Regional Chemical Modeling in Support of ICARTT

Topics:

- ➤ How good were the regional forecasts?
- ➤ What are we learning about the emissions?
- ➤ What are our plans for integrating models with observations?

Our Analysis Framework

MOZART Global Chemical Transport Model

Meteorological
Dependent Emissions
(biogenic, dust, sea salt)

Mesoscale Meteorological Model (RAMS or MM5) Influence Functions
Emission Biases/
Inversion

Anthropogenic & biomass burning Emissions

TOMS O₃

STEM Tracer Model
(classified tracers for regional and emission types)

Airmasses and their age & intensity Analysis

STEM Prediction
Model with on-line
TUV & SCAPE

Chemistry & Transport
Analysis

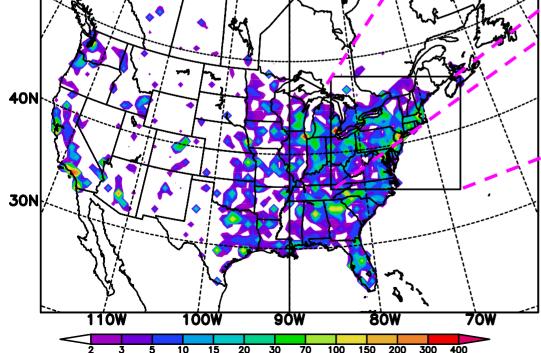
STEM Data-Assimilation Model

Observations

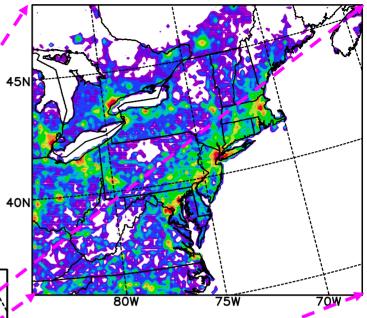
Analysis Done at 60 and 12 km Horizontal Resolution

NEI-1999 emission in 60km (below) and 12km (right) domains.

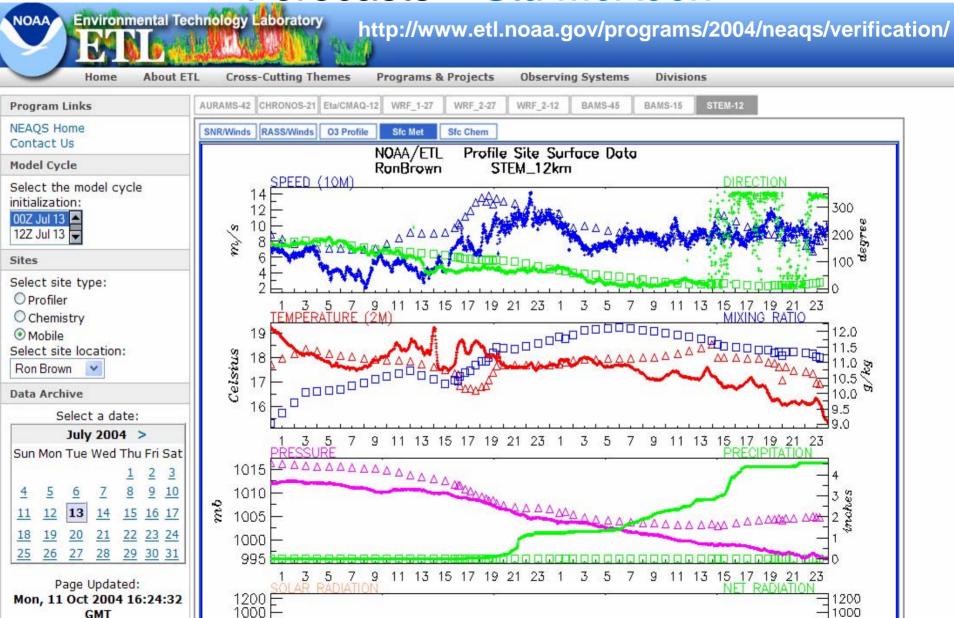
Mean CO Emission for Typical Summer day (10¹¹ Molecules/cm²/s)

