
19 is kind of arbitrary. That doesn't really have very good  
20 physical meaning.

21 If you have barometric pressure changes, you can

1 actually have air move through from the soil into the  
2 atmosphere or vice versa in which case a stagnant boundary  
3 layer doesn't really even make physical sense.

4 But this is still used in this particular  
5 model, and provides one way to estimate the emissions into  
6 the atmosphere. I have a slide which will  
7 show for allowing this to be temperature dependent or  
8 keeping this a constant, and compares it to some measured  
9 data that we collected in a methylbromide field experiment  
10 looking at flat fume shallow injection.

11 A new, newer, I guess I should say, boundary  
12 condition, this was developed by John Baker at Saint Paul,  
13 is shown here where you have micro meteorological  
14 information. You have the stability of the atmosphere.  
15 There is wind information, Reynolds number, Schmidt number  
16 (ph).

17 So basically, to use this boundary condition, if  
18 you have micro meteorological information, you can  
19 actually let what is occurring in the atmosphere control  
20 what is happening at the surface.

21 Can I have that next slide? So for that first

1 boundary condition, which is all soil base, you don't  
2 really have any knowledge of what is happening in the  
3 atmosphere.

4 For a constant mass transfer coefficient, you  
5 get a very simple, common flux curve that you would see if  
6 you were looking at isothermal conditions.

7 When you have allowed the temperature at the  
8 surface to be controlled by solar heating, you start  
9 seeing cyclic behavior. Some of this you actually see in  
10 the documents that we were given. This is kind of high at  
11 the beginning, decaying curve with cycles in it.

12 But you still see there is a fairly large  
13 mismatch between measurements and simulations. Now, if  
14 you allow atmospheric conditions to control what is  
15 happening, can I see the next slide, you can get a much  
16 better agreement of what is occurring.

17 You see this kind of sharp changes in behavior  
18 that you often see when you start looking at flux data  
19 that's taken at a higher frequency. So this would be an  
20 alternative. It is time consuming, there is some  
21 information that I know right now would not be available.

1     You couldn't go and plug this into PERFUM's tomorrow and  
2     get it to work.                     But it might be a direction  
3     that should be headed towards since it seems like the --  
4     there is just some -- it seems like this idea of  
5     transferring from one location to another is not going to  
6     be very easy to do using an indirect approach. You have  
7     to go out to a lot of fields.

8                 When you start looking at things like emission  
9     reduction strategies, it's going to be very difficult to  
10    get all the data you need in field studies. But in here  
11    all you have to do is change how you handle the boundary  
12    condition and then maybe run a study or two to verify that  
13    it makes some sense.

14                But anyway, there was some discussion earlier  
15    talking about whether this would work or not. I thought I  
16    would just put it up there so people could see that, at  
17    least in this case -- actually, I have done this for two  
18    situations, this study and looking at triallate (ph), which  
19    is a herbicide of lower volatility. It worked for that as  
20    well. That's only two studies which doesn't prove that it  
21    is right.

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1                   So I think that might cover everything. I think  
2 those are all my comments.

3                   DR. ROBERTS: If you think of something else,  
4 we'll open it up again.

5                   Dr. Bartlett I believe is next.

6                   DR. BARTLETT: I think I talked earlier about  
7 the first question a bit on the flux values and the  
8 standard error estimates and it has been covered a bit  
9 already.

10                  Then I guess I feel particularly wary about the  
11 limited amount of studies and the ability to generalize.

12                  And I believe you say -- and I totally  
13 appreciate how much the cost and effort to do each study.

14                  But there is quite a bit of variation in weather and  
15 time, which you see just in your monthly data itself.

16                  So that generates maybe another way of looking  
17 at variation too. But on the other hand, your approach  
18 tries to capture that.

19                  But there are significant regional differences  
20 in weather. And unfortunately, we can talk about later on  
21 the weather section, but there are other weather data

1 sources that might give us some idea of what some of those  
2 variations might be and do. So I guess I will delay it to  
3 the weather question.

4 As far as missing data, it seems like you are  
5 using standard approaches as far as missing weather data.

6 But of course you wouldn't do that if you were doing a  
7 study. During the study period, you wouldn't use missing  
8 data because some problems can happen there.

