

Assessing Model Performance Using Aircraft Data

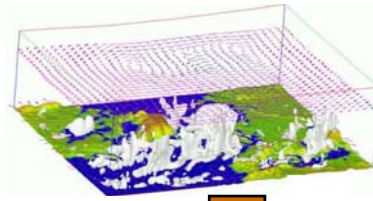
- Overview of the uses of observations by models.
- Using observations to evaluate models.
- Combining observations with models

How are observations used by models

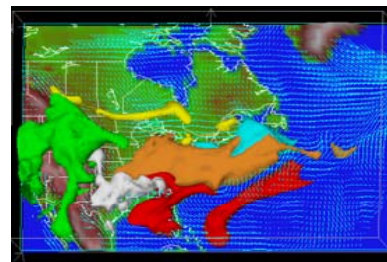
- “Direct” use in models
 - Boundary conditions (in limited area models)
 - Initial conditions (e.g., trajectory fill methods, and observation-based models)
 - As parameters (e.g., size distributions, optical properties)
- Evaluation
 - Point comparisons
 - Profiles
 - Different types of air masses, processes, etc.
- Data assimilation (formally combining observations and models)

Air Quality Modeling: Improving Predictions of Air Quality (analysis and forecasting perspectives)

Met model

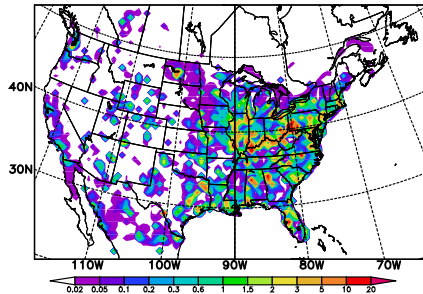


CTM

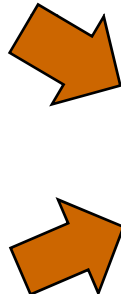
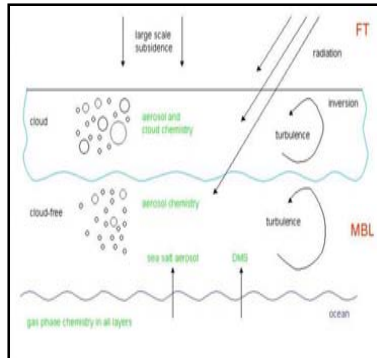


Emissions

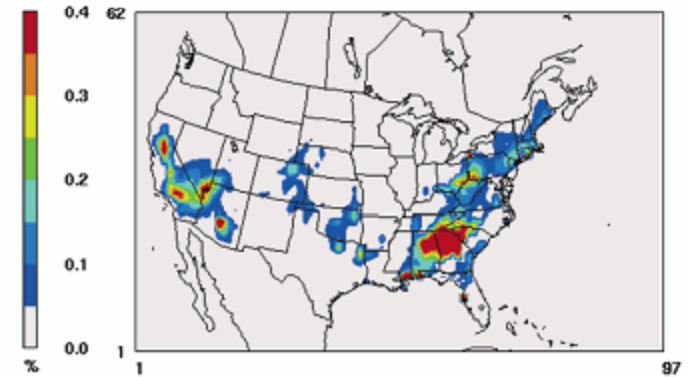
Mean SO2 Emission for Typical Summer day (10⁶ Molecules/cm²/s) in NEI1999



