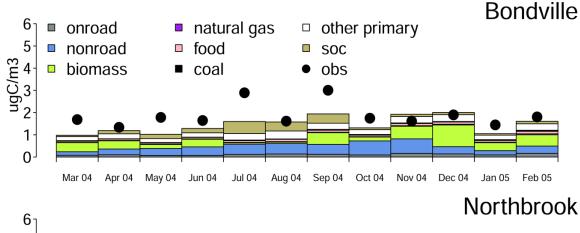
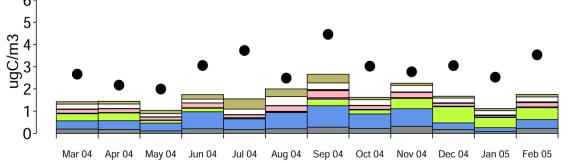
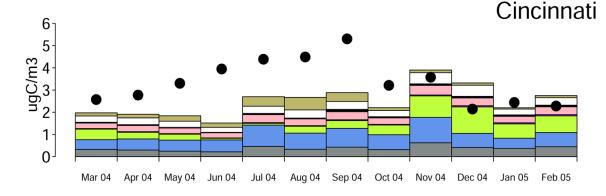
US ERA ARCHIVE DOCUMENT

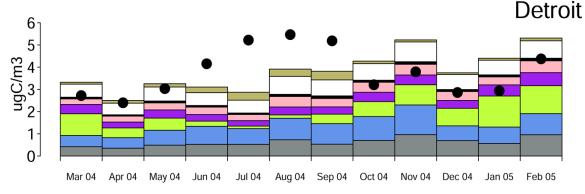
What can year-long hourly measurements tell us about emissions? Ann Arbor, MI, 4-5 March 2014 Benjamin de Foy and Jamie Schauer Saint Louis University / University Wisconsin - Madison This research is funded by U.S. EPA - Science To Achieve Results (STAR) Program Grant # RD 83455701

CMAQ model contributions to total organic and elemental carbon by month and site with observations









iagnostic Air Quality Model Evaluation
Source-Specific Primary and
econdary Fine Particulate Carbon
apelenok et al., ES&T 2013

Inverse Modeling: Mercury in Milwaukee

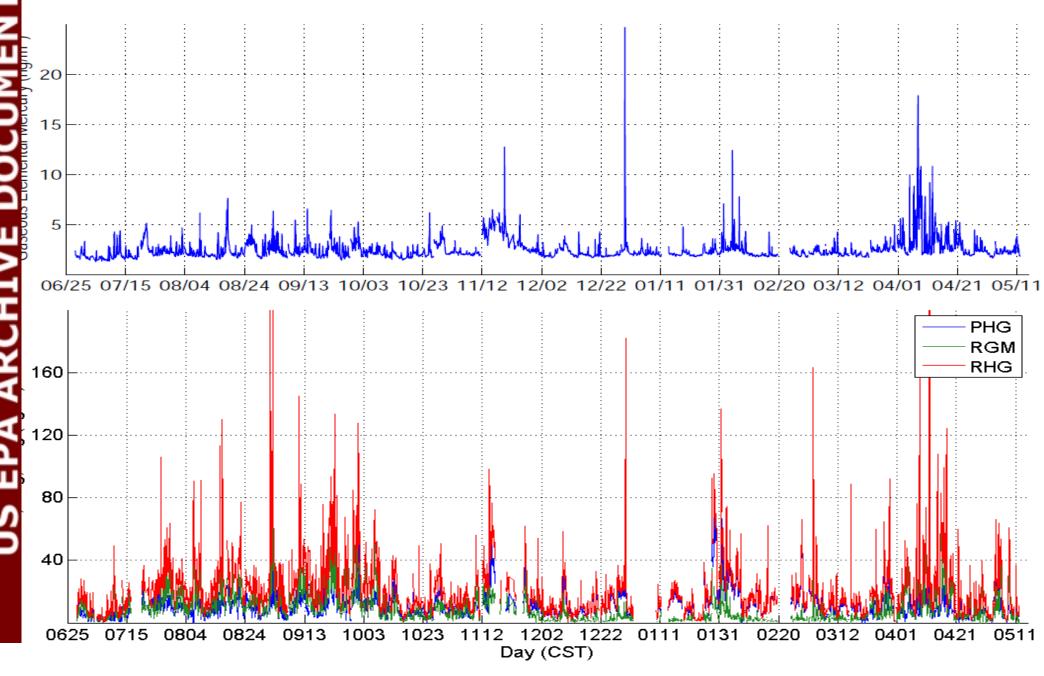
Gaseous Elemental Mero Milwaukee impacted by: **Gaseous Elemental Mercury at**

- Local urban sources
- Ohio River Valley + regional Ohio River Valley sources
 Forest fires
 Lake outgassing

3. de Foy, J. Heo, and J.J. Schauer, "Estimation of direct emissions and atmospheric processing of reactive mercury using inverse modeling," Atmospheric Environment, 2014.

