











Chemical transport models for exposure assessment

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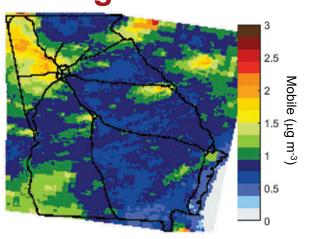
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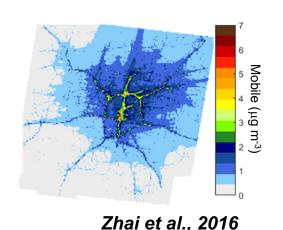
Modeling Atmospheric Scales

Regional



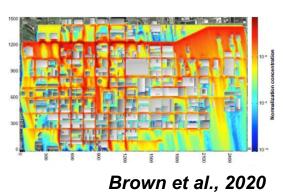
Zhai et al., 2018 >400km

Neighborhood



~1-100km

Roadway



<100m

Decreasing Spatial Resolution

Chemical Transport Model

- Atmospheric chemistry
- Transport processes
- All pollutants simulated!

Dispersion (or CFD)

- Some turbulence
- Single pollutant
- No chemistry



Spatial Resolution of Health Studies

County - uniform throughout state

GA county resolution ~15-40km

Zip code - based on population density

Downtown ATL ~4km, ATL MSA ~13km, rural GA ~25km

Census tract - based on population density

Downtown ATL ~1.5km, ATL MSA ~15km, rural GA ~30km

Increased spatial resolution for health analysis

Location of residents, gridded data (250m)

Spatial resolution of exposure estimates = health resolution!

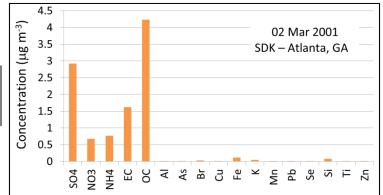


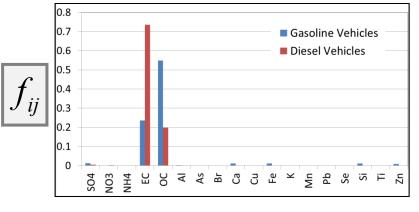
Source Impacts (Chemical Mass Balance)

$$C_i = \sum_{i}^{I} f_{ij} \cdot S_j + e_i$$

<u>Inputs:</u>

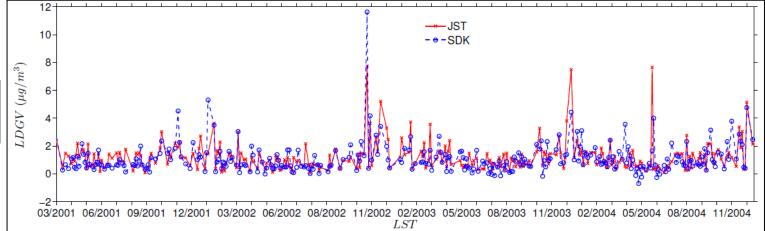
 C_i





Result:

