

Modeling the Air Quality Impacts of Specific Sources

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Background

- Two methods for analyzing the impacts of specific sources:
 - Indirect Emission Reduction (a.k.a. “brute-force” method)
 - Two model simulations are conducted with different levels of emissions. Difference between the results yields the impact of the source.
 - Reliable for large sources such as power plants
 - Direct Decoupled Method (DDM)
 - Direct calculation of impact; more efficient
 - Impact of small sources
 - First-order accurate DDM for linear response; higher-order for non-linear response
 - Incorporated in CMAQ for ozone and PM_{2.5}



Overview

- Summary of 2 recent projects (both based on FAQs):
 - Unal, Hu, Chang, Odman, Russell, “Impacts of aircraft and airport-related emissions on air quality and the importance of comprehensively prescribing the timing, location, mode and type of emissions with application to the **Hartsfield-Jackson** Atlanta International Airport,” *Atmospheric Environment*, vol. **39**, no. 32, pp. 5787–5798, October 2005.
 - Unal, Hakami, Odman, “Adaptive grid modeling for predicting the air quality impacts of biomass burning” (application to prescribed burning at **Fort Benning**)
- Synopsis of ongoing/future work



Hypothesis 1

- There are uncertainties in emission inventories for various sources.
 - Aircrafts / airports
 - Biomass burning
 - ...
- These uncertainties can be reduced by:
 - Incorporating new information
 - Using more accurate data
 - Finding and correcting errors



Hypothesis 2

- There are uncertainties in model results due to inadequate resolution. These can be reduced by:
 - Increasing the models' grid resolution
 - *Adaptive Grid Model*
 - Using better resolved data (e.g., terrain, land-use)
 - Improving physical parameterizations (e.g., urban canopy)



Impacts of Aircraft/Airport Emissions



Objectives

- Determine the impact of aircraft emissions on regional air quality for $PM_{2.5}$ and O_3
 - Hartsfield-Jackson International Airport
 - Annual LTO > 420,000
- Improve emissions estimation and processing for aircrafts
 - $PM_{2.5}$ estimation
 - Temporal and spatial distribution



PM_{2.5} Emissions from Aircraft

- A “First Order Approximation” was introduced by FAA (Wayson and Fleming, 2001)

$$E(PM_{2.5}) = 0.6 \times (SN)^{1.8} \times (FF)$$

- **E**: Mass of PM_{2.5} emissions
 - **SN**: Smoke Number
 - **FF**: Fuel Flow rate
- NASA is conducting a more detailed study



PM_{2.5} Emission Estimates at Hartsfield-Jackson

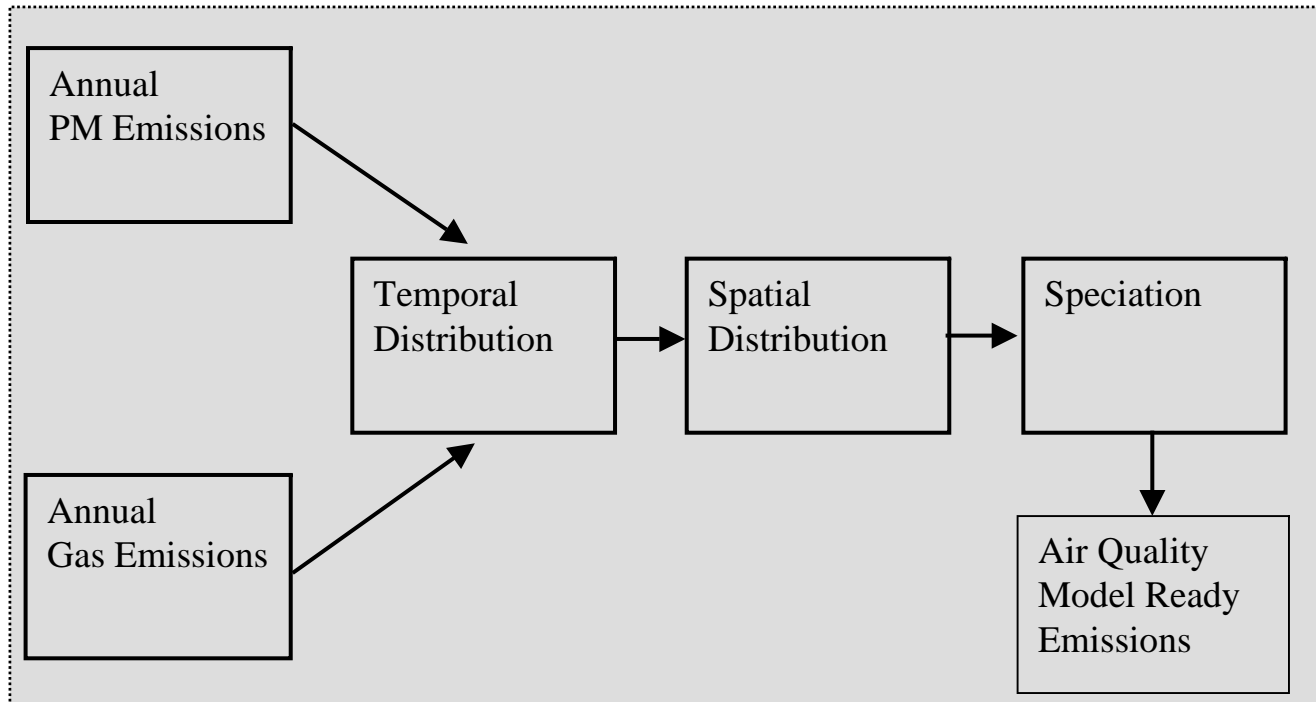
- Characteristic Value:
 - Based on International Civil Aviation Organization (ICAO) database
 - 70 tons/year
- Mode Specific:
 - We used statistical relations b/w take-off and other modes