9 And the issue of going through the origin, I  
10 kind of agree it is either there is another factor at work  
11 or nonlinearity. It is not unusual to have nonlinearity  
12 as you approach zero.

13 I mean, I think it is very unusual to have  
14 linearity when you are close to zero. And if you have a  
15 lot of data points close to zero, that's a problem.

16 I mean, there are other ways to deal with  
17 detection limit problems than conventional, which is  
18 one-half zero and full and truncated data. There are ways  
19 to project, come up with other data points, the censor  
20 data techniques to develop that. I don't know if it's  
21 worthwhile to do that.

1           I have somewhat reluctance when you force it to  
2   zero, but you already have quite a bit of differences on  
3   your slopes from the different studies.

4           And there are, of course, ways to put the  
5   studies together and use the data as you get more.

6           As far as -- so I pretty much agree with a lot  
7   of the discussion of Dr. Yates. I'm much more familiar  
8   with the physical process soil emission modeling. It  
9   would be nice to know how this would compare with that.  
10   But maybe -- and I'm less familiar with volatiles, but it  
11   seems like a lot can be learned about that.

12           I guess that gets into the issue of  
13   generalization again. If you are going to different crops  
14   and canopy, I'm not sure if fumigation is applied to  
15   situations where you have crops already growing or is that  
16   always pre-planting?

17           DR. REISS: I believe it is mostly pre-planting.

18    I think they are also going to look at other  
19   applications. I don't know if there is anybody from  
20   Arvesta that wants to step in. But everything we have  
21   looked at to date has been pre-planted.



1 DR. BARTLETT: I guess kind of a general thing  
2 that I have been thinking about that applies to a lot of  
3 this is when the comment came up about mass balance. And  
4 I think what I'm trying to synthesize what is happening  
5 here now that I know more about this substance and  
6 methylbromide is that quite a bit of it is emitted within  
7 the first week or so.

8 And it gets to me, again, the question of the  
9 persistence of the substance in people's bodies that you  
10 may be focusing on the first 24 hours but you may have a  
11 lot less variation if you look at exposure of, let's say,  
12 a house in the boundary area for a week or something like  
13 that, because the variation seems to wash out after that a  
14 little bit.

15 DR. REISS: Can I answer that?

16 DR. BARTLETT: Sure.

17 DR. REISS: For methylbromide, they have looked  
18 at that in a different way than these source of models.  
19 DPR has calculated data or measured data in schools, for  
20 example, I think to get longer term averages.

21 I don't know what the case will be with methyl

1 iodide. It is a much less persistent compound.

2 But you really can't deal with those sort of  
3 recirculation issues very easily in a model. You could,  
4 but not in this type of model. So you would have to deal  
5 with that in another way.

6 So it is not a goal of the model to get at that  
7 long term or weekly concentration. But it is not  
8 necessarily something that's going to be ignored in the  
9 risk assessment process. It is just not being dealt with  
10 right now.

11 DR. BARTLETT: I think with each one of your  
12 field studies you have the first 24 hours but you go on  
13 beyond that. Right?

14 DR. REISS: That's correct.

15 DR. BARTLETT: So it would be of interest to  
16 look at what would you do if you looked at the longer  
17 period and you have more consistency -- does that account  
18 for a lot of the variance, I guess it is not clear to me,  
19 the variance between them?

20 And maybe the cumulative emission curve that we  
21 just saw there might bring your data together more. And

1 you are just seeing -- and then as far as exposure, it  
2 does make sense if people are living and working on the  
3 periphery to look at that as a cumulative exposure over  
4 that time period. I'm not sure if this is -- once it gets  
5 in your body how long does it stay there? Do we have some  
6 idea of that?

7 MR. DAWSON: Our toxicologist is sitting over  
8 here. It's relatively short lived. Again, as I said this  
9 morning, I think if it is capable of looking at --  
10 averaging over the several days based on what the data  
11 show, we potentially could be interested in that as part  
12 of what we would do in a risk assessment.

13 Don't forget, it is not only this particular  
14 case, but we're looking at all the different fumigants.  
15 They all have different toxicology profiles. So we would  
16 be interested in varying durations.

17 It is going to be dependent on the data. As we  
18 look across the different chemicals, you see a very  
19 different kind of emission profile. So we're definitely  
20 interested in that kind of capability.