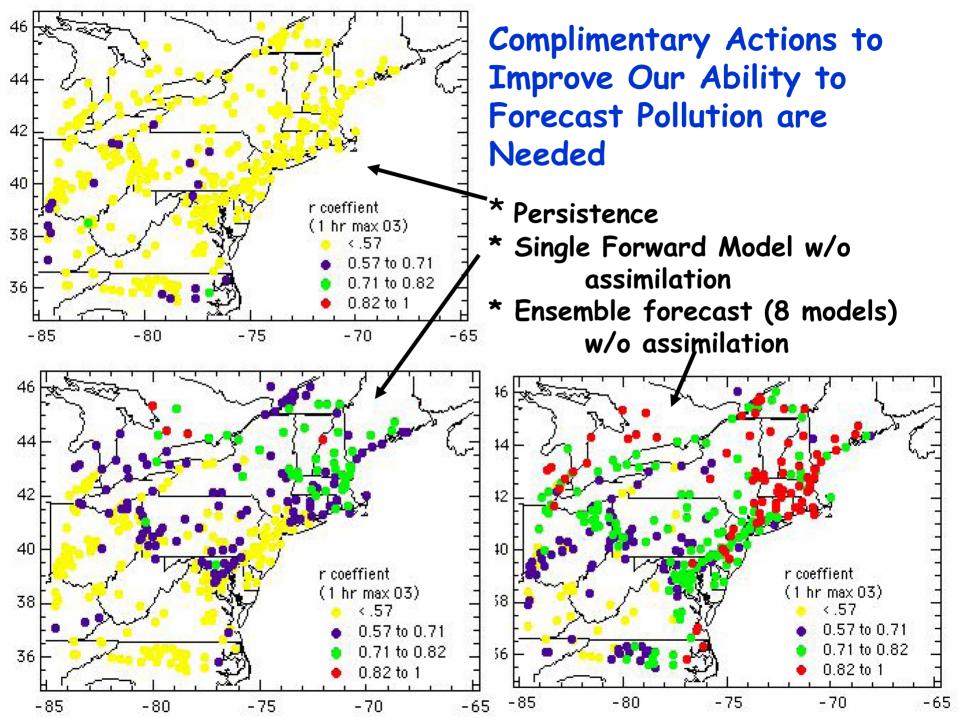


Mean CO Emission for Typical Summer day (10" Molecules/cm²/s)