$$SN_{climb-out} = 0.86 \times SN_{take-off}$$

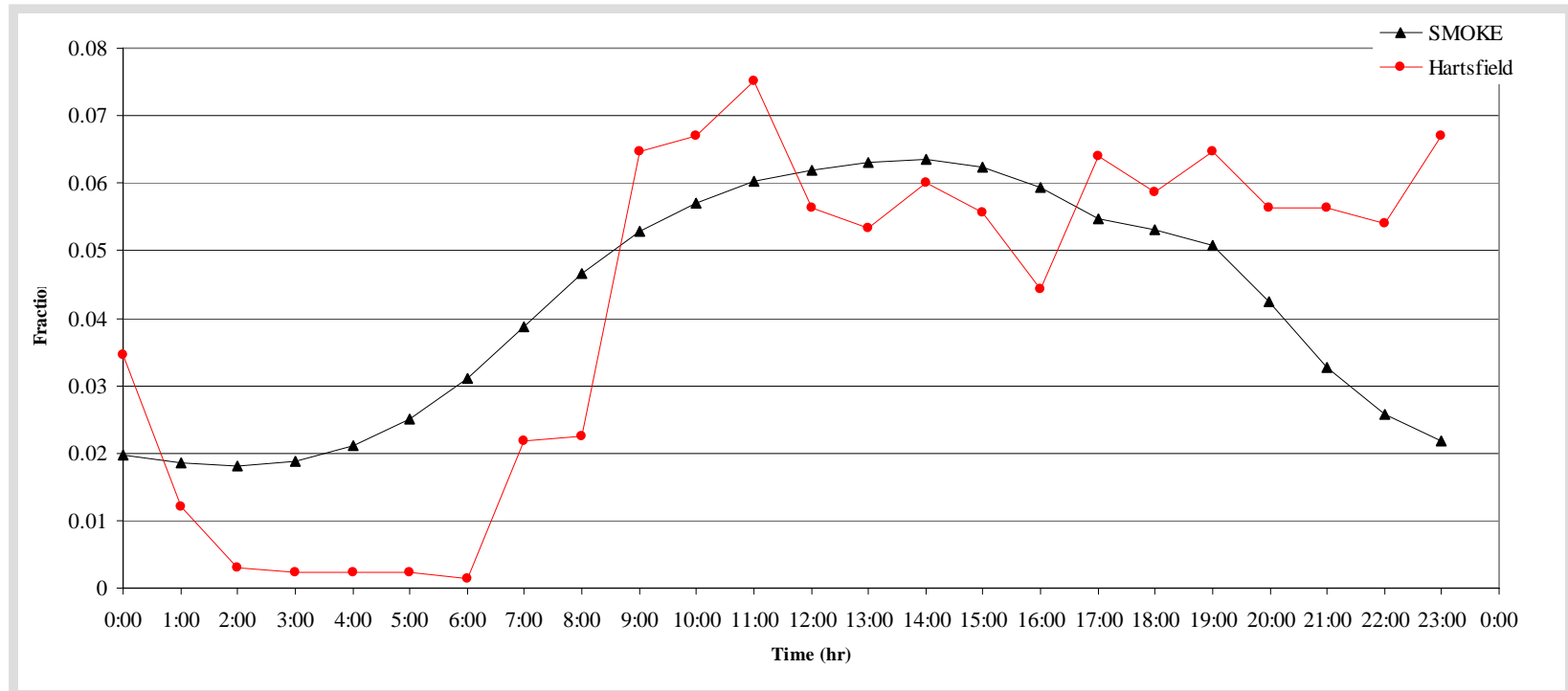
- 27 tons/year



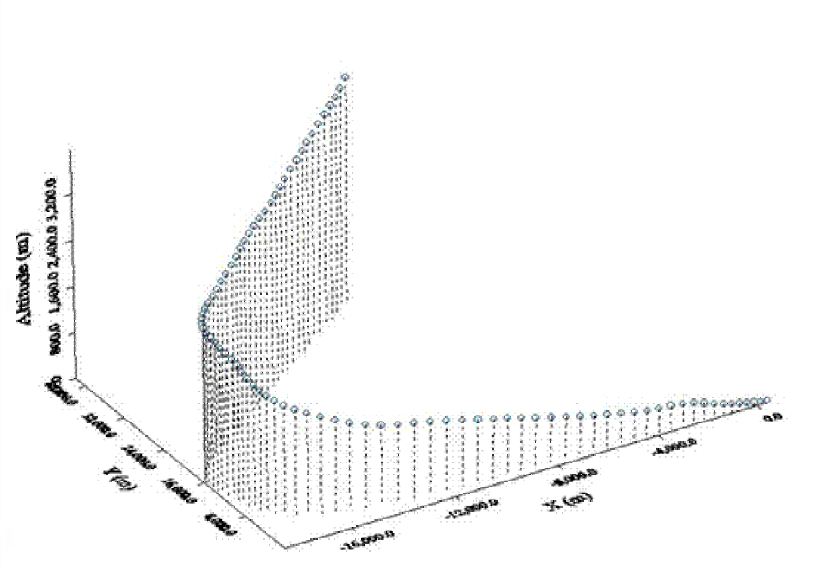
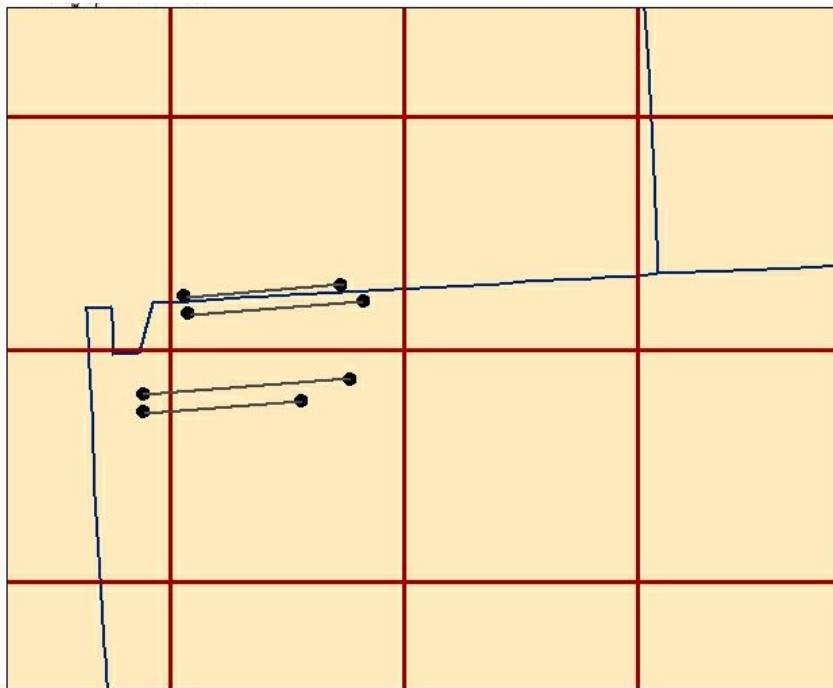
Emissions Processing