21 DR. BARTLETT: I guess it is the opposite point

1 I made before when I was concerned about the one two hour  
2 peak as far as acute exposure and then as far as something  
3 that might build up within as a cumulative over a few  
4 days.

5 But I'm glad to hear you are looking at both.  
6 But that's something the model could generate as well.  
7 Right now you are doing 24 hours.

8 But in the field studies, I think it is  
9 essential that you continue to do the field studies in  
10 that way, but maybe presenting the data in cumulative  
11 emission.

12 DR. REISS: In terms of the people or the  
13 distances right surrounding the field, the model could and  
14 hopefully eventually will look at shorter and longer term  
15 durations to deal with other fumigants and other issues.

16 DR. ROBERTS: Dr. Majewski?

17 DR. MAJEWSKI: I think it is hard to add much  
18 more than what Dr. Yates already presented. I agree with  
19 most everything he said. All my questions, in fact, have  
20 been answered previously. But I would like to reiterate  
21 that or go through the questions as an exercise.

1                   What refinements are needed? Basically, I think  
2   as the model progresses more field data will help refine  
3   the model.

4                   I don't think that you can take results from one  
5   area and apply it to a completely different area or state  
6   like taking Central Valley data and applying it to  
7   Minnesota. I think that would be inappropriate use. You  
8   may be able to do that in various areas of the Central  
9   Valley.

10                  I think that's about all I can add.  
11   Like all models, I think it needs more field validation,  
12   basically.

13                  DR. ROBERTS: Dr. Small?

14                  DR. SMALL: I'll sort of reiterate and follow up  
15   a little bit on Dr. Yates' discussion of the mechanistic  
16   models. I think you are probably going to need some type  
17   of a dual strategy in which you are working on the  
18   mechanistic models as a long term objective.

19                  But it is probably the case that that's not  
20   going to be ready for short term use, so that you are  
21   still going to have to have some more empirical approach.

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I think given that, the indirect method is a reasonable place to start. I think it is a good approach for estimating the emissions at a particular site for a particular observation period.

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I think when you start applying it to different conditions and different sites, though, I think you recognize that the type of variabilities that you are representing aren't probably representative of those.

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And that gets back to the chart that we saw before.

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I think the model breaks down in a couple of different ways. One of them is that the standard error method that you have for uncertain emissions, if you apply them to individual time periods and you sample those independently, you get still a further smoothing.

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If you have, say, four time periods in a day or six time periods in a day, you start off with a coefficient of variation that's 20 percent, you are going to be dividing that by the square root of four or the square root of six in terms of your daily variability.

1           So I think that's part of the reason why you got  
2   those results where those upper tail distributions seem to  
3   be pretty flat and close in. Because I think you are  
4   doing a number of things that tend to be variance  
5   reduction procedures.

6           By assuming the independence, I think, if for  
7   some reason the actual emissions due to variability or  
8   site conditions are higher during one period, they are  
9   probably going to be higher during the next.

10          Now, once you get into sort of larger scale  
11   variations, normal distribution or even a T distribution,  
12   which is symmetric, really isn't going to fly because you  
13   are going to start generating negative values.

14          You are not really going to be able to characterize  
15   coefficients of variation of 50 percent or 100 percent  
16   with a normal distribution or a T distribution.

17          If you start using a log normal distribution,  
18   you might get into mass balance issues where you start  
19   generating more than one hundred percent of your  
20   application.

21          So I think one possible hybrid approach might be

1 to first consider the total mass emitted either during the  
2 entire period or during the first day, and use a  
3 distribution that would constrain that between zero and  
4 one. The logical choice for that is a beta distribution.

5           If you go back to the distribution that Dr.  
6 Yates fit with a log normal, if you want to constrain  
7 that, that's one approach. But if you want to constrain  
8 that between zero and one, you could use a beta  
9 distribution. His particular plot actually looked like a  
10 mixture of two beta distributions, because it wasn't  
11 really that smooth. But that's one approach.

12           Having that then as sort of characterizing your  
13 uncertainty in the total amount that is emitted, either  
14 over the entire period or over the first day.

15           If you want to work with a daily averaging  
16 period, then you might be able to take representative  
17 profiles that sort of start off a little higher in the  
18 beginning, have some diurnal variation to them and a  
19 random component, and somehow constrain them so that you  
20 end up with a sampled or simulated value of the fraction  
21 that was applied.