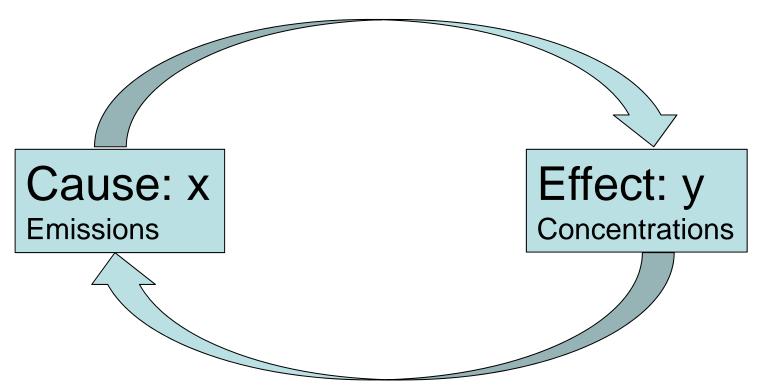
3. de Foy, C. Wiedinmyer, J.J. Schauer, "Estimation of mercury emissions rom forest fires, lakes, regional and local sources using measurements in Milwaukee and an inverse method," Atmospheric Chemistry & Physics, 2012.

Speciated Mercury Measurements in Milwaukee



Inverse Modeling





Inverse Problem: $x = H^{-1}y$

ktbook Intro: "Parameter Estimation and Inverse Problems" by Aster, Borchers and Thurber aphical Intro: "A Conceptual Introduction to Geophysical Inversion" by Andy Ganse, U. Washington

Inverse Modeling: Bayesian Formulation Simplifies to Least-Squares Inversion when Error Covariances are Diagonal

ayesian Formulation:

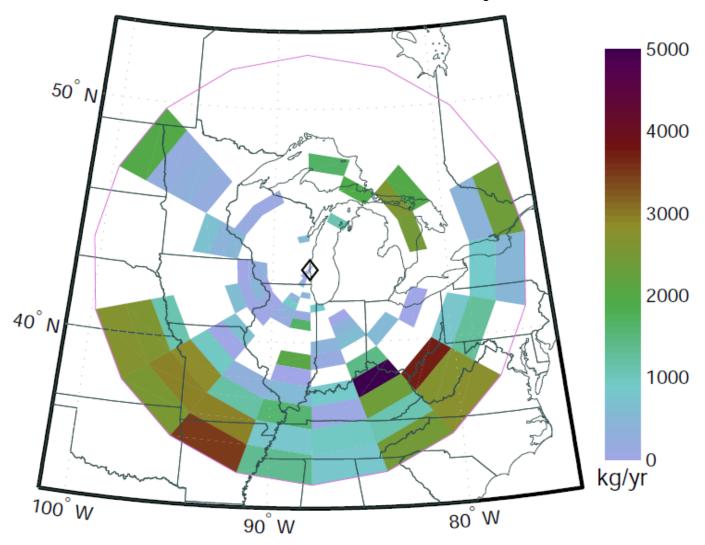
$$I = (\mathbf{H}x - y)^T \mathbf{R}_a^{-1} (\mathbf{H}x - y) + x^T \mathbf{R}_b^{-1} x$$

Simplifies to:
$$J = (\mathbf{H} \boldsymbol{x} - \boldsymbol{y})^T (\mathbf{H} \boldsymbol{x} - \boldsymbol{y}) + \alpha^2 \boldsymbol{x}^T \boldsymbol{x}$$

Solution in a single step of least-squares:
$$J = \left\| \mathbf{s} \cdot (\mathbf{H}''\mathbf{x} - \mathbf{y}'')
ight\|_2$$

