

Development of Two Model Fusion Techniques Utilizing CMAQ and RLINE to Obtain PM_{2.5}, CO, and NO_x Concentrations at 250m Resolution over Atlanta, GA



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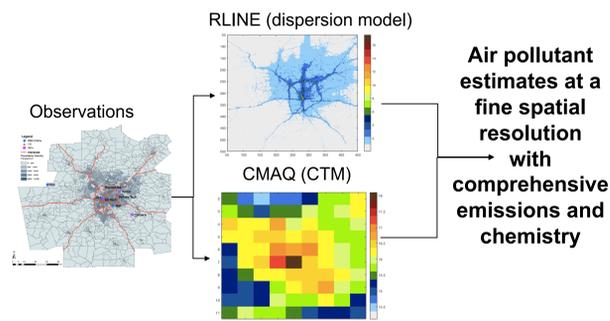


Introduction

Birth cohort study with individual-level residence data needs air pollutant concentration fields at a fine spatial resolution with minimal temporal and spatial bias. Inaccurately capturing intraurban variability in air pollutant concentrations can affect risk ratio estimates.

- Air quality measurements may not capture full spatial variability due to lack of monitoring stations
- Dispersion models simulate small-scale variations but do not simulate chemistry or regional emissions
- Chemical transport models (CTM) can simulate chemistry and local and regional emissions but usually at a coarse resolution

Objective: Develop and apply model fusion approaches that combine observation-fused CTM and dispersion model outputs to obtain fine particulate matter (PM_{2.5}), carbon monoxide (CO) and nitrogen oxides (NO+NO₂=NO_x) estimates at a 250m resolution that retain chemistry and comprehensive emissions



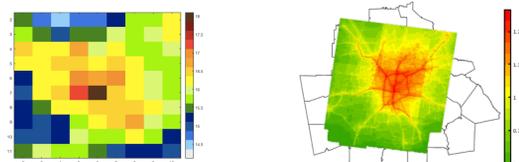
Model Inputs

Chemical Transport Model: Data-Fused CMAQ (OBS-CMAQ)

- Daily values for 2003-2008
- Fused with observations and evaluated in Friberg et al. (2016)

Dispersion Model: Observation-Calibrated RLINE (OBS-RLINE)

- RLINE is a line-source model for near-surface releases of primary inert pollutants chosen to model concentrations near roadways
- Annual averages for 2003-2008
- Ran and calibrated with observations and evaluated in Zhai et al. (2016)

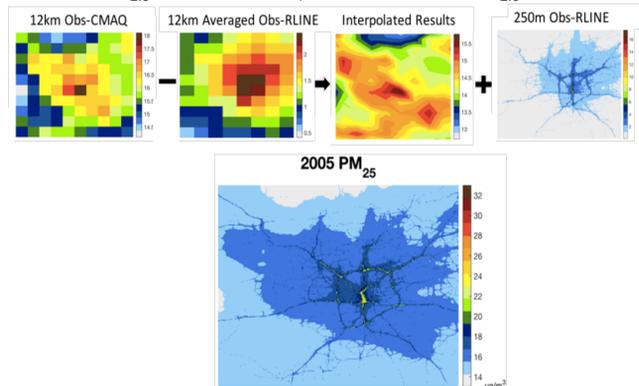


2005 Annual Averages for 12km CMAQ (left) and 250m RLINE (right)

Additive Approach (PM_{2.5})

$$PM_{250m} = [(CMAQ_{12km} - \overline{RLINE}_{12km})_{interpolated}] + RLINE_{250m}$$

- Spatially average RLINE values to match grid of CMAQ (12 km)
- Subtract 12km averaged RLINE concentrations from 12km CMAQ values to remove mobile impacts on PM_{2.5}, resulting in urban background estimates (i.e. particulate matter resulting from all secondary formation and primary sources except mobile emissions)
- Spatially interpolate urban background using triangulation-based linear interpolation algorithm to obtain spatially smooth estimates at 250m grids
- Add RLINE PM_{2.5} to results of step 3 to add mobile PM_{2.5} back into model