1 I'm kind of working out here as I go kind of an  
2 approach that might allow you to capture a larger amount  
3 of variability but still be consistent with the mass  
4 balance constraint. You don't want to emit more than you  
5 apply.

6 You might have caught in my questioning earlier  
7 that I was not enamored with the approach of using the  
8 second day's emissions, applying them to the omitted first  
9 day emissions because you got started later on.

10 I think that once you get into the second day,  
11 as you have shown, a significant portion of the mass has  
12 already been lost. So you would expect the emissions on  
13 the second day to be significantly lower during that time  
14 period than on the first day even though they are at the  
15 same time of the day.

16 DR. REISS: I may not have been clear about  
17 that.

18 We didn't miss any periods in the field studies.  
19 We didn't miss any periods during the field studies.

20 It is purely a matter of numbering the hours.  
21 Say if you had 19 hours in the first -- you know, it said

1     that we had 19 hours in the first day. That was 19 hours  
2     from the start of the application and the start of the  
3     measurements to just the beginning of the measurements for  
4     the next day.

5                 So when I'm using that first 24 hours, it is the  
6     first 24 hours after the application. So it isn't --  
7     we're not missing anything.

8                 DR. SMALL: Good. I think a little  
9     clarification would help on that issue.

10                I had some thoughts too on the intercept term.  
11     I think in your documentation you would be a little  
12     clearer as to what a nonzero intercept might represent  
13     physically, particularly on the background issue. I think  
14     that would help.                   A statistician would  
15     actually be baffled at the approach that you have taken  
16     here, which is one in which if you get a statistically an  
17     intercept which is statistically significantly different  
18     than zero, then you reestimate it and force it to be zero.

19                Because a statistician would say if it is  
20     statistically significantly different than zero that shows  
21     that it really is significantly different than zero and

1     you ought to allow it to be. In contrast to the case, if  
2     it is not statistically significantly different than zero,  
3     then you could argue let's force it to go through zero.

4             So you have taken an opposite approach that a  
5     statistician would take. I think whatever you end up  
6     doing with that, you need to motivate a little bit more  
7     with the physical reasons why you may have background  
8     concentrations, particularly if you have done multiple  
9     tests in an area at that site or at other sites.

10            DR. REISS: The physical reason is you can't use  
11     the intercept term to calculate the adjusted flux rate.

12            So you have the possibility of underestimating  
13     the flux rate if you have a positive intercept, which is  
14     typically the case if you have an intercept.

15            DR. SMALL: Unless you argue that it is from  
16     other sources other than from that field.

17            DR. REISS: I think that's highly unlikely. It  
18     could be from diffusion processes that are not accounted  
19     for in the model. I think that's more likely the reason.

20            So I think the reason that it is done is just  
21     not underestimate the flux rate. Make sure you predict

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1     those --

2             DR. SMALL:   It's conservative.

3             DR. REISS:   Yes, it is conservative and you make  
4     sure you predict those high values.   That's the idea.

5             DR. PORTIER:   Could I have a clarification  
6     question on that before you go on, if you don't mind?

7             DR. SMALL:   Yes.   Then I just have one minor  
8     last thing.   Go ahead.

9             DR. PORTIER:   When you do the regression and the  
10    intercept is significantly nonzero, is it normally  
11    positive or negative?

12            DR. REISS:   It is typically positive.

13            DR. PORTIER:   You are in good shape.

14            DR. SMALL:   That's common.   That's very common.

15     There is a number of reasons why that could happen.   Even  
16     if your errors in your measurements are non normal, you  
17     are never going to get negative concentrations.   You don't  
18     report negative concentrations.   You will sometimes report  
19     higher concentrations at those very low levels.

20            So you still have kind of a log normal error  
21     structure in your measurement rather than normal.

1 DR. PORTIER: That only happens because you get  
2 a lot of zeros. You can still have a negative intercept  
3 with a couple of positive zero values if you didn't have  
4 all those zeros at zero.

5 DR. REISS: It is possible. It is just much  
6 more likely --

7 DR. PORTIER: But if it were negative, then your  
8 statement wasn't true. By forcing it to zero, you would  
9 tend to underestimate the flux rate. So as long as you  
10 are on the positive side you are always okay.