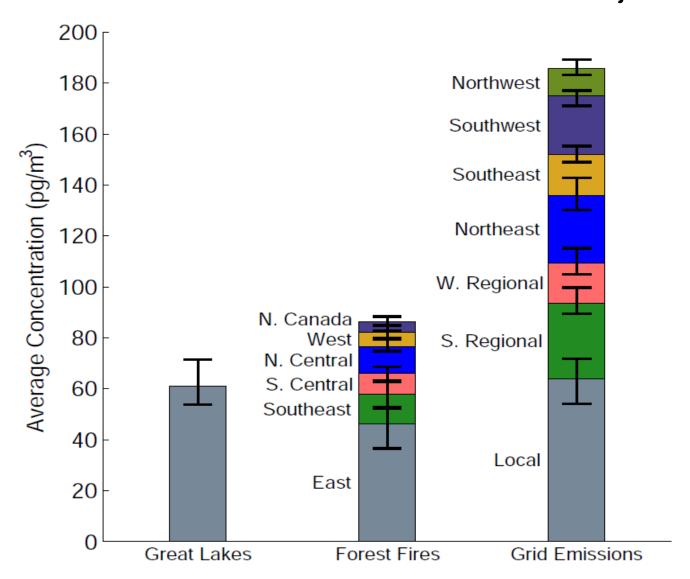
B. de Foy, C. Wiedinmyer, J.J. Schauer, "Estimation of mercury emissions from forest fires, akes, regional and local sources using measurements in Milwaukee and an inverse method," Atmospheric Chemistry & Physics, 2012.

Gridded Emissions of Gaseous Elemental Mercury Estimated from Back-Trajectories



B. de Foy, C. Wiedinmyer, J.J. Schauer, "Estimation of mercury emissions from forest fires, akes, regional and local sources using measurements in Milwaukee and an inverse method," Atmospheric Chemistry & Physics, 2012.

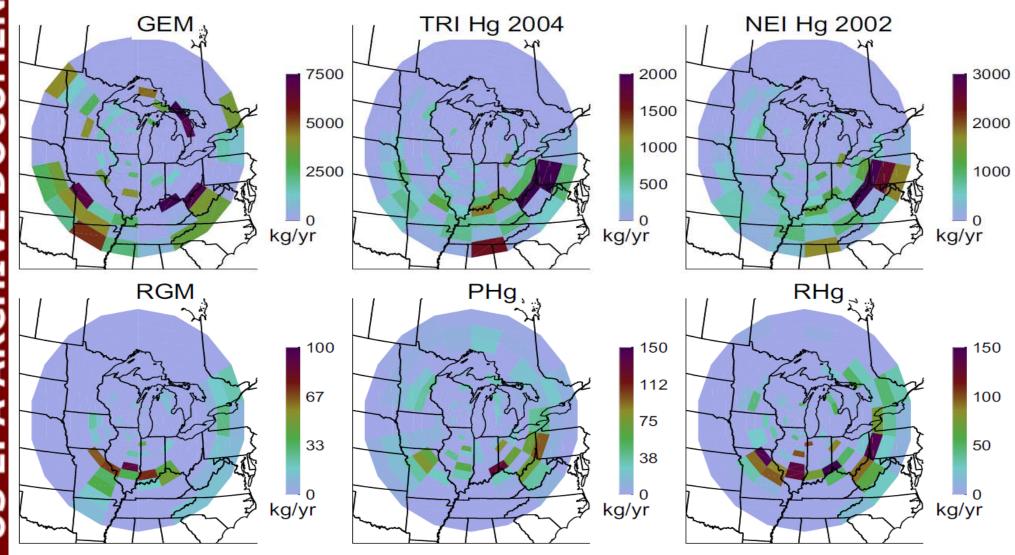
Source Group Impacts on Gaseous Elemental Mercury in Milwaukee Based on CAMx Simulations and Back-Trajectories.



B. de Foy, C. Wiedinmyer, J.J. Schauer, "Estimation of mercury emissions from forest fires, akes, regional and local sources using measurements in Milwaukee and an inverse method," Atmospheric Chemistry & Physics, 2012.