Temporal Considerations



Spatial Considerations

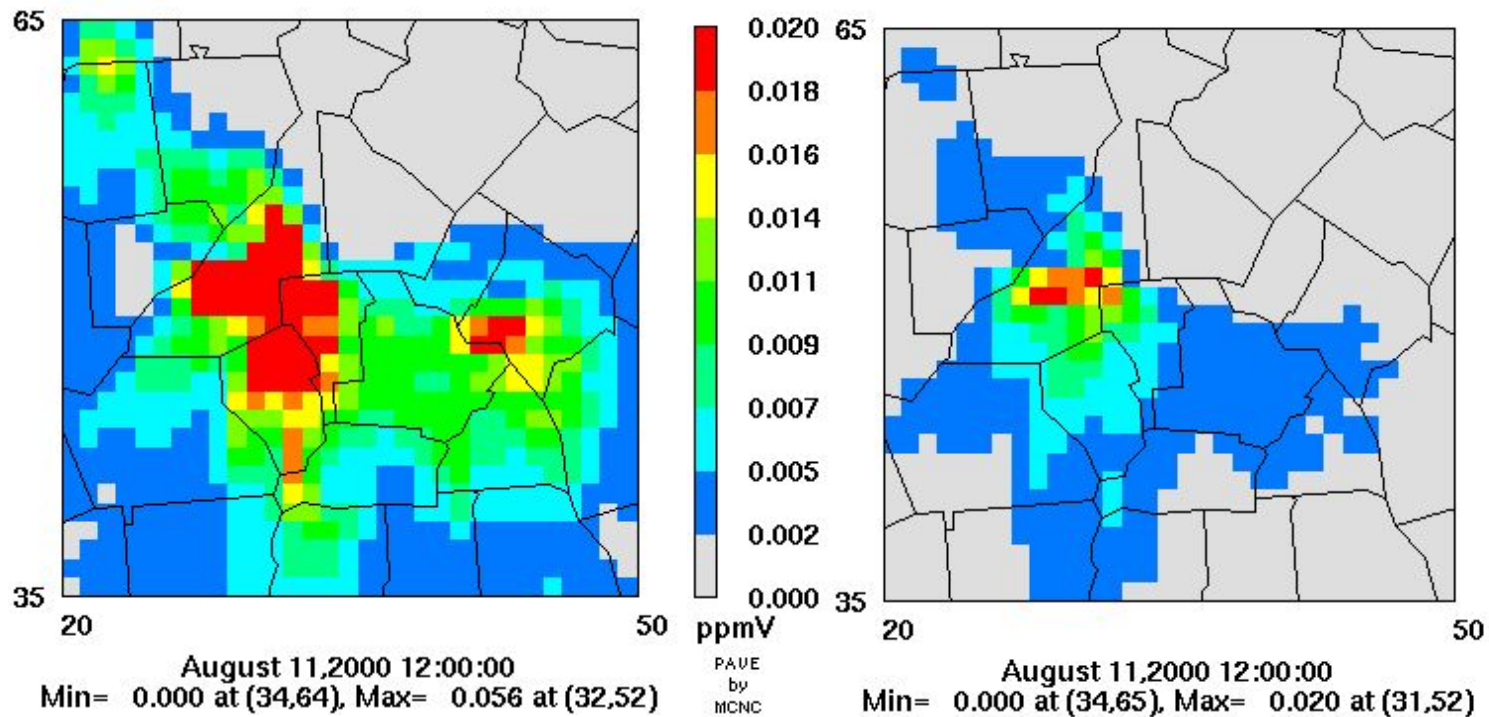


Chemical Considerations

Composition of PM _{2.5} emissions (%)			
EC	OC	SO ₄	NO ₃
65.87	29.21	4.60	0.32



Aircraft Impact - Ozone



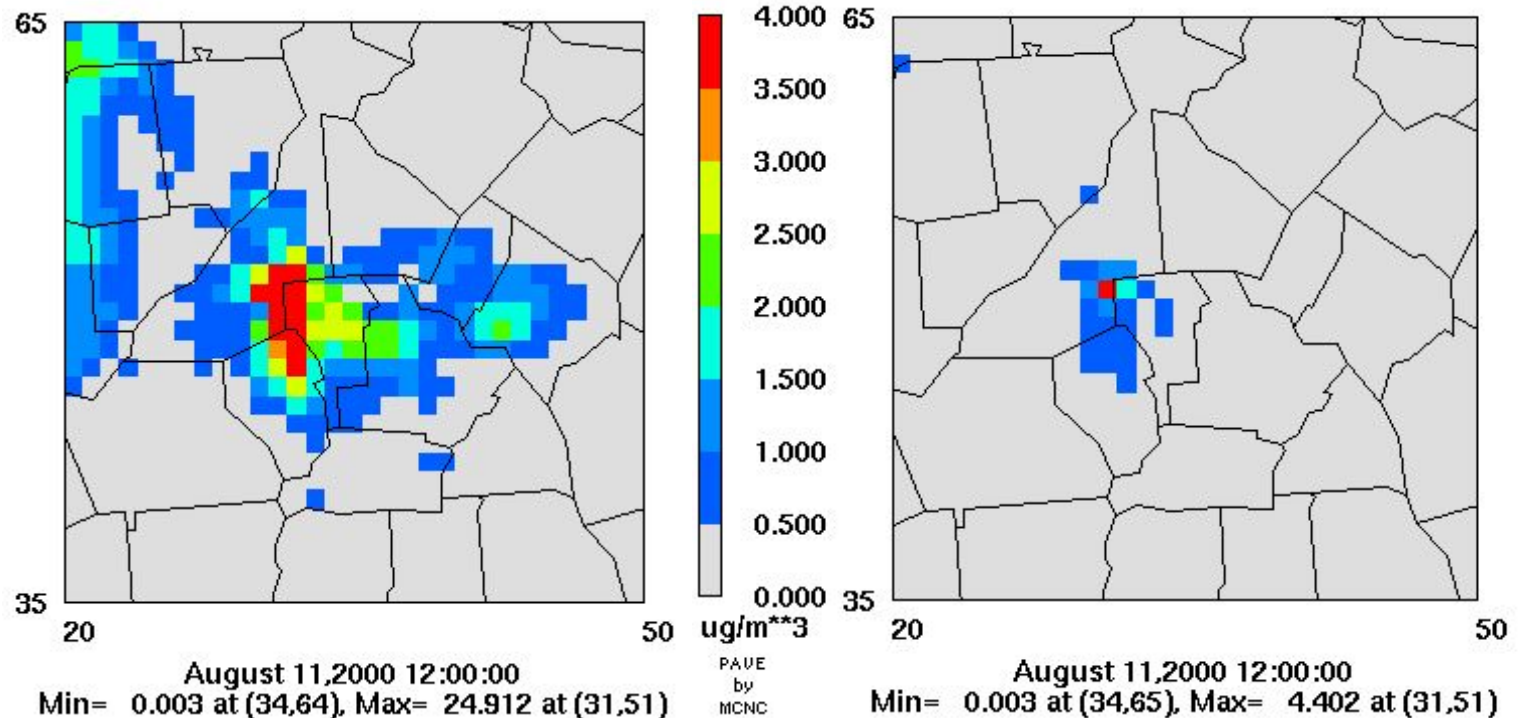
Characteristic Value
& 1st layer

Mode Specific
& 3-D

Maximum sensitivity of regional concentrations to aircraft emissions