11 DR. SMALL: I just have one last comment.

12 In a document, it brings up the issue of the log  
13 transformation before the regression.

14 That's inappropriate because the linear model is  
15 the mass balance model. If you want to be -- consider  
16 with just the physical superposition mass balance  
17 approach, you have to stick with the linear equation.

18 Now, if the errors are big, they may be non  
19 normal because you don't want negative concentrations.  
20 That's a much more sophisticated statistical model. You  
21 have to use numerical methods, numerical maximum

1     likelihood estimation techniques to do that.                     I  
2     don't know that it is worth it, given the sort of bigger  
3     picture that we have laid out here, which is that you are  
4     really not all necessarily all that interested in skinning  
5     the last bit of statistical accuracy for that study at  
6     that site but rather considering the variation from site  
7     to site that could occur.

8                 And I think this approach, in the long term,  
9     using a mechanistic model perhaps in the short term  
10    considering sampling your percent of your application that  
11    volatilizes and then distributing it in some reasonable  
12    way would be a good approach for that.

13                DR. REISS:   Thank you.

14                DR. ROBERTS:   Dr. Wang?

15                DR. WANG:    I would like to comment on PERFUM's  
16    capability to address multiple and linked application  
17    events versus the single source scenarios.

18                I think it is a big plus to be able to use this  
19    model to look at multiple events or applications at the  
20    same time.   And that should probably be explored a bit  
21    further beyond the example you presented.

1           It is true that wind direction -- one moment you  
2   may only have one wind direction in close proximity. But  
3   there could be -- that could change. And also from a  
4   probabilistic approach, that direction may not be just in  
5   one direction. It could be in -- covers a large region,  
6   angle in the angular orientation.

7           That means it is going to translate to the risk  
8   factors if you have closely related fields that's  
9   fumigated, and that maybe you want to expand on those  
10  areas.

11           Also, the distribution of those fields that  
12  maybe fumigated in a fairly close time period and how they  
13  going to be very uniformly distributed or very far apart  
14  like the example you presented.

15           They may be somewhat random or they may not be  
16  very random since the logistics of soil fumigation tends  
17  to optimize those operations.

18           So those things may need to be considered when  
19  you try to -- when you come up with some additional  
20  examples to help address those scenarios.

21           DR. ROBERTS: Are there other panel members that

1 would like to contribute comments? Dr. Winegar?

2 DR. WINEGAR: I would like to make a couple  
3 comments about -- first of all, using a single emission  
4 flux factor for a single study to apply to different  
5 regions of the country.

6 That kind of reminds me of the use of AP42 kind  
7 of emission factors where you look in a book and you have  
8 some type of emission source and you apply it to that  
9 source across the board.

10 In my experience in doing source testing for a  
11 fair number of different types of sources and trying to  
12 reconcile that with emission factors, it doesn't match up  
13 very frequently. And so -- I mean, you have heard other  
14 people comment about the same thing. So I'm just  
15 concerned about that.

16 DR. REISS: It is a good comment. When we have  
17 seven different studies, we have applied it for all the  
18 studies, and the range in the amount that was emitted  
19 during the first 24 hours is between 35 and 60 percent,  
20 that's a variability that we're concerned about and  
21 concerned about whether there is more variability there.



1           However, when you look at like AP42 factors,  
2   they can vary by orders of magnitude, the uncertainties on  
3   those. Just by mass balance considerations we're  
4   certainly doing better than order of magnitude. Whether  
5   it is possible that there is 70 or 80 percent example out  
6   there, that's possible.

7           But just by mass balance limitations it  
8   constrains the uncertainty in that direction.

9           DR. WINEGAR: And also from just a pragmatic  
10   point of view, questions will always be raised, well, you  
11   used it here versus -- and the study was there, was that  
12   any good.

13           So my other comment is in regards multiple  
14   application events. You stated that you thought it was  
15   pretty improbable for more than one event to be done --  
16   application to be performed within a certain area.

17           I got to thinking about some monitoring that I  
18   have performed for the alliance of the methylbromide  
19   industry a couple of years ago on behest of DPR where we  
20   did ambient monitoring for methylbromide in a couple  
21   locations. One of them was around Santa Maria.

1           There was one situation where we had -- there  
2   was a large field to the northwest of the center of town  
3   where there was multiple fumigations that had occurred.  
4   It was a large field, so they did it sections at a time.  
5   But they were like one day after the other.