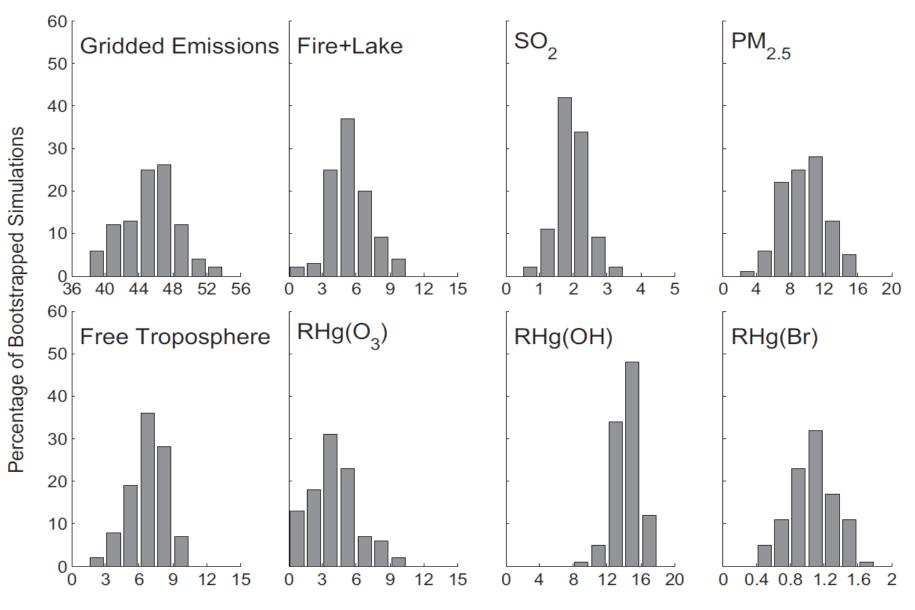
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Gridded Emissions of Speciated Mercury Compared with TRI and NEI



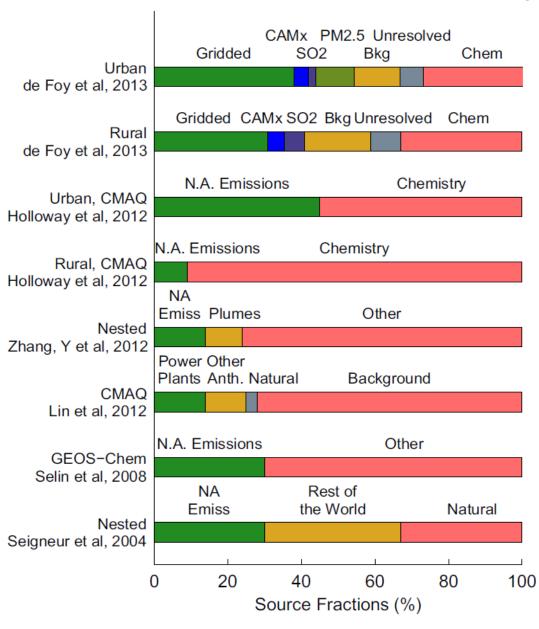
B. de Foy, J. Heo, and J.J. Schauer, "Estimation of direct emissions and atmospheric processing of reactive mercury using inverse modeling," *Atmospheric Environment*, 2014.

Contributions to Reactive Mercury in Milwaukee Uncertainty Analysis using Bootstrapping



Source Term Contribution to RHg Estimated by the Inverse Model (%)

Source Attribution of Reactive Mercury: Current inverse model suggests that a greater fraction is directly emitted compared to previous modeling studies



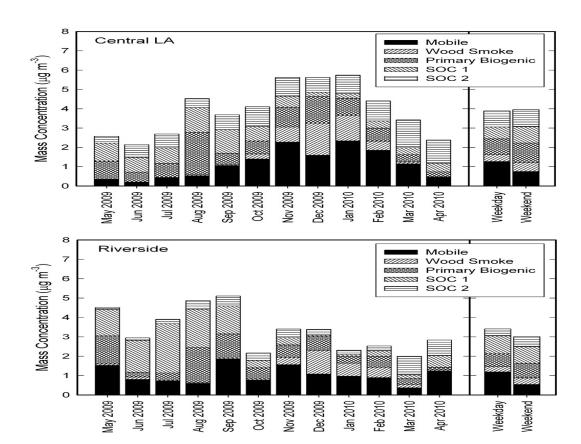
Los Angeles: **PMF Source** Apportionment of PM2.5 **Organic Carbon**

Heo et al., "Source appoint

M2.5 organic carbon using molecular

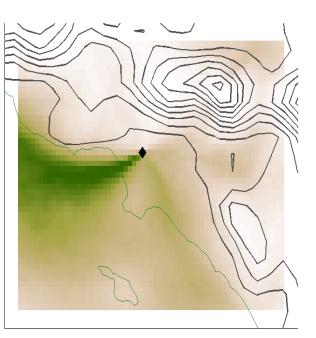
Matrix Factorization an Heo et al., "Source apportionments of arker Positive Matrix Factorization and mparison of results from different eptor models"

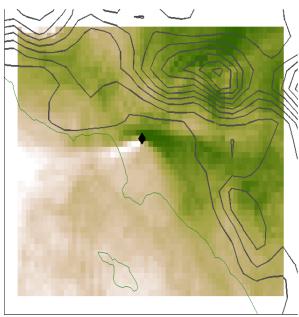
tmospheric Environment, 2013.