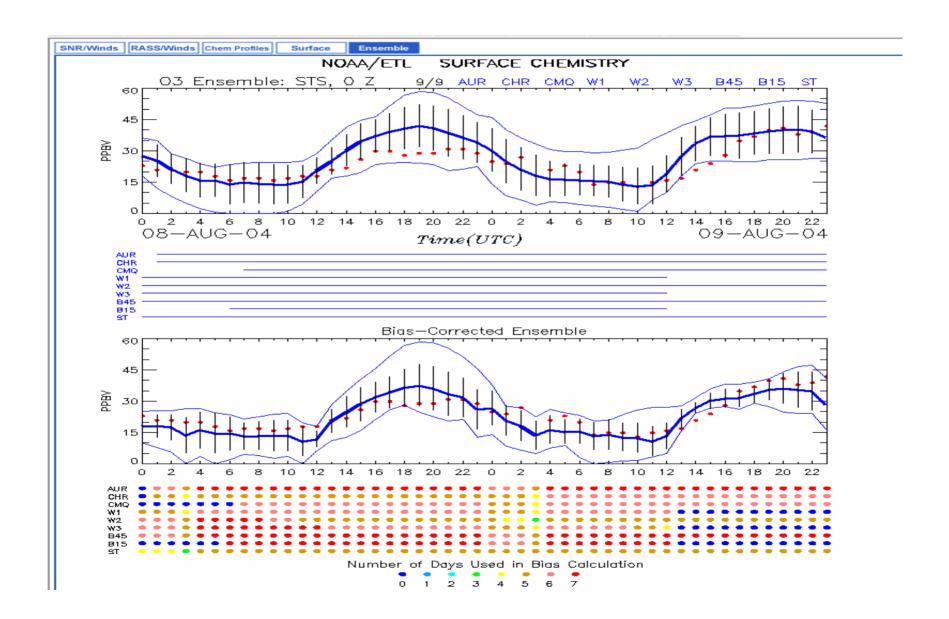


Extensive Real-Time Evaluation of Regional Forecasts – *Stu McKeen*

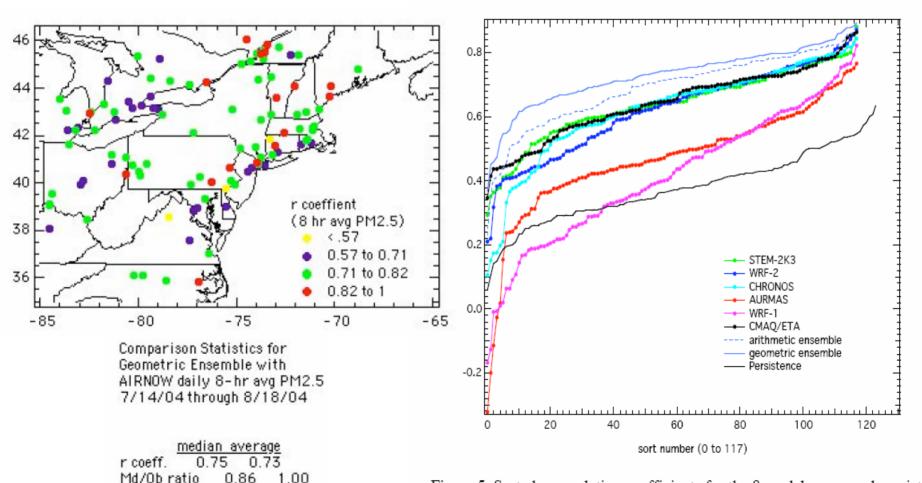




Ensemble Techniques Help!



Ensemble Methods Also Work for PM2.5 Forecasting



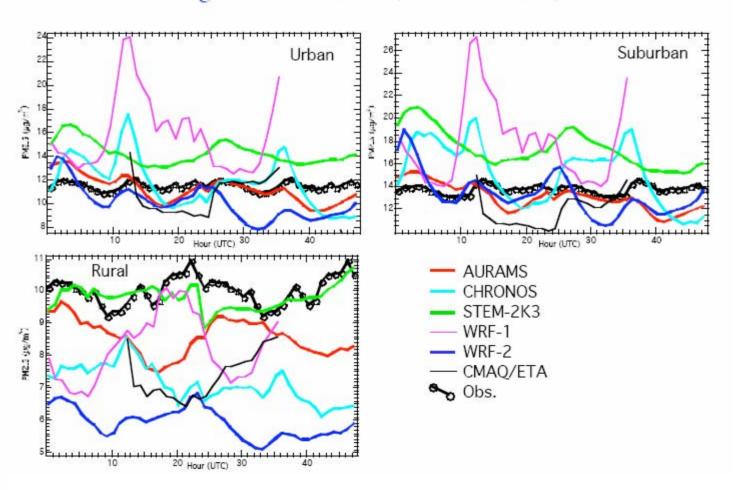
ratio RMSE

Skill factor(%)=

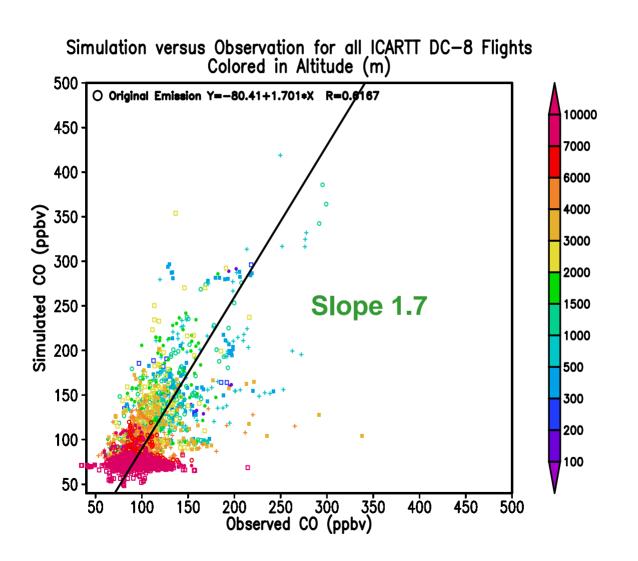
Figure 5. Sorted r-correlation coefficients for the 8 model cases, and persistence

But take little comfort...... We have a long way to go !!

PM2.5 Average Diurnal Profiles, summer 2004, in NE U.S.