6           Downwind, approximately a mile at least, much  
7   more than any of the buffer zones would have predicted,  
8   there was one of the highest concentrations that we  
9   measured. And so I think that scenario can occur in those  
10   kind of situations.

11           DR. REISS: Yes, I agree it can occur.

12           DR. WINEGAR: That was an example of how it  
13   really could be impacted.

14           DR. REISS: The ideal circumstance would be some  
15   sort of Monte Carlo model where you would randomly choose  
16   the probability. You would have a probability for there  
17   being a multiple application, how close it was to the  
18   source you were interested in and the direction from the  
19   source you were interested in.

20           I would like to develop a model like that to  
21   look at this problem, but I can't. I don't know how to

1 assign any of those probabilities. I don't know what data  
2 I can use to assign any of those probabilities.

3 I think we're going to have to look at more in a  
4 worse case situation. There are for methylbromide risk  
5 mitigation measures that restrict the intervals between  
6 applications when you have multiple fields and also the  
7 distance between applications, I believe, also, for  
8 different growers, for example.

9 Those mitigation options are obviously going to  
10 be on the table to deal with this issue.

11 DR. ROBERTS: Dr. Seiber?

12 DR. SEIBER: I would kind of chime in on the  
13 same point. I think you have probably the elements to do  
14 multiple sources. We did it in the Salinas Valley during  
15 a period of peak methylbromide fumigation way back in 1995  
16 and worked up the data after that.

17 We used ISC model and we also compared it with  
18 CALPUF just to get kind of a sense of which model would  
19 work better in that multiple source situation. We had  
20 eleven fixed monitoring sites. And we had roughly 20  
21 applications that were done within about a week or so

1 period.

2           So we basically said, well, if this field was  
3 treated four days ago and based on Yates' emission factors  
4 and ours and some of the others up there, here is kind of  
5 a boundary of what that field could be doing today, four  
6 days later.

7           And then we had a receptor at a fixed site. So  
8 you kind of -- what you basically do is add the  
9 contributions of multiple sites.

10           So it wasn't a perfect study. It had all kinds  
11 of limitation. Mainly a resource issue, but I think the  
12 approach it looked like it actually could work. And we  
13 were able to get a margin of exposure for the people in  
14 north Salinas, south Salinas depending on where they were  
15 relative to the monitoring sites.

16           So again, there is all kinds of flaws in it.  
17 But I think it is just a tour de force. If you can spend  
18 enough time on it, you can probably get it done.

19           In that case I had a graduate student that  
20 basically got burned up trying to calculate all the  
21 numbers. But at least he finally got it done.

1 DR. REISS: I think with multiple applications  
2 there is two issues really. The one which I have tried to  
3 deal with at least in the prototype scenarios we gave is  
4 how would another field affect the buffer zone for this  
5 particular field.

6 I think the question you are looking at is more  
7 the regional concentrations.

8 DR. SEIBER: Right, exactly.

9 DR. REISS: I think one of the other models you  
10 might hear from may look at that issue. But we didn't  
11 design this model to look at that issue. It is really  
12 totally different from a modeling perspective, a totally  
13 different scenario.

14 DR. ROBERTS: Dr. Spicer?

15 DR. SPICER: One of the comments I had was with  
16 regard to this plot of the measured concentration versus  
17 predicted concentration for these flux rates and this idea  
18 of the non zero intercept.

19 I think the point was made earlier that if you  
20 had -- if you did not include the non zero values in the  
21 intercept that, in essence, your slope was going to be

1 increased and therefore the flux estimates were going to  
2 be increased.

3 And I agree with that from just simply looking  
4 at the slope of the line. But I think the problem is  
5 that those non zero intercepts are really a failure of the  
6 dispersion model at that point. Because, in essence, what  
7 the concentrations are telling you is that the cloud is  
8 somewhere that's not predicted.

9 DR. REISS: I think it could be, and I'm  
10 speculating here, diffusive, some diffusive transfer,  
11 minor amount of diffusive transfer which causes those very  
12 small concentrations in the upwind direction.

13 DR. SPICER: Let's suppose you estimated the  
14 stability class incorrectly. Then the value of sigma Z is  
15 changed, then the dispersion estimates are off.