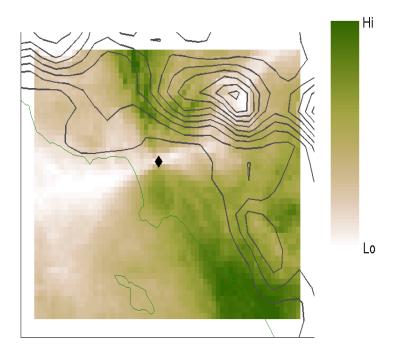




Back-Trajectory Analysis for Los Angeles: Carbon Monoxide, May 2009 – April 2010







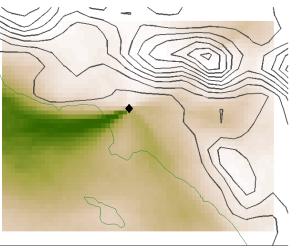
Residence Time
Analysis shows
dominant flow pattern:
westerly flow from the
Pacific Ocean

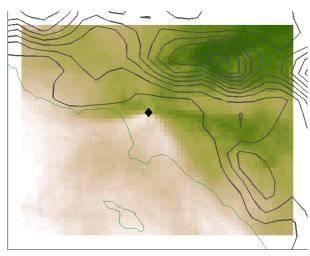
Concentration Field
Analysis shows high
CO concentrations are
associated with
downslope winds and
night time drainage
flows.

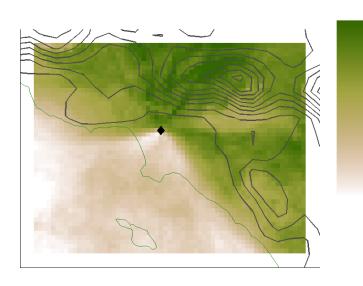
Column Concentration Field Analysis shows larger amounts of CO in the boundary layer are associated with flow from the South Coast and from the Central Valley.

Back-Trajectory Analysis for Los Angeles: PMF using 24hr samples, May 2009 – April 2010

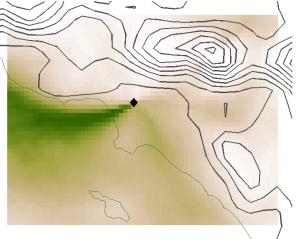
Mobile Marker

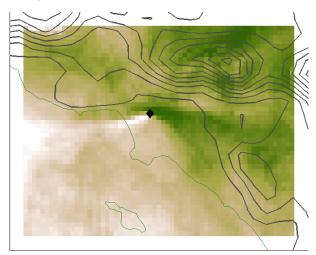


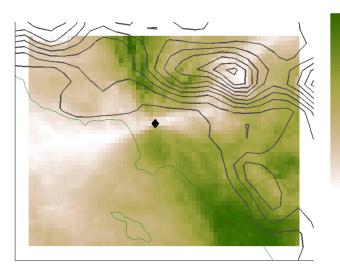




Carbon Monoxide (Hourly AQS Data)