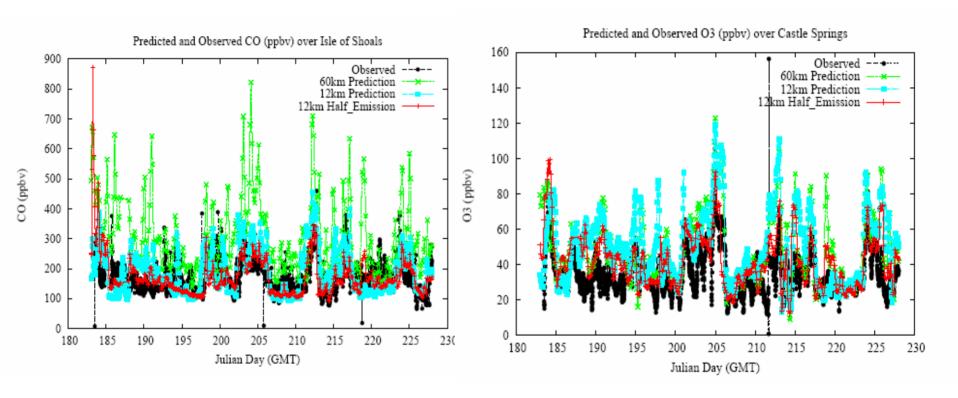
Emission Issues



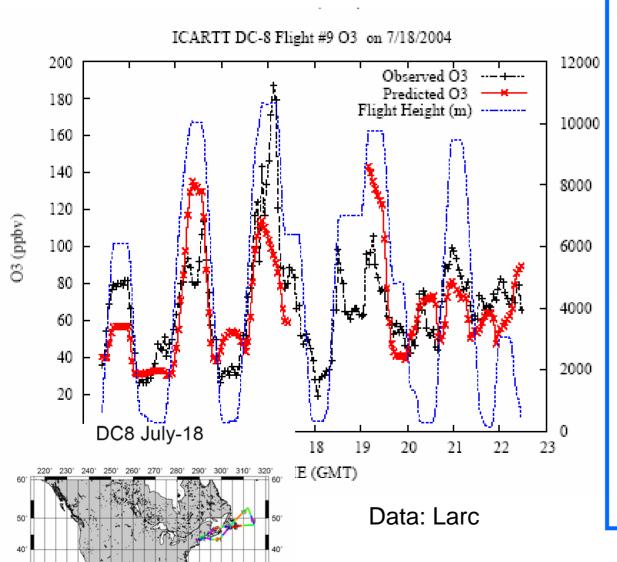
Sensitivity Runs Using Reduced Emissions -- Correlations between STEM simulations and Measurements for All DC-8 flights

Species	60 km simulation with original NEI-1999v3 emission		$60~\rm{km}$ simulation with half CO, $\rm{NO_x}$ and $\rm{SO_2}$ emissions	
	Slope	R	Slope	R
СО	1.70	0.62	0.83	0.66
NO _y	4.11	0.48	1.50	0.48
PILS SO ₄ ²⁻	2.52	0.75	0.78	0.75
SAGA SO ₄ ² -	3.06	0.74	1.13	0.74
PILS NH ₄ +	0.35	0.35	0.33	0.48
SAGA NH ₄ +	1.60	0.64	1.08	0.66
O_3	1.13	0.46	0.97	0.55
Ethyne	0.21	0.50	0.26	0.51
URI HCHO	0.82	0.84	0.89	0.84
H ₂ O ₂	0.56	0.70	0.47	0.67