Aircraft Impact - PM_{2.5}



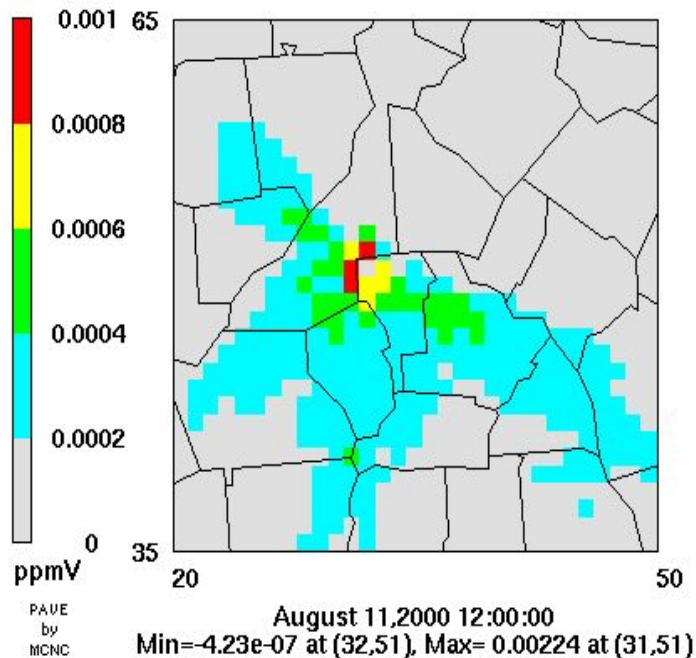
Characteristic Value
& 1st layer

Mode Specific
& 3-D

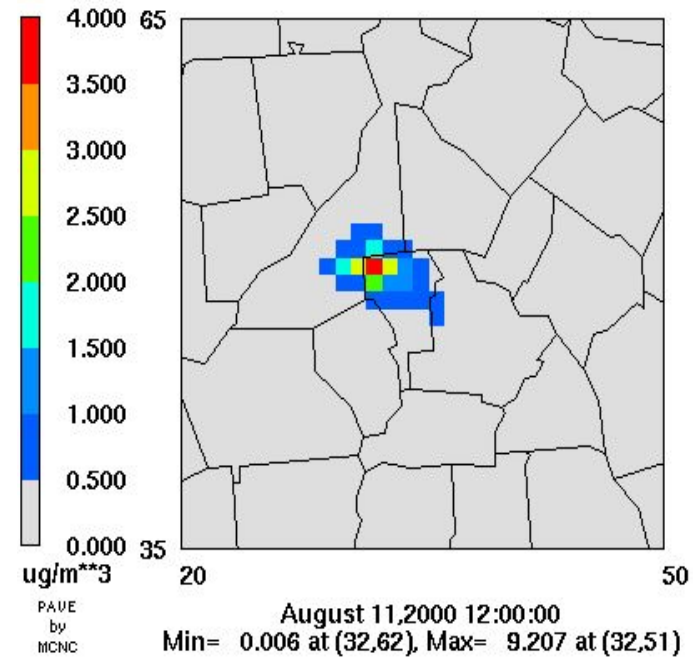
Maximum sensitivity of regional concentrations to aircraft emissions



Ground Support Equipment Impact



Ozone



PM_{2.5}

Maximum sensitivity of regional concentrations to Ground Support Equipment emissions