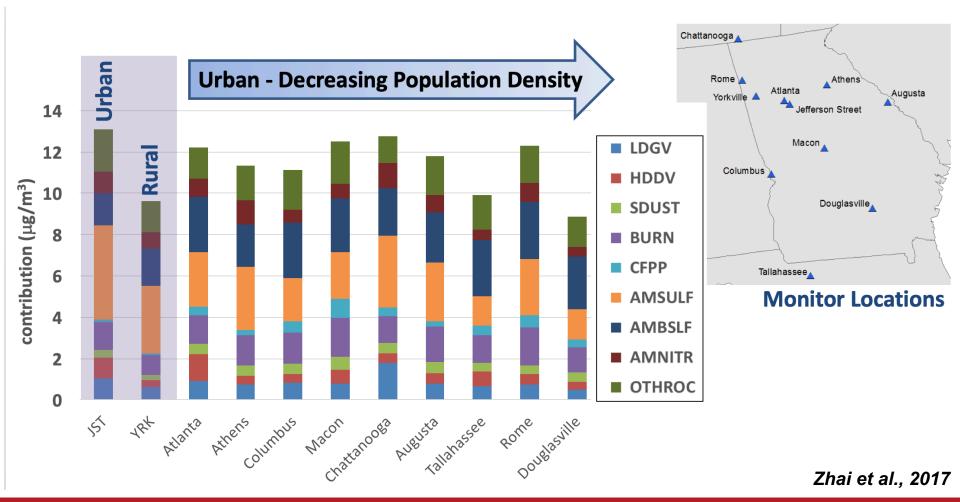






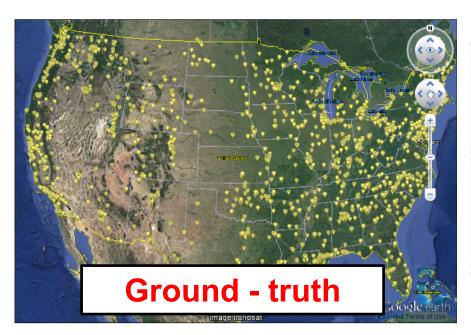
Georgia Source Apportionment

- Average CMB results, 11 monitors 2002 to 2010
- Higher mobile contributions in urban area





Ground Observations and AQ Model





Environmental Protection Agency (EPA)

Ground Based Stations
Surface PM_{2.5} Concentrations

National Aeronautics and Space Administration (NASA)

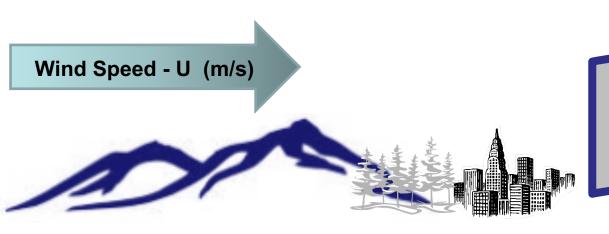
Chemical Transport Modeling Global Aerosol Concentrations

http://gmao.gsfc.nasa.gov/research/aerosol/modeling/nr1 movie/aerosols geos5.mp4



Numerical Weather Prediction

- Weather Research & Forecasting (WRF) Model
 - Late 1990's, National Center for Atmospheric Research (USA)
- Numerical Simulation of Atmospheric Flows
 - Meters to Hundreds of Kilometers
 - Navier-Stokes Equations
 - Inputs: Terrain, Land Use, Observations
 - Outputs: Atmospheric Physics, Moisture, Temperature,
 Precipitation, Radiation



Numerical Simulation of Physical Processes and Meteorology



Emissions Modeling

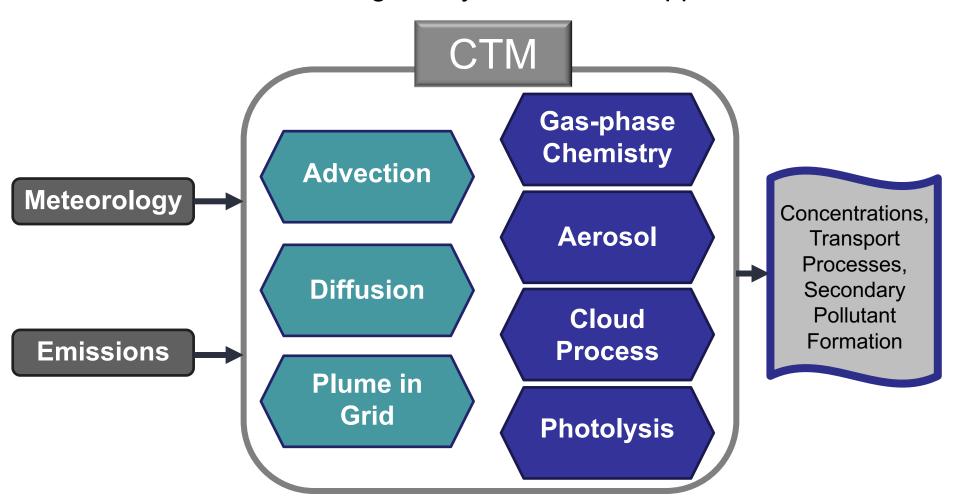
- National Emissions Inventory (NEI)
 - USA Environmental Protection Agency (EPA)
 - States report annual average by county
- Sparse Matrix Operator Kernel Emissions (SMOKE)
 - Model spatial and temporal patterns
 - Area, biogenic, mobile, point source emissions
 - Emissions factors for each source, species