Clear Improvement in Surface Predictions

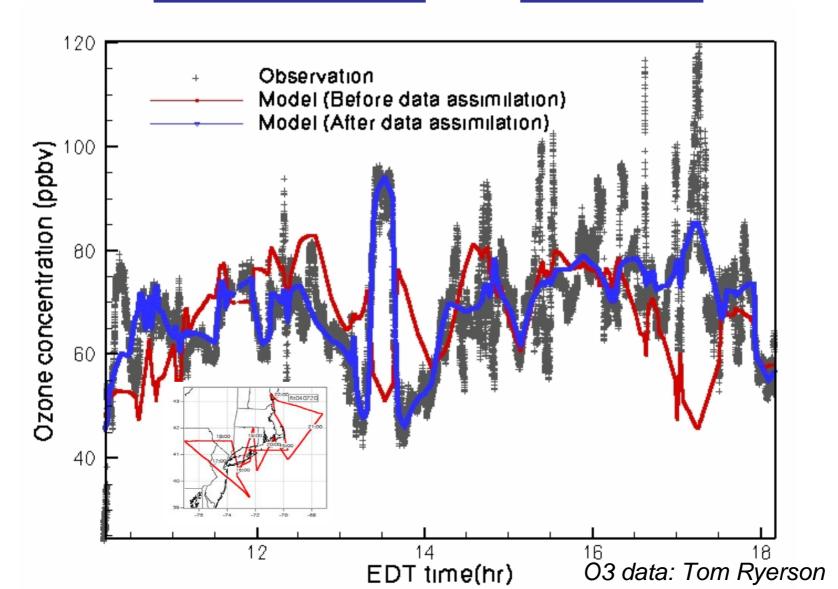


Integration of Measurements & Models



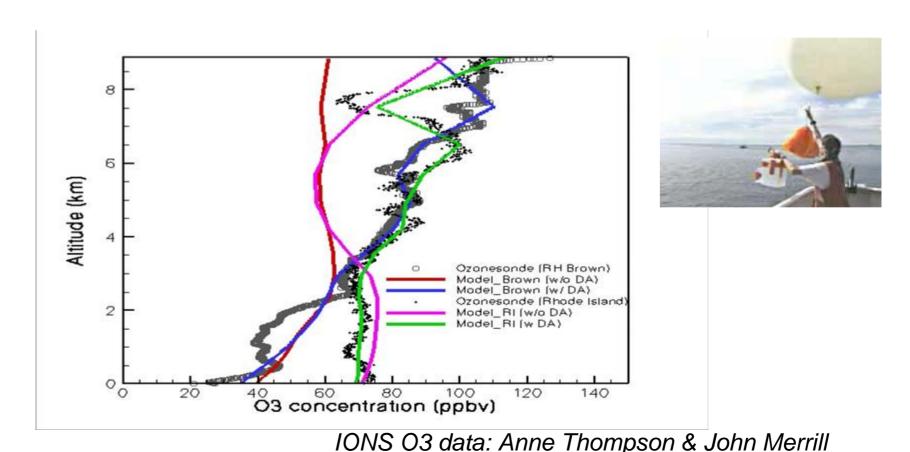
- Cost functional measures the modelobservation gap.
- Goal: produce an optimal state of the atmosphere using:
 - Model information consistent with physics/chemistry represented
 - Measurement information consistent with reality
 - within errors

Reanalysis of Ozone using <u>Surface</u> as Well as <u>Ozone Profile</u> and <u>Aircraft</u> Data