Findings

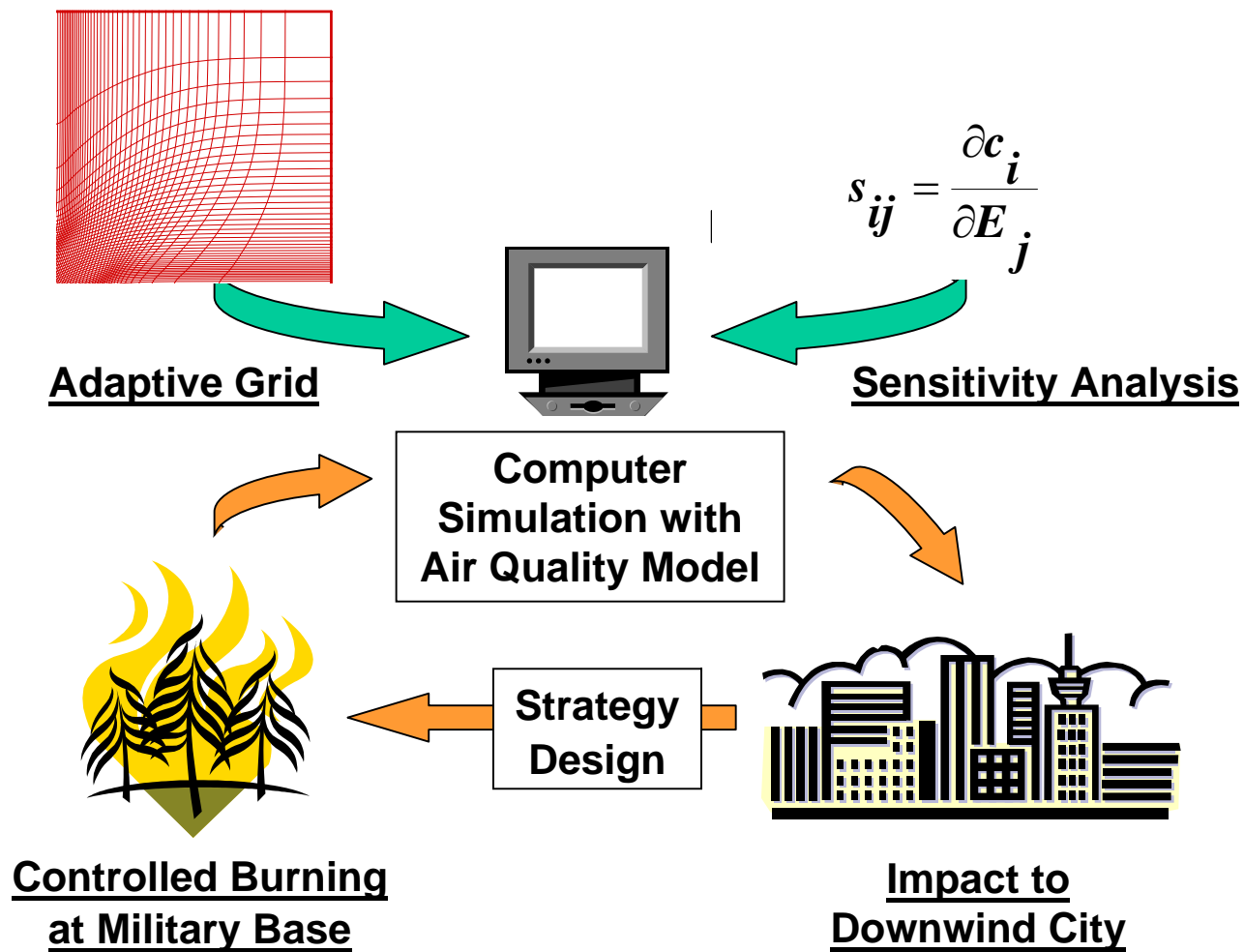
- Large uncertainties in aircraft PM_{2.5} emission estimates
- Maximum aircraft impacts for the August 11-20, 2000 period are estimated as:
 - Ozone:
 - 56 ppb (Characteristic Value)
 - 20 ppb (Mode Specific and Spatial Distribution)
 - PM_{2.5}:
 - 25 µg/m³ (Characteristic Value)
 - 4 µg/m³ (Mode Specific and Spatial Distribution)
- Ground support equipment impacts are smaller and more local



Impacts of Biomass Burning



Objectives



Biomass Burning Emissions

- First Order Fire Effects Model (FOFEM) Version 5, was utilized.
 - Fuel type was assumed as natural fuel with long leaf pine
- For speciated VOCs and NO_x, emission factors provided by Battye and Battye (2002) were used.

