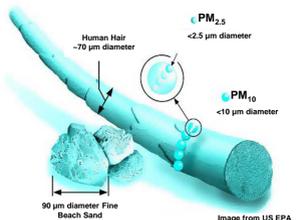


US EPA ARCHIVE DOCUMENT

ABSTRACT

In many epidemiological studies, particulate matter (PM) is associated with increased risk for adverse cardiopulmonary events. Due to cost and participant burden of indoor and personal measurements, health studies often estimate exposures using ambient measurements. However, ambient levels do not necessarily reflect personal exposures since indoor levels can differ from ambient levels, and people spend considerable time indoors. To reduce this exposure error, which adds uncertainty and bias to risk estimates, we are developing an exposure model for individuals (EMI) in cohort health studies. A critical aspect of EMI is estimation of the ambient contribution to concentrations within individual homes, where people spend most of their time. A mass-balance indoor air quality (IAQ) model was linked to an air exchange rate (AER) model to predict ambient-generated indoor PM_{2.5} mass from ambient concentrations, meteorology, and questionnaire data on housing characteristics and operation. First, AER model predictions were compared to 642 daily AER measurements across 31 detached homes in central North Carolina. For individual model-predicted and measured AER, median absolute difference was 43% (0.17 hr⁻¹). Second, IAQ model was evaluated with concurrent daily measurements of ambient and residential indoor-outdoor PM_{2.5} mass. Using cross validation, model predictions were compared to measurements of ambient-generated indoor PM_{2.5}, which were derived from sulfate (outdoor tracer of PM_{2.5}). For individual model predictions and measurements, median absolute difference was 20%. This study demonstrates the ability of EMI to predict residential indoor PM_{2.5} of ambient origin in support of developing exposure metrics for health studies.

AIR POLLUTION HEALTH STUDIES

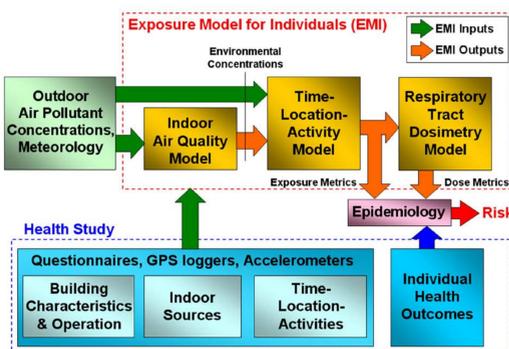


- Respiratory effects: reduced lung function, exacerbation of asthma
- Cardiovascular effects: myocardial ischemia, endothelial vasomotor dysfunction

Challenges of health studies:

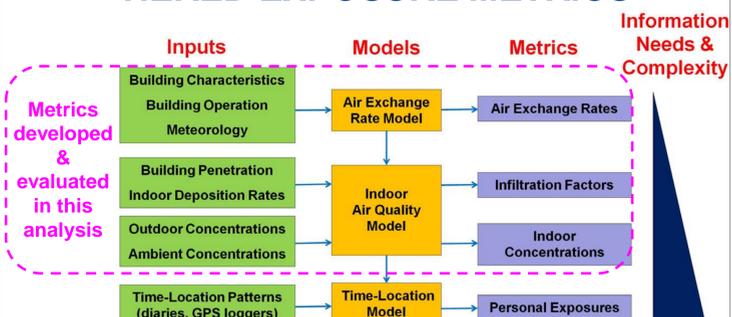
- Possible exposure errors from using surrogates (e.g., ambient levels) add uncertainty and potential bias to risk estimates
- Cost, participant burden of personal exposure monitoring

EXPOSURE MODEL FOR INDIVIDUALS

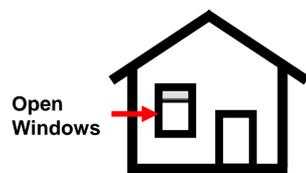


- Predicts exposure metrics for individuals in cohort health studies
- Inputs include outdoor concentrations, meteorology, and questionnaires

TIERED EXPOSURE METRICS



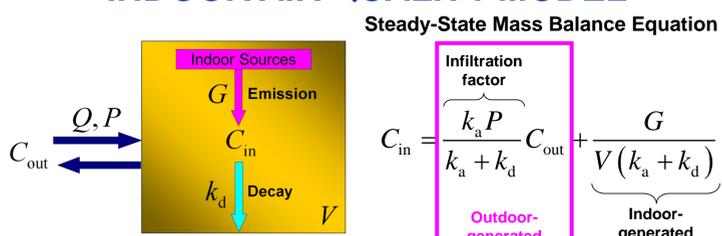
AIR EXCHANGE RATE (AER) MODELS



Air Infiltration Pathways Natural Ventilation Pathway

Models	Air Infiltration	Nat. Vent.	Meteorology
SF (Breen et al)	Yes	No	No
LBL (Breen et al)	Yes	No	Yes
LBLX (Breen et al)	Yes	Yes	Yes