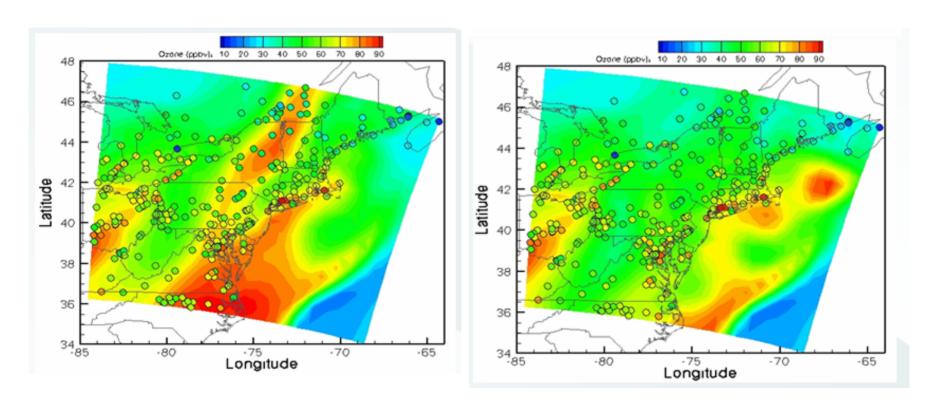


Getting the Vertical Distributions Right is Critical

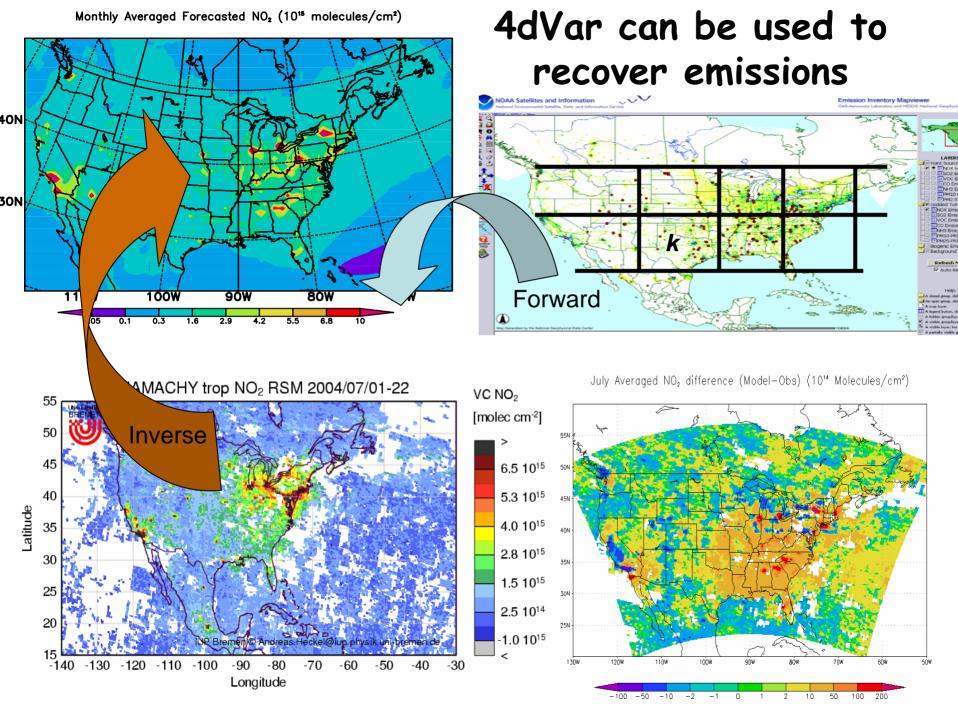
Current models have a difficult time...so data assimilation is important



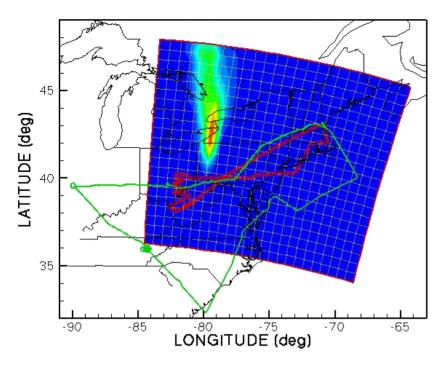
Ozone Forecasts (left) and Reanalysis (Right)

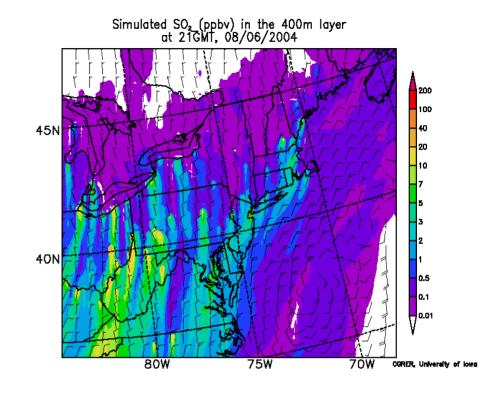


Circles represent observations (locations and values)









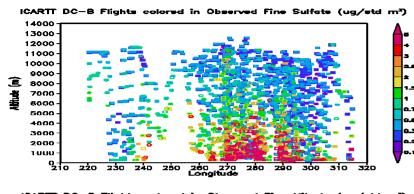
Adjoint Tools Can Also Help in the Characterization of Emissions

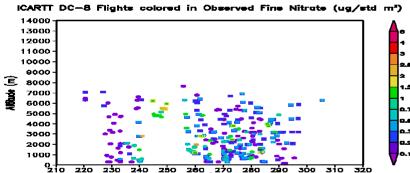
Preliminary Results: CO emission scaling factor ~ 0.7.