Hourly Speciated Emissions
Rate for Each Grid Cell



Chemical Transport Modeling (CTM)

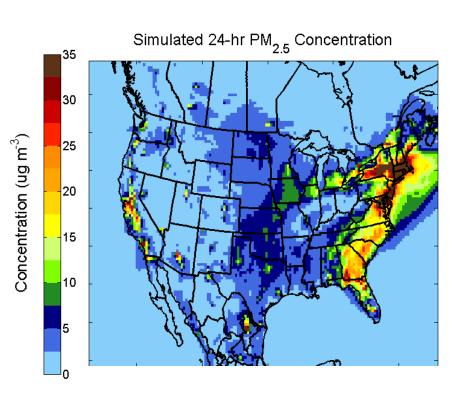
- Community Multiscale Air Quality (CMAQ) Model
 - USA-EPA Model, Regulatory & Research Applications

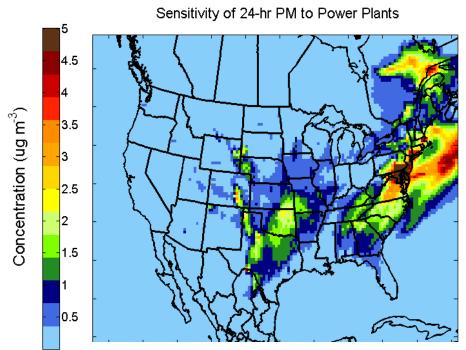




CMAQ-Direct Decoupled Method (DDM)

CMAQ Sensitivity to Emissions Perturbations 36km Resolution, Contiguous USA Domain







Spatial Source Apportionment Model

Receptor Models PMF, CMB, CMB-GC

- Monitoring network data
- Spatially limited
- Different results for each method

Source Oriented CMAQ-DDM

- Spatiotemporal source impacts
- Atmospheric processes modeled
- Results do not match observations

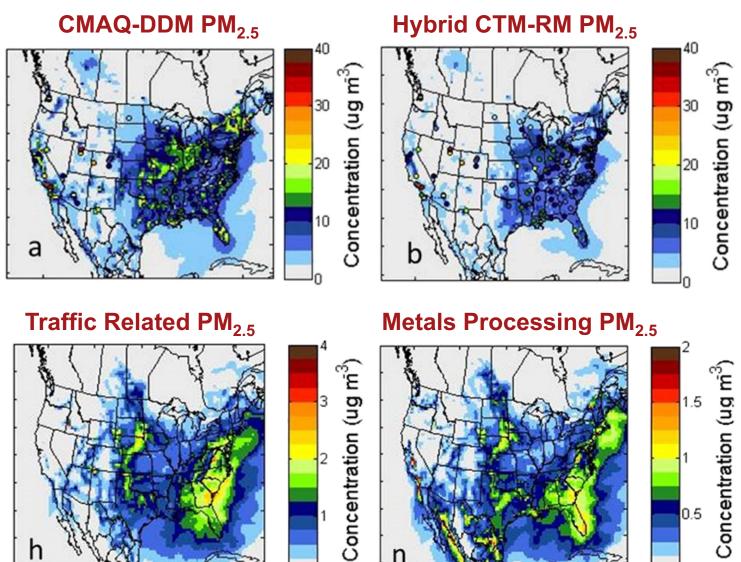
Novel SA Method

Hybrid CTM-RM

Daily and Spatial Field



Spatial Source Apportionment Model



Ivey et al., 2015



CTM Modeling for Exposure Estimates

Community Multiscale Air Quality (CMAQ) Model

Species Concentrations, No Metals (CMAQ)