16 So there are all kinds of reasons why that would  
17 be the case. But it focuses on that dispersion question.

18 The point is that if there is concentration where the  
19 model predicts it, then no flux, ever how large in the  
20 model, will predict that mass there simply because the  
21 distribution coefficients are incorrect.

1           Now, what that means, though, is you have now  
2 missed some mass that's gone from the plot going downwind  
3 and not being predicted by the model. So there is some  
4 mass going downwind that is not included in the field that  
5 can be predicted by the model.

6           Because now what I'm doing -- what you are doing  
7 is you are fitting a straight line through the points that  
8 you have measured that the model tells you do exist.

9           Now by doing that, you correctly modeled the  
10 maximum concentration that the model sees. Therefore, you  
11 have the mass right that's inside the plume. But now  
12 there is an area outside the plume predicted by the model  
13 that you don't have. And therefore your flux estimates  
14 are too low.

15           DR. REISS: Well, I mean the flux estimate  
16 predicts the concentration in all directions. So you give  
17 it a flux estimate and the model will predict the  
18 concentration at all of the points around there.

19           DR. SPICER: No, I'm talking about getting the  
20 flux estimate from this plot. The point is that you are  
21 comparing them -- the maximum concentration that you

1 measure versus the maximum concentration that you predict.

2 And you are lining those two things up. You  
3 are saying to get that maximum concentration that I  
4 measure, I have to multiply the flux by .43 or .62 or  
5 whatever to get the right flux ratio based on my basis.

6 But that model prediction incorporates a certain  
7 amount of mass by virtue of the Gaussian dispersion model.

8 The point is you have now measured mass outside that  
9 distribution that's not being accounted for.

10 Therefore, your flux estimates are too low as  
11 opposed to being too high like it seems like everyone is  
12 thinking at this point.

13 DR. REISS: I'm not sure I agree with that. I  
14 think particularly when you constrain the intercept  
15 through zero you are deriving the flux that statistically  
16 best explains the data.

17 It is the one that minimizes that -- you have  
18 got the flux rate that minimizes the residuals between the  
19 predicted and the observed values. I can't see that that  
20 would be an underestimate in any way.

21 DR. SPICER: But the model values are always



1 going to predict a certain amount of error regardless of  
2 what the slope is. Because the model values that are at  
3 zero predicted concentration are simply going to add an  
4 increment to the mean square error when you actually do  
5 the minimization of the mean square error to get the least  
6 squares fit.

7 The point is that because those values are zero  
8 measured concentration, they are essentially thrown out as  
9 far as having any influence on the predictions concerned.

10 DR. REISS: No. Certainly, they are  
11 incorporated into linear regression and they are part of  
12 the residuals that you calculate.

13 DR. SPICER: But they will be a part of the  
14 residual that cannot change because the predicted values  
15 are always going to be zero, though.

16 DR. REISS: The residual could change, because  
17 if the predicted -- well, you are right. If you constrain  
18 it through zero, it will always be zero.

19 DR. SPICER: Therefore, if you leave those  
20 points out of the slope determination, it won't make any  
21 difference as long as the model predicted concentration is

1 zero.

2 DR. REISS: I'm not sure I have my mind wrapped  
3 around your question. I may need to think about it a  
4 little more. But I think from a risk assessment  
5 standpoint, the most important thing we're concerned about  
6 is that we're predicting that maximum concentration  
7 correctly.

8 That's what is going to be the risk driver. And  
9 whatever method we're going to use, we're going to  
10 optimize it to make sure we predict that maximum  
11 concentration accurately and we don't underpredict it.

12 DR. SPICER: There is no question about that.  
13 The point I'm trying to make is that your ultimate goal is  
14 not the maximum concentration. It is the flux. It is  
15 that slope of the line that you are using to say gives you  
16 these calibrated fluxes which I believe may actually be  
17 underpredicting the actual flux instead of overpredicting,  
18 which of course is the direction that you don't want to  
19 have.

20 If you were in a situation where you are always  
21 overpredicting the flux, overestimating it, then it would

1 be a conservative approach as far as the modeling is  
2 concerned.

3 But I'm afraid that by virtue of the fact that  
4 you have cloud where you don't have predicted cloud to be,  
5 you are missing some mass. Since you are missing some  
6 mass, you have to be missing some flux.