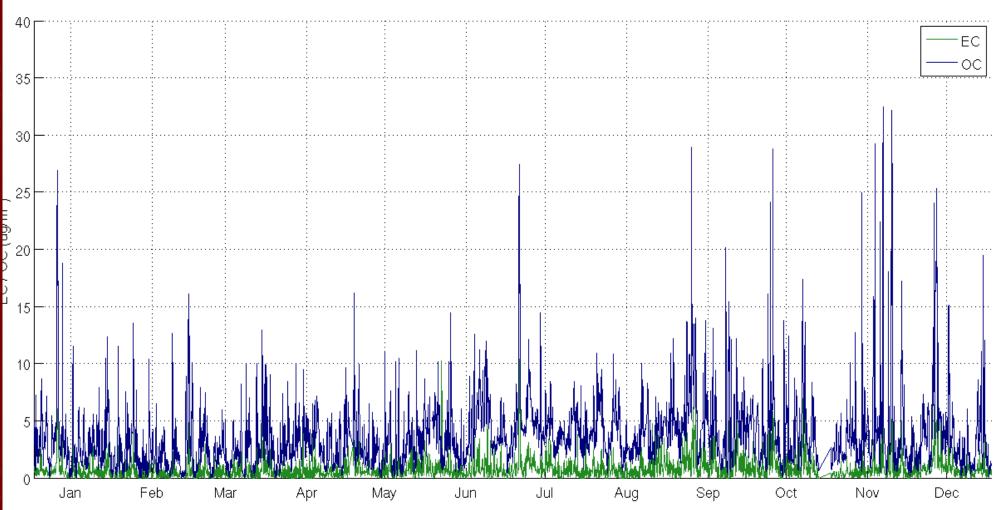




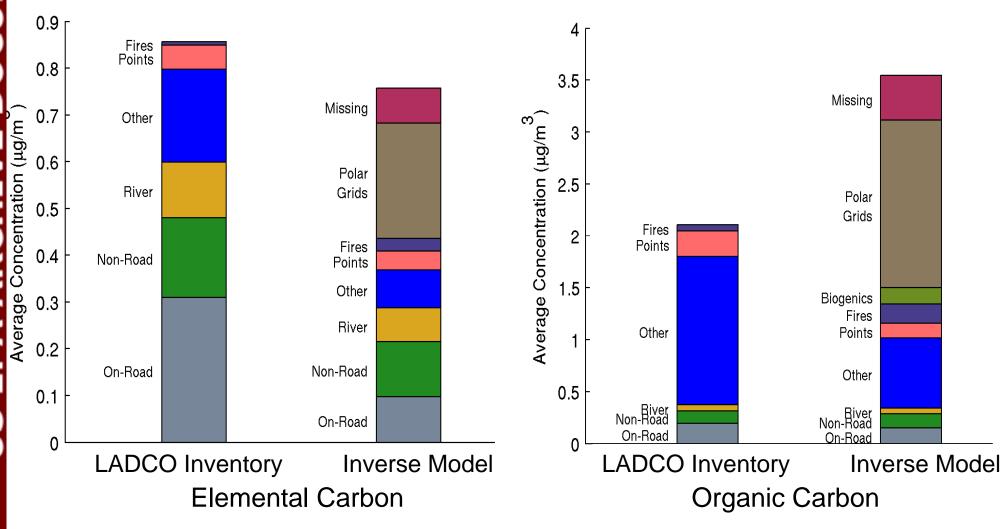




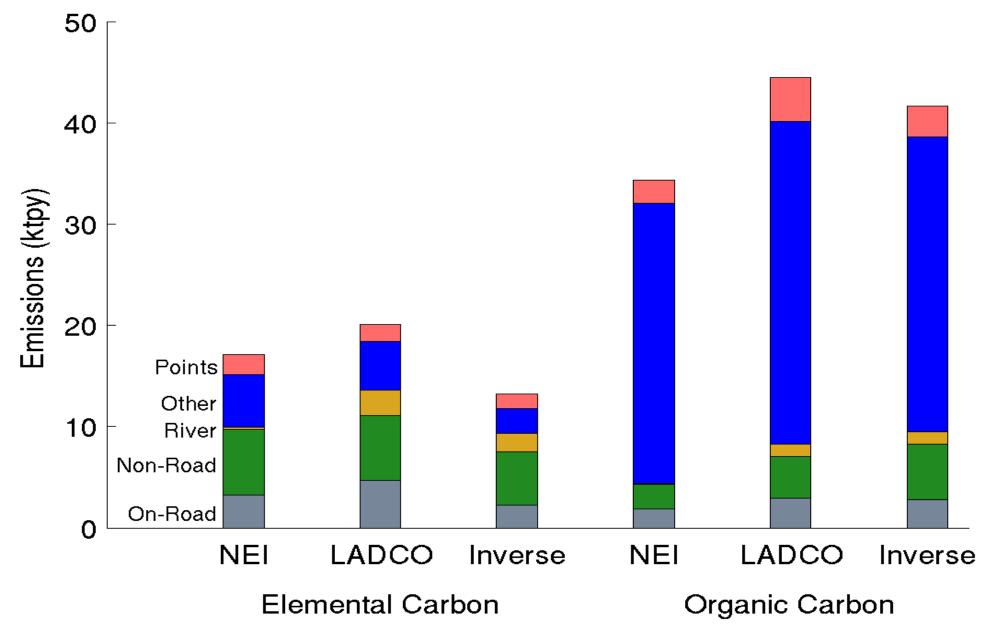
East – St. Louis Supersite: Continuous Hourly EC/OC Measurements for 2002