- 12km Eastern USA, 8 years: 2001-2010
- 4km Georgia, 5 years: 2007-2012

Source Impacts (CMAQ-DDM)

- 36km Continental USA, 3 years: 2005-2007
- 20 source categories

Un-bias Numerical Results for Health Assessments

CMAQ Concentrations Adjusted to Observations

- Species Concentrations
 CMAQ Data Fusion Model (Friberg et al., 2016 & 2017)
- Source Apportionment CMAQ-DDM Hybrid (Hu et al., 2014; Ivey et al., 2015, 2016, & 2017)

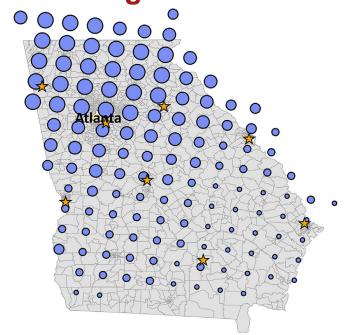


Georgia Hybrid CTM-RM Sources

Annual Average Dust



Annual Average On-road Diesel



Dust

- 0.30 0.44
- 0.45 0.57
- 0.58 0.71
- 0.72 0.86
- 0.87 1.09

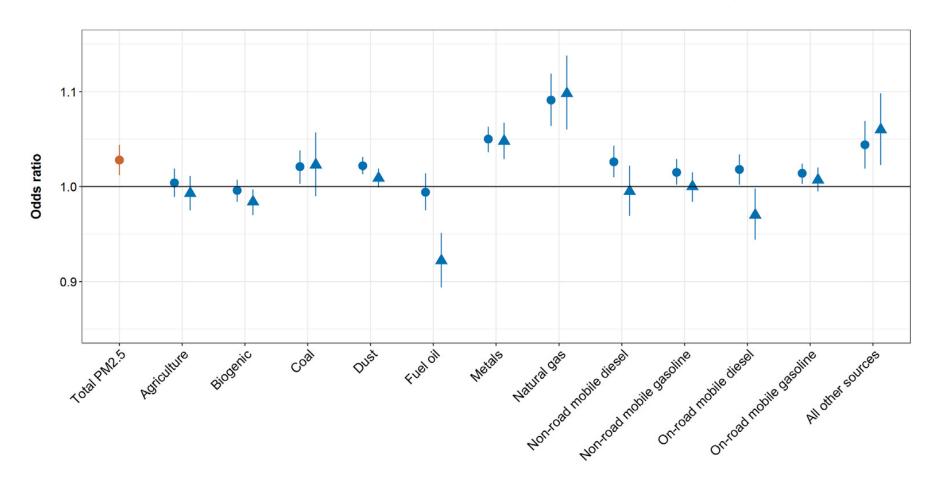
On-road diesel

- 0.44 0.53
- 0.54 0.62
- 0.63 0.73
- 0.74 0.91
- 0.92 1.23



Georgia Pediatric Asthma by Source

ZIP code ED visits for 0-18 years of age





CTMs for Non-tailpipe Exposures

Strengths

- Transport due to changing wind and meteorology conditions
- Secondary pollutant formation
- Model exposure for pollutant mixtures
- Increased number of source categories (e.g., 20+ vs. ~6)
- Improved source apportionment for secondary species

Limitations

- Near roadway gradients need smaller grid cells (~100m)
- Turbulence at small scales difficult to simulate in CTM
- Resolution depends on modeled emissions
- CPU intensive, especially CMAQ-DDM



References

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