7 DR. REISS: You could be missing some mass in  
8 that upwind direction, but by virtue of doing a linear  
9 regression, you are getting some more -- you are getting  
10 that mass back in another direction by slightly  
11 overpredicting.

12 When you look at what is causing that intercept,  
13 it is extremely small concentrations compared to the  
14 maximum concentration that you are seeing in the field.

15 Whatever you do, whether it is a mass balance or  
16 a minimization of the mean square errors, those handful of  
17 maximum concentrations, three or four, are going to  
18 dominate whatever calculation you do.

19 DR. SPICER: I guess that's the whole point, is  
20 that I don't disagree in that regard. But what this does  
21 tell you is that it appears that what you are relying on

1 is the atmospheric dispersion to tell you something about  
2 the flux even though it is very close to the source.

3 And so the net result is that any sort of  
4 uncertainties in that atmospheric dispersion modeling  
5 right there is going to be an uncertainty that's reflected  
6 back into the flux rates.

7 And that's an issue that I think at this point  
8 in time no one knows which way it can go, because all you  
9 have are these concentration measurements at one single  
10 level.

11 If you had vertical concentration measurements  
12 somewhere, presumably, along the center line would be the  
13 best approach of course, but that's not always easy to do,  
14 but if you had those then you would have a much better  
15 estimate then of what the flux could be.

16 DR. REISS: We'll take that under consideration  
17 as we design more studies. Thanks.

18 DR. ROBERTS: Dr. Baker?

19 DR. BAKER: The emission fluxes are determined  
20 by the back calculation, which are based on point  
21 measurements and you were talking about consideration for

1 future field studies.

2 Is there any way to enhance or modify the  
3 program? Are there line methods that might be appropriate  
4 --

5 DR. REISS: Line methods?

6 DR. BAKER: -- suitable lasers where you get a  
7 line average concentration or something like that and you  
8 can vary that with height?

9 DR. REISS: No, I'm not -- I mean, if we were to  
10 do a variation by height, we would just likely set up  
11 different samplers at different heights. That would be  
12 the way to do it.

13 We are going to look at in a future a field  
14 study we have planned that's coming up, we're going to  
15 have an arc of monitors at a farther distance from the  
16 field to provide some further validation for how well it  
17 predicts concentrations farther down.

18 And that would, at least indirectly, get at the  
19 vertical dispersion coefficient as well.

20 DR. BAKER: Are they going to be vertically  
21 arranged?

1 DR. REISS: The plan right now is not to  
2 vertically arrange them. By virtue of the fact you are  
3 assuming some vertical dispersion and you have  
4 measurements that are significantly far apart vertically,  
5 horizontally, you would get some indication about bias  
6 that way.

7 DR. BAKER: There was a series of tests done  
8 several years ago with nitrogen tetroxide where they  
9 looked at different sampling arcs. I think it was 800  
10 meters and 150 meters or something like that.

11 The point is that the only thing that allowed  
12 you to sort out the distribution coefficients and those  
13 sorts of things was the fact that you did have the  
14 vertical measurements in addition to the horizontal ones.

15 DR. REISS: We'll take a look at it. I would  
16 note that this is a great data set in terms of looking at  
17 really even validating air models.

18 There are not a lot of data sets out there where  
19 you have on-site meteorological measurements and  
20 concentrations you can plug into your model. So it is  
21 something we look forward to putting out in the

1 literature.

2 DR. ROBERTS: Anymore comments on the issue of  
3 flux measurements? Let me ask the agency if the feedback  
4 and responses to this question were clear or whether you  
5 would like some clarification?

6 MR. DAWSON: Clear and thank you for a very  
7 informative look at this issue.

8 DR. ROBERTS: With that, then, let's go ahead  
9 and adjourn this session for today.

10 We will reconvene tomorrow morning at 8:30.  
11 There is plenty more discussion to go. I think we have  
12 five more questions.

13 I will look forward to seeing everyone at 8:30.

14 I would, however, like to ask the panel members to meet  
15 in a very short session in our meeting room here so we can  
16 discuss the issue of writing up the minutes.

17 With that, unless anyone needs to bring anything  
18 up, let's go ahead and adjourn for today and reconvene  
19 tomorrow morning at 8:30.

20 - - -

21 [Whereupon, at 5:20 p.m., the

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1 meeting recessed.]

2 -oo0oo-



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FRANCES M. FREEMAN