INDOOR AIR QUALITY MODEL



where: C_{in} = home indoor PM_{2.5} concentration (μg/m³)
 C_{out} = home or ambient outdoor PM_{2.5} concentration (μg/m³)
 P = PM_{2.5} penetration coefficient (dimensionless)
 k_d = PM_{2.5} indoor decay rate (h⁻¹)
 k_a = air exchange rate (h⁻¹) = Q/V
 G = indoor-generated PM_{2.5} source strength (μg/h)
 Q = air flow rate through building (m³/h)
 V = building volume (m³)

CASE STUDY: RTP PM PANEL STUDY

Daily measurements (24 h avg.) at 31 detached homes for 7 consecutive days in 4 consecutive seasons (2000-2001)

- PM_{2.5} mass: ambient, home outdoor & indoor
- Sulfate: home outdoor & indoor
- Home air exchange rates
- Questionnaires (e.g., building characteristics)

Exposure Metric

Models

Air exchange rates (AER)

SF, LBL, LBLX
(Breen et al)

PM_{2.5} mass infiltration factors

$$\hat{F}_{inf} = \frac{k_a P}{k_a + k_d}$$

Indoor PM_{2.5} mass concentrations (from home outdoor meas.)

$$\hat{C}_{in}^{home_out} = \hat{F}_{inf} C_{home_out}$$

Indoor PM_{2.5} mass concentrations (from ambient outdoor meas.)

$$\hat{C}_{in}^{amb_out} = \hat{F}_{inf} C_{amb_out}$$

CROSS VALIDATION FOR PREDICTIVE INDOOR AIR QUALITY MODEL

Method to (1) evaluate model *output uncertainty* with independent data not used for parameter estimation, and (2) estimate *parameter uncertainty*

- Parameter uncertainty
 - Removed samples from one home at a time (validation sample), and estimate parameters with remaining data
 - Evaluated model with validation sample
 - Calculated Jackknife estimates and confidence intervals for parameters (P, k_d)

Number of homes	Number of samples	P	95% CI of P	k_d	95% CI of k_d
31	591	0.84	(0.74, 0.93)	0.21	(0.13, 0.29)

Model Evaluation Metric

$$|\mu| = \frac{|\hat{k}_a - k_a|}{k_a} \times 100$$

$$|\mu| = \frac{|\hat{F}_{inf} - F_{inf}|}{F_{inf}} \times 100$$

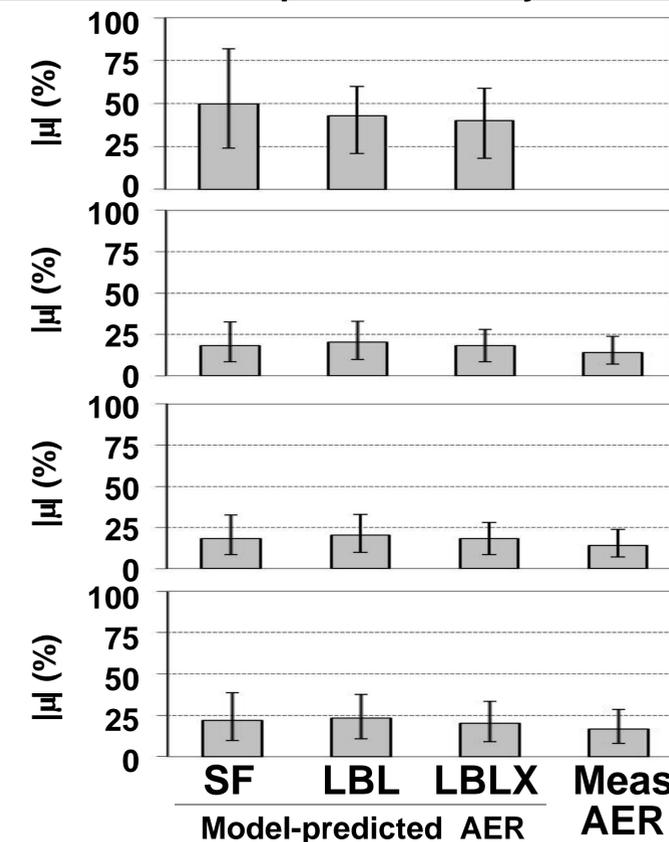
where: $F_{inf} = \frac{C_{sulfate, in}}{C_{sulfate, out}}$

$$|\mu| = \frac{|\hat{C}_{in}^{home_out} - C_{in}|}{C_{in}} \times 100$$

where: $C_{in} = F_{inf} C_{home_out}$

$$|\mu| = \frac{|\hat{C}_{in}^{amb_out} - C_{in}|}{C_{in}} \times 100$$

Output Uncertainty



CONCLUSION

- Demonstrates feasibility of developing tiered exposure metrics for individuals in health studies
- Sensitivity analysis showed 40% error of predicted AER yields only 12% error for predicted indoor conc.

- Median $|\mu|$ for indoor conc. increased by 2-4% using ambient, instead of home, outdoor meas.
- Median $|\mu|$ for indoor conc. varied by 2-3% using different AER models

REFERENCES

Breen et al. Environ. Sci. Technol. 44:9349-9356, 2010.
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