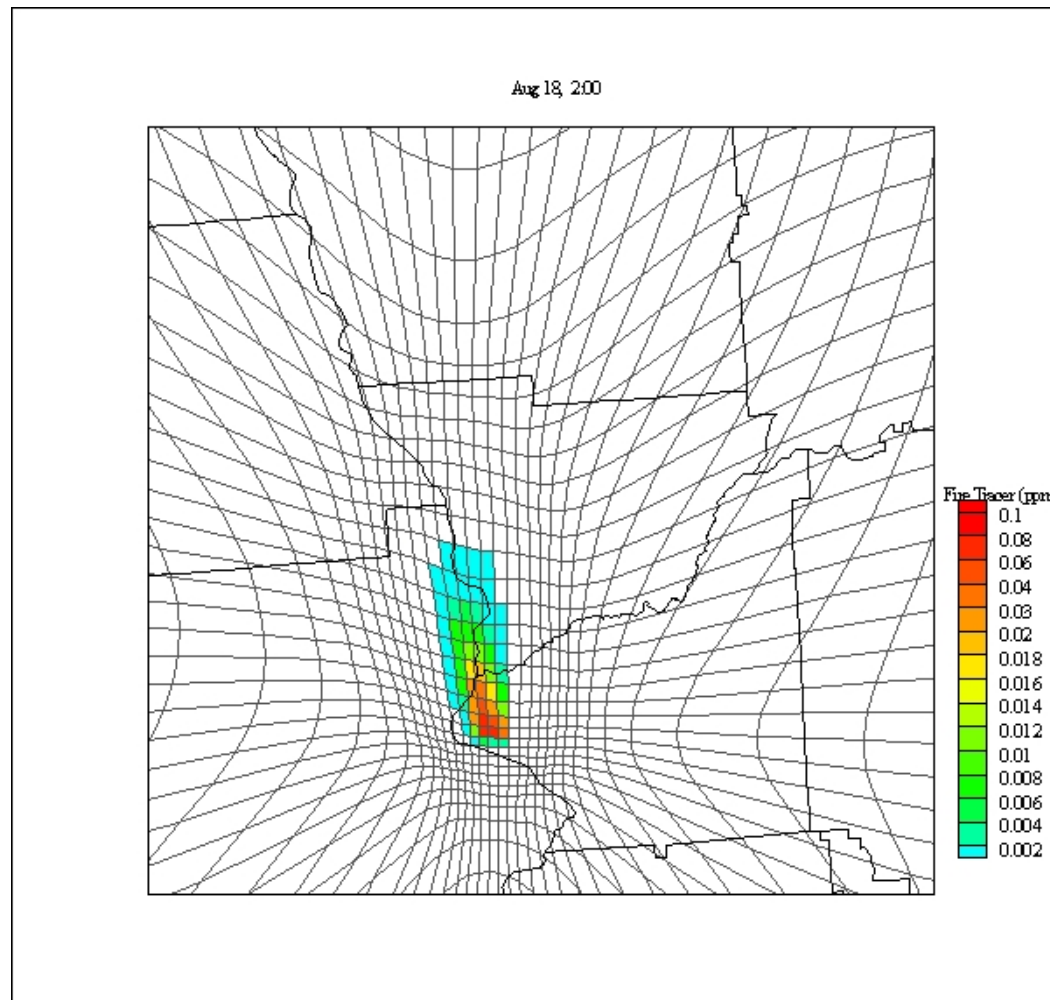


Adaptive Grid Model

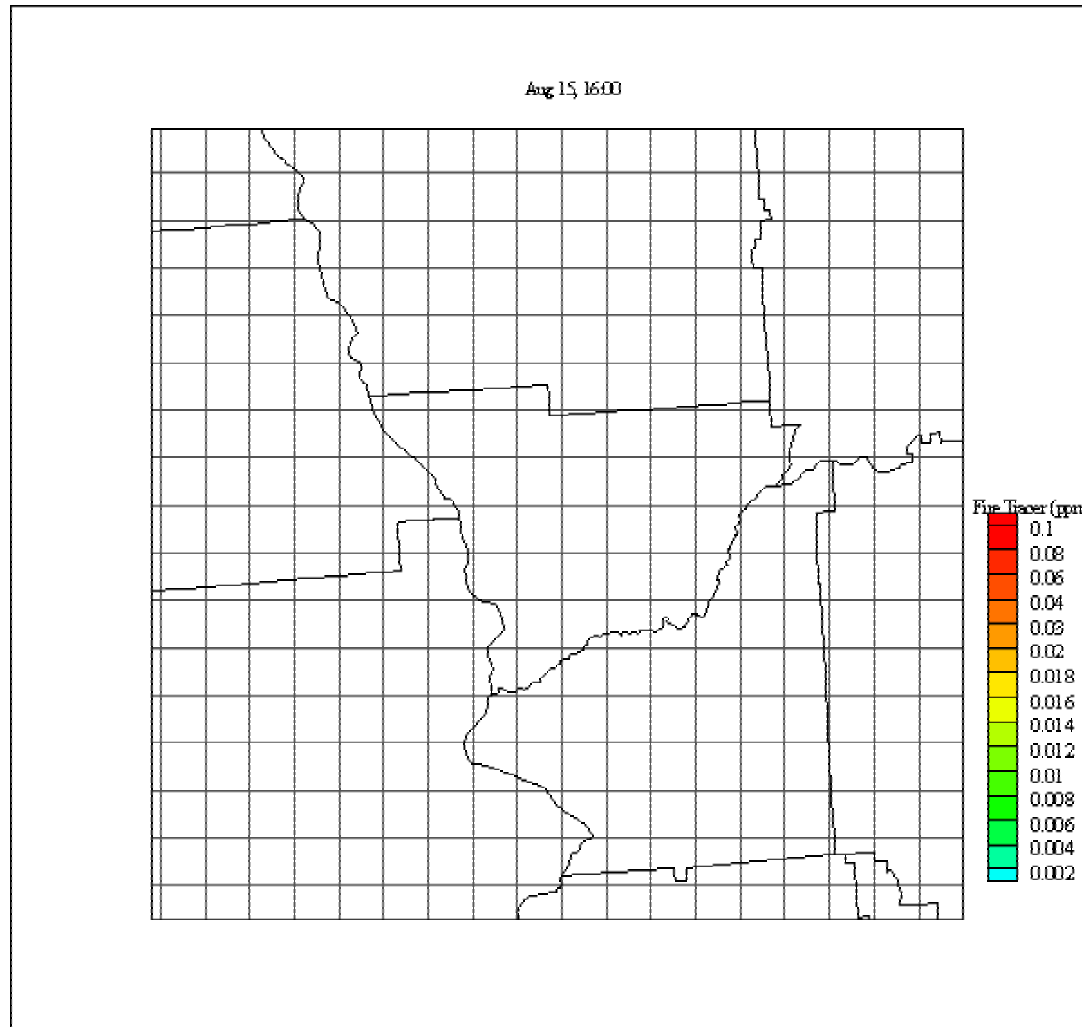
- Our adaptive grid technique is a *mesh refinement* algorithm where the number of grid cells remains constant and the structure (topology) of the grid is preserved.
- Grid nodes move continuously during the simulation. Grid cells are automatically refined/coarsened for a more accurate solution.
- A weight function controls the movement of the grid nodes according to user-defined criteria (change in NO_x gradients here).