Chemical, Aerosol, Removal modules

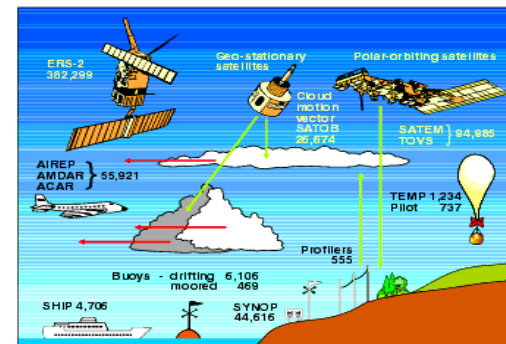


Predicted Quantity: e.g., *ozone AQ violation*



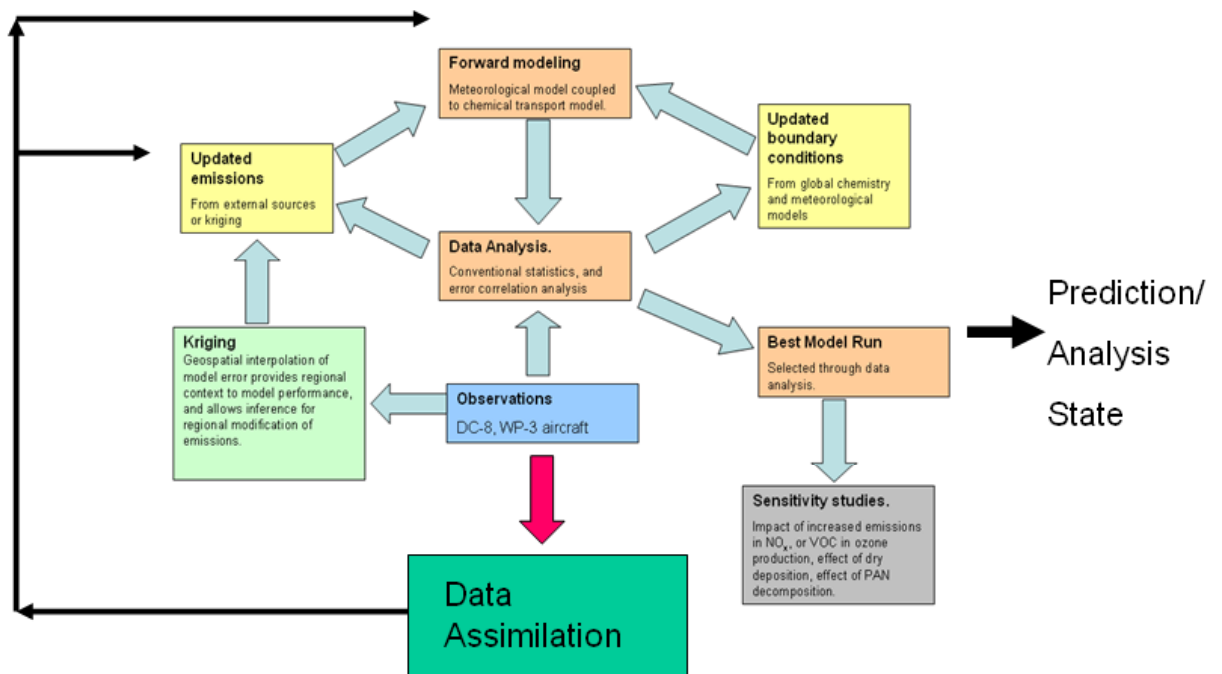
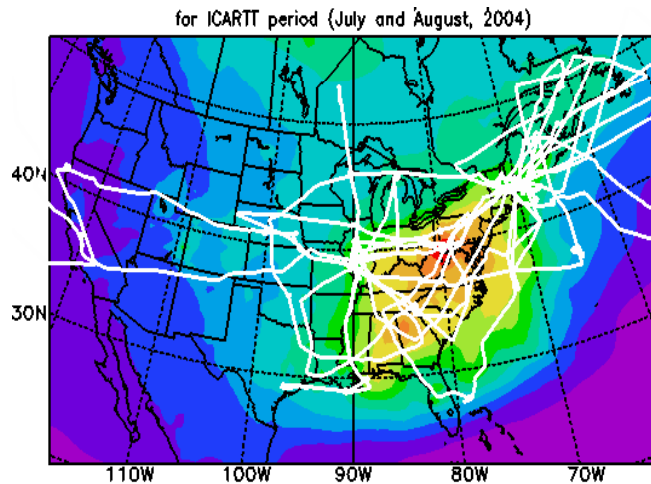
How confident are we in the models & predictions?
What do the observations tell us about the quality of the calculation?

Observations



Experiments such as ICARTT employ mobile “Super-Sites” and Provide a Comprehensive Set of Observations