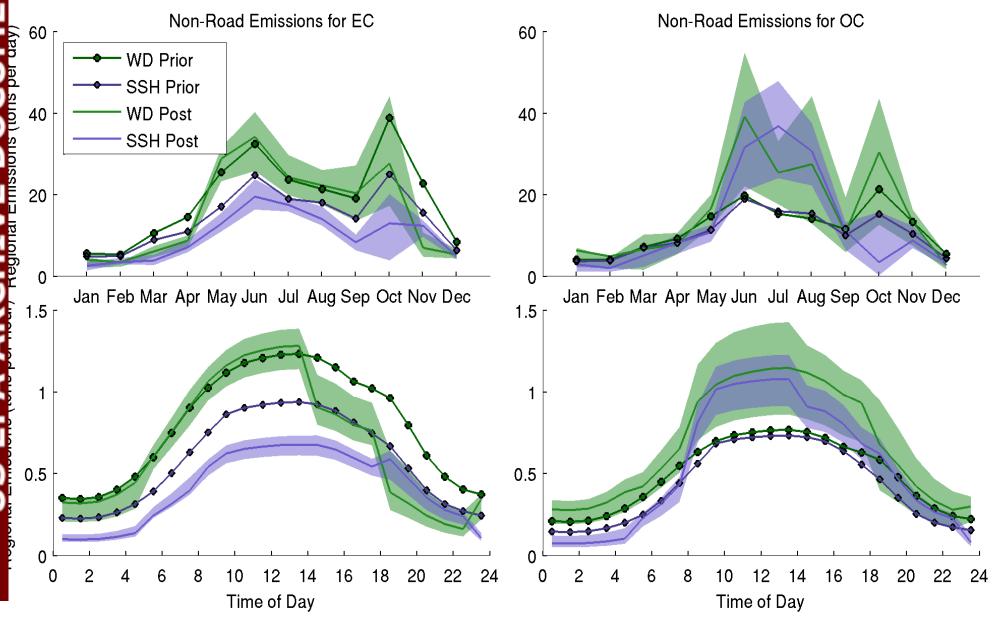
Contributions by Source Types using the LADCO Inventory and the Inverse Model Results



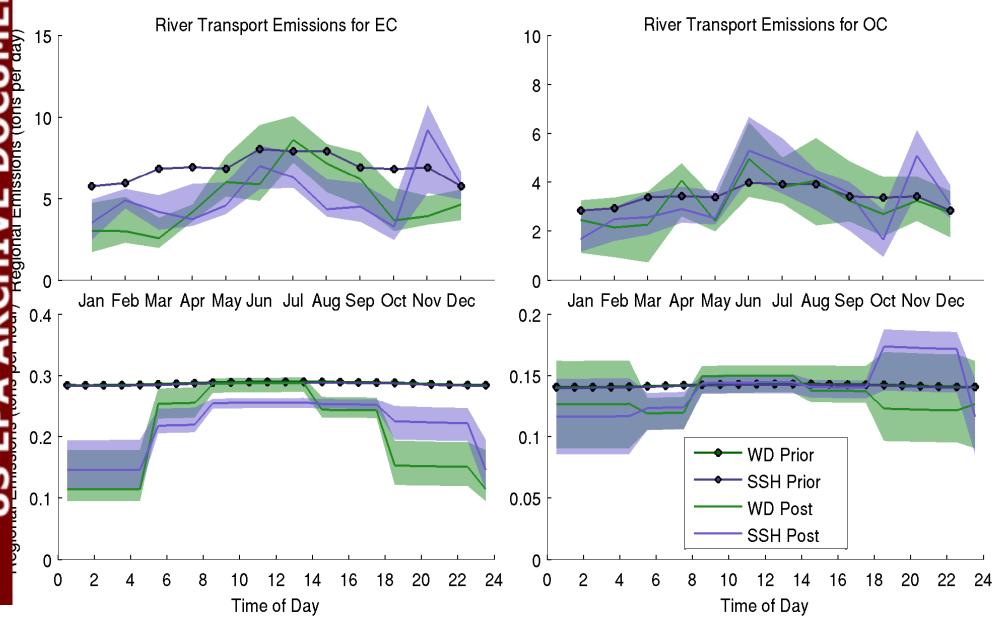
Emissions by Source Type For St. Louis and the Surrounding Region