Adaptive Grid Model

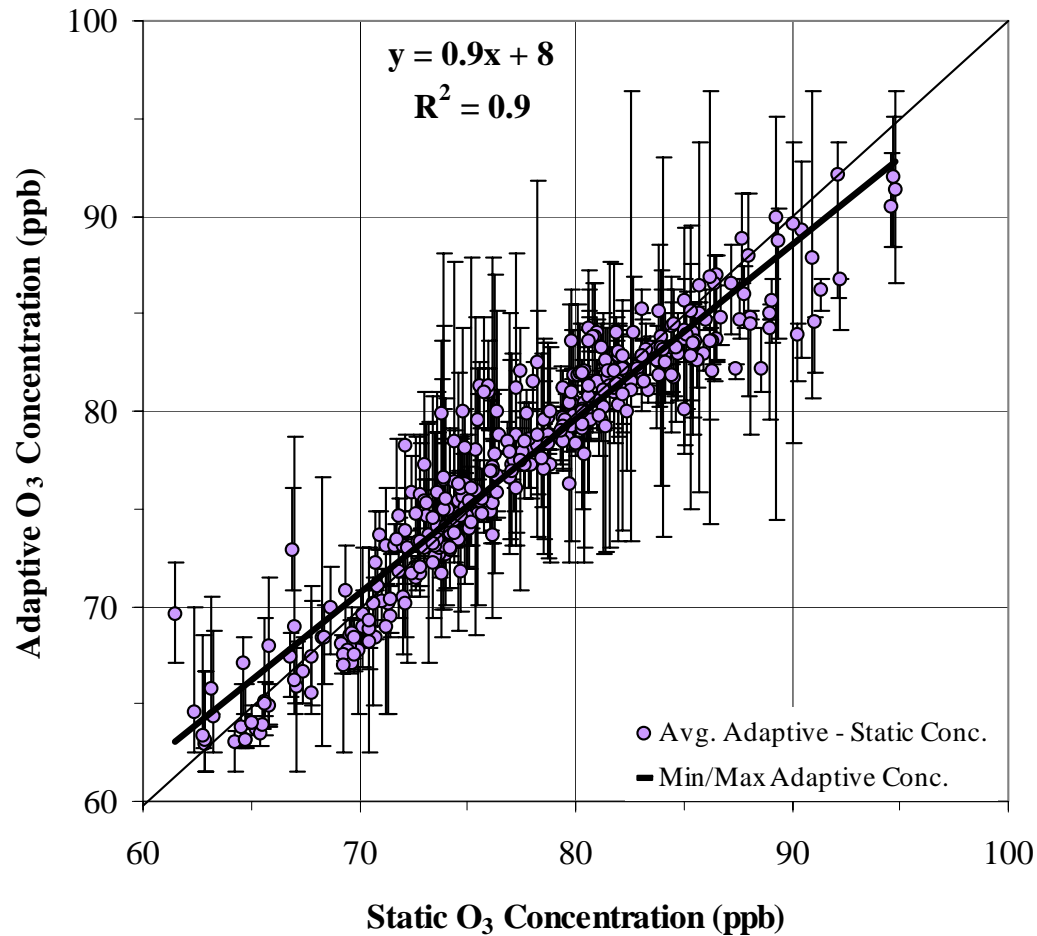


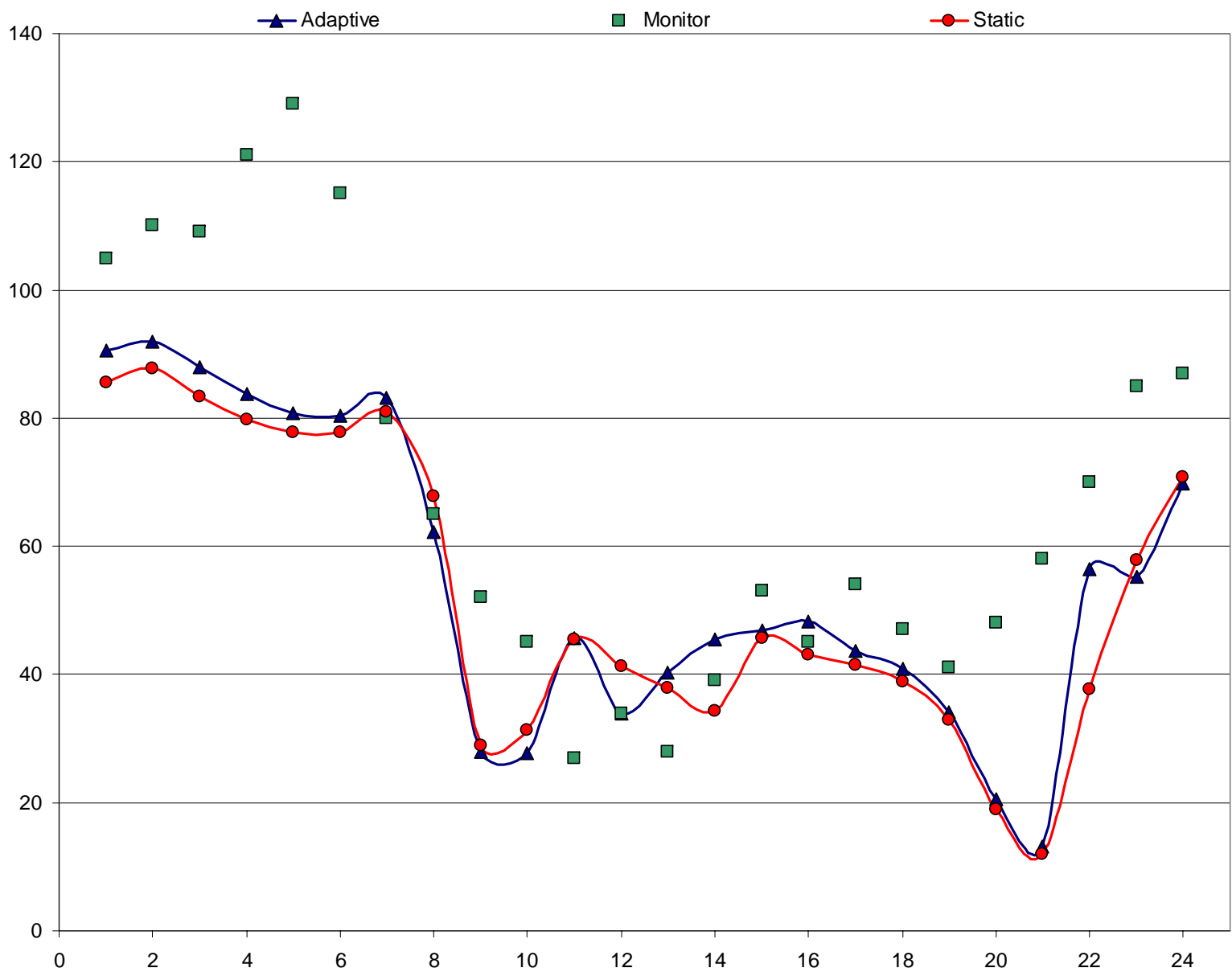
Adaptive Grid Model



O₃ Concentration: Adaptive vs. Static Grid

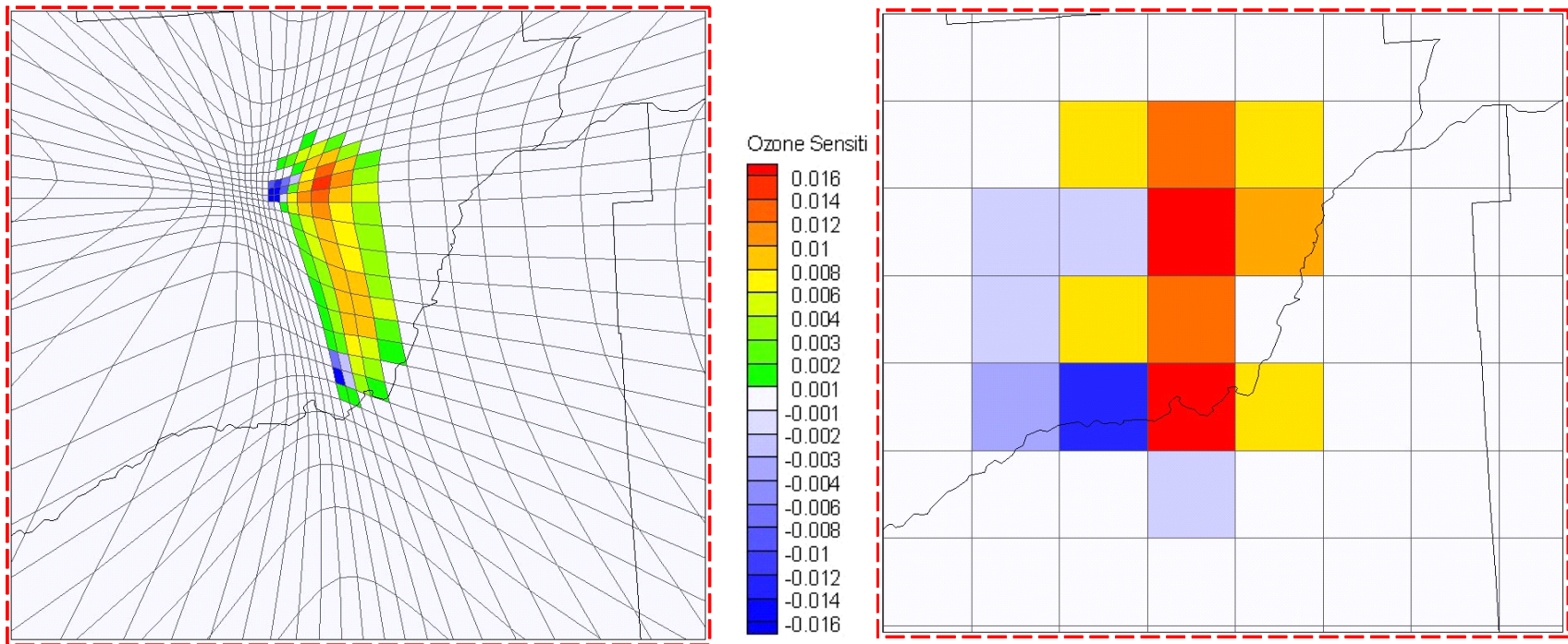
August 15, 21:00 (GMT)





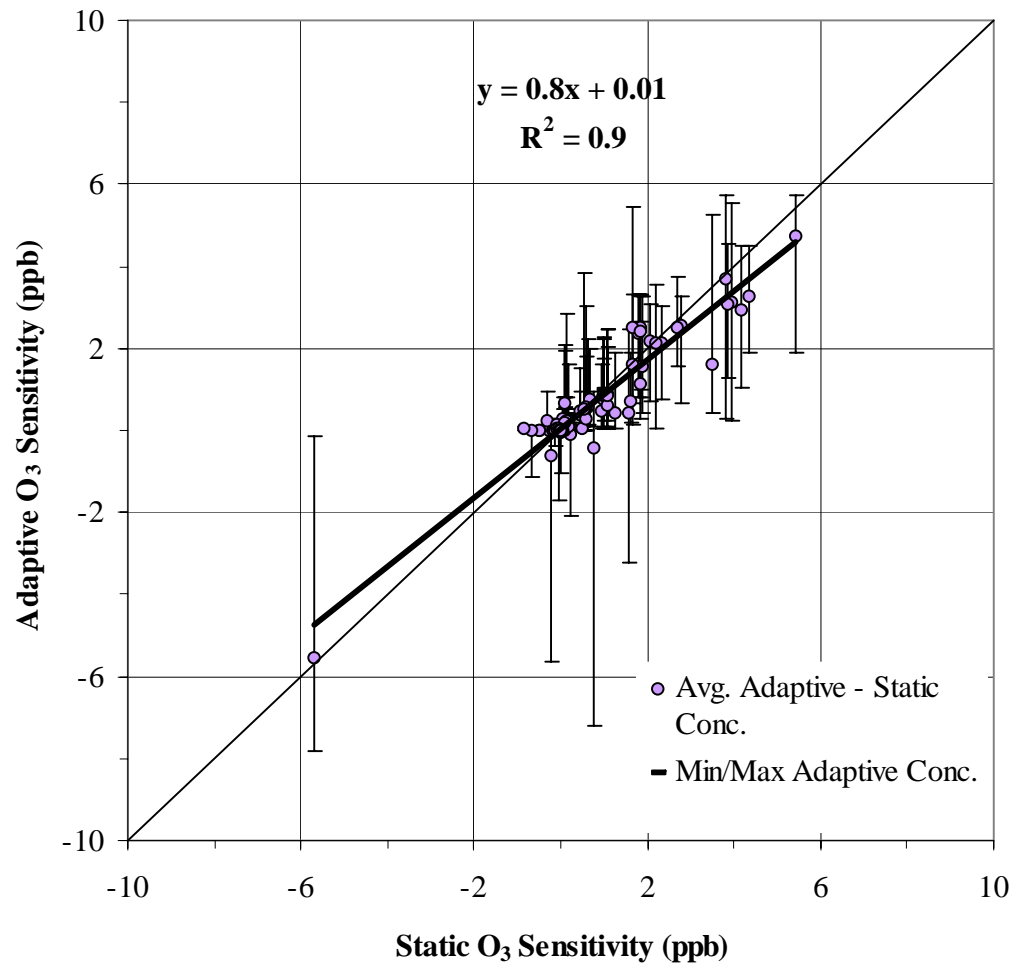
O₃ Sensitivity: Adaptive + DDM vs. Static + BF

August 15, 20:00 GMT



O₃ Sensitivity: Adaptive + DDM vs. Static + BF

August 15, 20:00 GMT



Findings

- Two techniques, Adaptive Grid Modeling and Direct Sensitivity Analysis (DDM), were merged to determine the impact of biomass burning.
- The impact of fires at Fort Benning during the modeled episode (August 15-18, 2000) ranged from 16 ppb reduction to 7 ppb increase in O₃ concentrations.
 - Impacts on Columbus were minimal due to wind directions
- Compared to static grid, concentration gradients were better resolved by Adaptive Grid.
- DDM with adaptive grid and Brute Force with static grid showed different ozone impacts.



Next Steps: Fine-Scale Forecasting

- 2006:
 - PM_{2.5} forecast using “state-of-science” models
 - WRF (meteorology) and CMAQ (air quality)
- 2007 and beyond:
 - Fine-scale air quality forecasts (archived for health effects)
 - More accurate emissions
 - High resolution for sources and receptors
 - Sensitivity forecasts
 - Individualized forecasting products



Acknowledgements

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