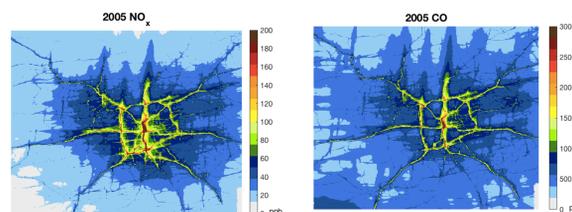


Overall, concentrations of primary roadway PM_{2.5} are placed in their respective locations inside CMAQ grids after removing average roadway primary PM_{2.5} from the CMAQ estimates to avoid double counting.

Multiplicative Approach (NO_x & CO)

The additive approach can lead to unphysical negative results if RLINE_{12km} is higher than CMAQ_{12km}. To avoid this phenomenon, a multiplicative approach that scales RLINE by CMAQ using a linear adjustment factor was developed.

$$Gas_{250m} = \left[\left(\frac{CMAQ_{12km}}{\overline{RLINE}_{12km}} \right)_{interpolated} \right] * RLINE_{250m}$$



Performance & Evaluation

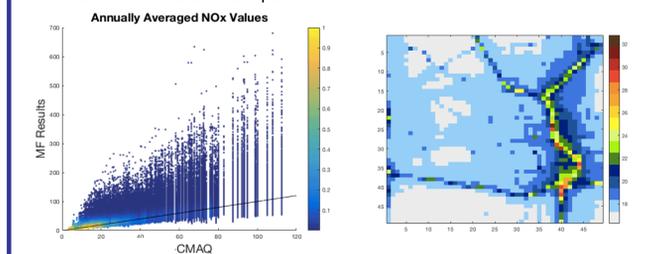
Spatial R values; reported is median over years 2003-2008 with minimum and maximum

SPECIES	MODEL FUSION	OBS-CMAQ	OBS-RLINE
PM _{2.5}	0.60 (0.34–0.77)	0.61 (0.37–0.80)	0.38 (0.16–0.62)
CO	0.98 (0.97–1.00)	0.84 (0.78–0.91)	0.96 (0.93–1.00)
NO _x	0.84 (0.76–0.89)	0.78 (0.72–0.83)	0.74 (0.68–0.77)

Evaluation statistics; median over years 2003-2008 with minimum and maximum

	24-hr PM _{2.5}	1-hr max CO	1-hr max NO _x
Normalized Mean Error (%)			
Model Fusion	9.9 (9.0–12.1)	23.8 (15.6–25.7)	39.6 (35.3–55.3)
Model Fusion Withholding	27.3 (26.5–29.1)	40.4 (36.4–45.0)	61.0 (59.5–74.7)
Normalized Mean Bias (%)			
Model Fusion	6.9 (1.5–8.0)	0.2 (-1.7–8.4)	4.3 (0.9–22.2)
Model Fusion Withholding	7.8 (4.4–11.1)	1.1 (-2.2–10.1)	8.3 (4.4–26.5)
Temporal R			
Model Fusion	0.99 (0.92–0.99)	0.93 (0.92–0.95)	0.98 (0.89–1.0)
Model Fusion Withholding	0.77 (0.73–0.79)	0.54 (0.52–0.59)	0.60 (0.58–0.62)

Model fusion withholding represents a 100% withholding test, i.e. no observations were fused with the CMAQ inputs.



Model Fusion versus CMAQ (left) and model fusion results within one 12-km grid (right) showing the distribution of results within one 12km grid that these approaches capture

Conclusions

- Model fusion approaches simulate steep spatial gradients within one 12km grid while retaining comprehensive emissions and chemistry, which minimizes spatial and temporal biases
- Model input biases affect model fusion performance; calibrations with observations should be made to inputs *a priori*
- Additive method should be used unless background is very small
- Methods could be applied to other models, locations, and pollutants

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