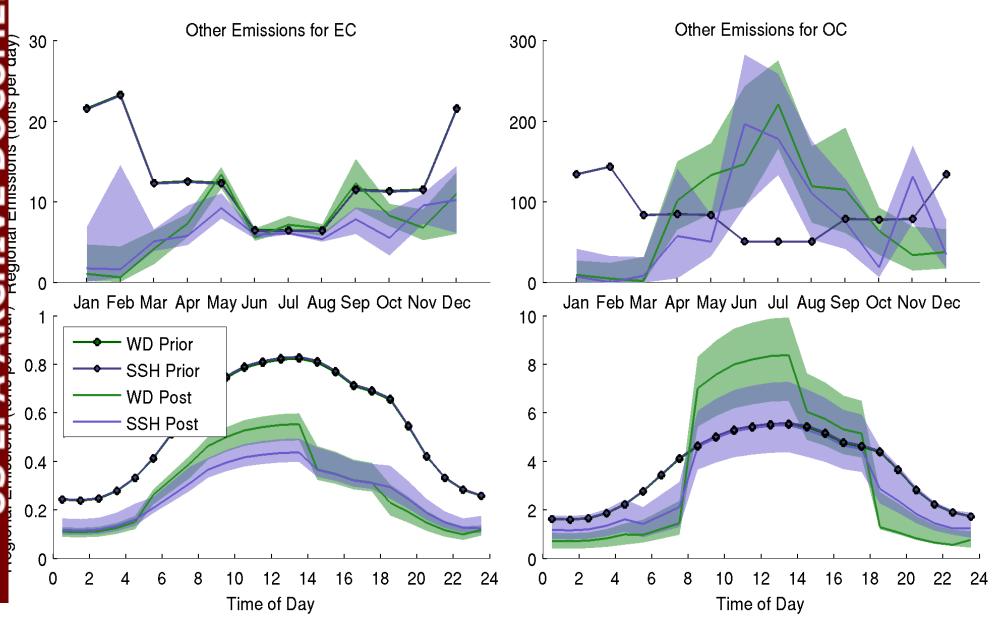
Diurnal and Monthly Emission Profiles for Non-Road Emissions LADCO Prior Inventory shown with Solid Markers Inverse Model Range based on Bootstrapping shown with Shading



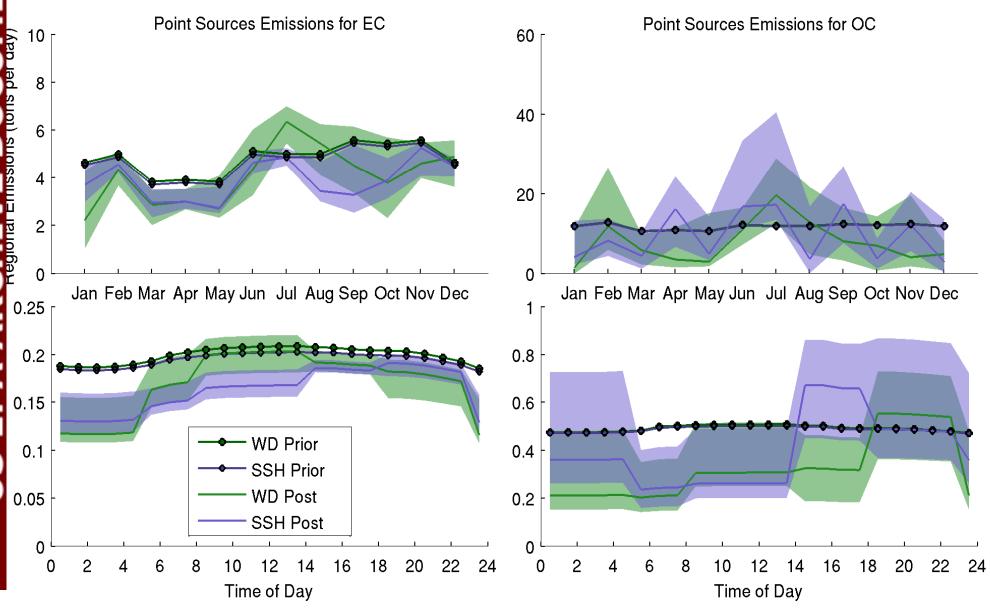
Diurnal and Monthly Emission Profiles for "Marine/Aircraft/Rail" LADCO Prior Inventory shown with Solid Markers Inverse Model Range based on Bootstrapping shown with Shading



Diurnal and Monthly Emission Profiles for "Other" Emissions LADCO Prior Inventory shown with Solid Markers Inverse Model Range based on Bootstrapping shown with Shading



Diurnal and Monthly Emission Profiles for Point Source Emissions LADCO Prior Inventory shown with Solid Markers Inverse Model Range based on Bootstrapping shown with Shading



Diurnal and Monthly Emission Profiles for On-Road Emissions LADCO Prior Inventory shown with Solid Markers Inverse Model Range based on Bootstrapping shown with Shading