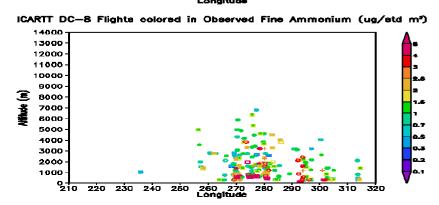
Regional Distributions of Aerosols

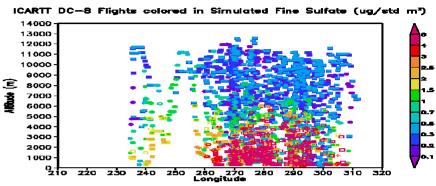
Observed (PILS)

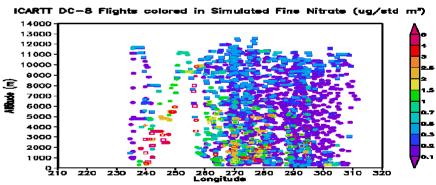


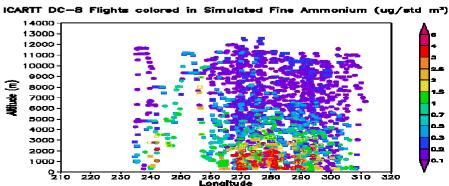






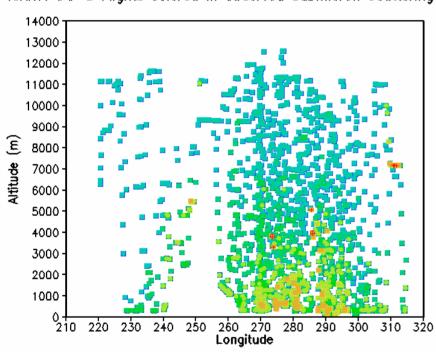


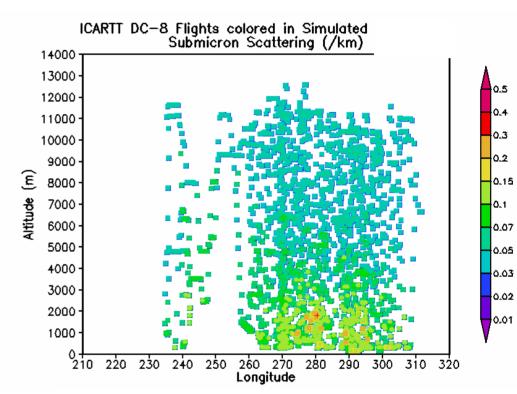




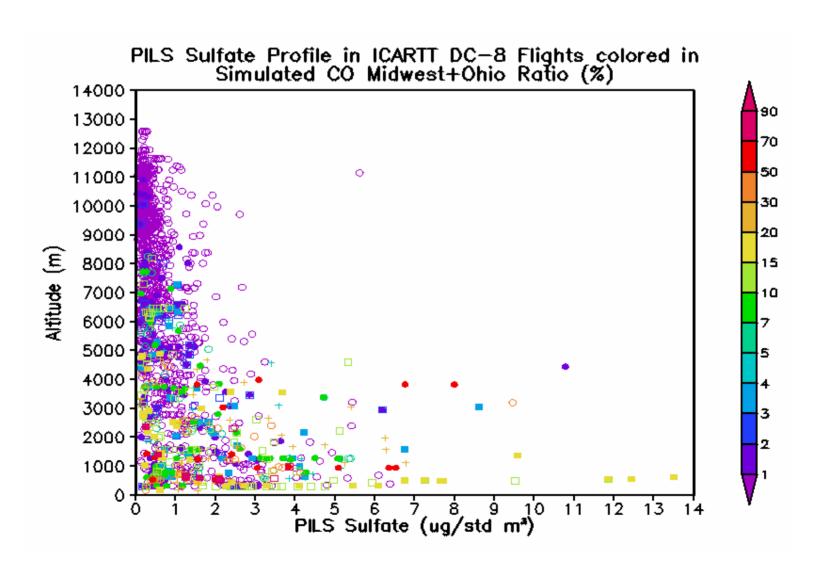
Observed and Predicted Submicron Scattering







STEM Source Region Tracers Can Be Used to Sort Data & Complement Trajectories



Future Plans

- Improve Base Emissions -- Update base year inventory (Streets and Vukovich), Biomass burning (others)
- > Emission inversions
- Re-analysis using aircraft, surface, satellites, sondes (Ozone, CO, NOy, HCHO)
- Analysis of aerosols and optical properties, by better linking observations and models
- Better understand and constrain physical removal processes (dry and wet)

We will submit our model products along the flight tracks