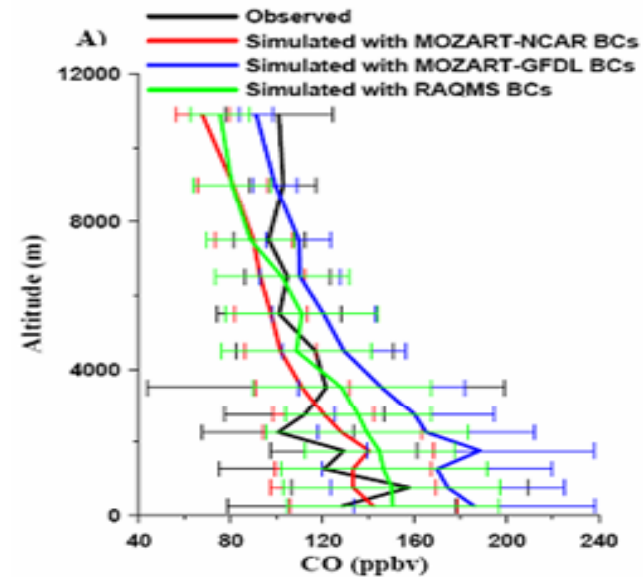
Analysis Approach



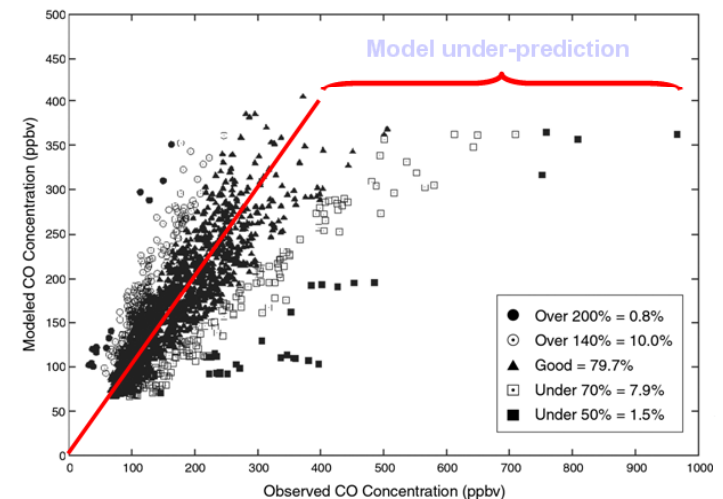
What do the agreements and disagreements with observations tell us?

Possible Reasons for Discrepancies:

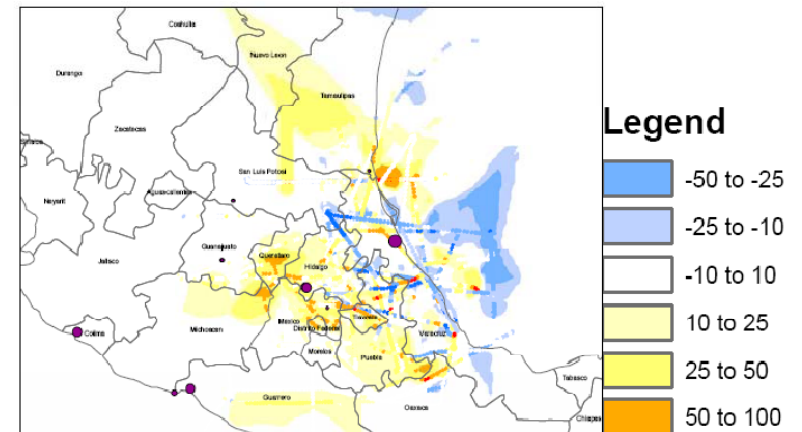
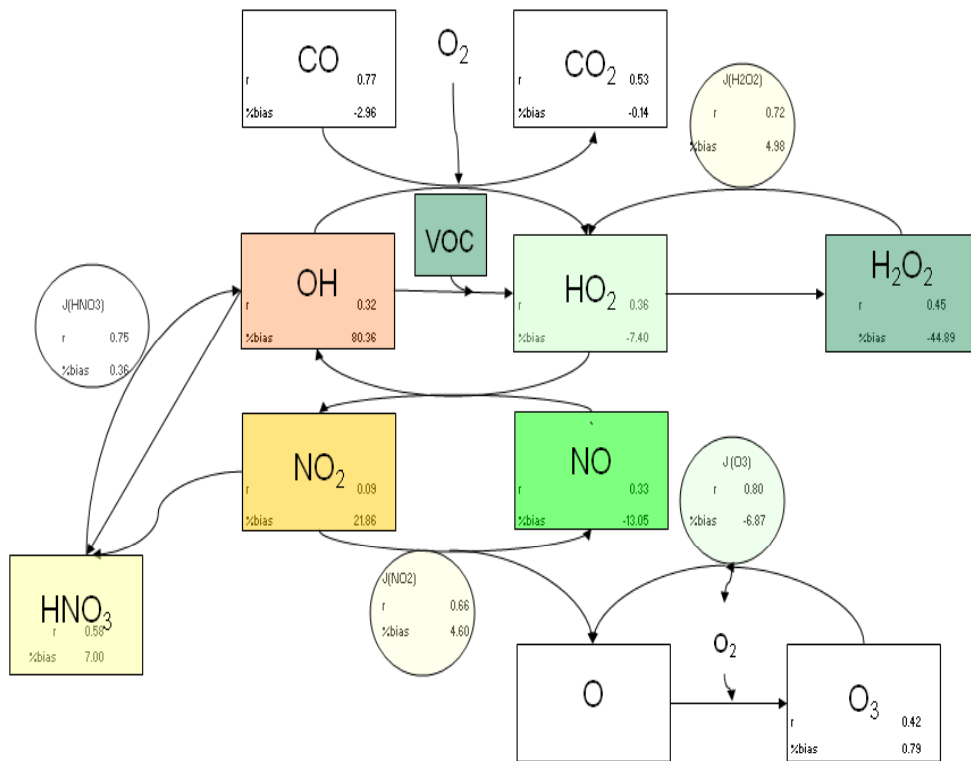
- Emissions
- Meteorology
- Chemical processes
- Inaccuracy of measurements and representativeness



Post-mission analysis has shown that the inventory seems good for most species, except for high CO and BC observations in the Yellow Sea



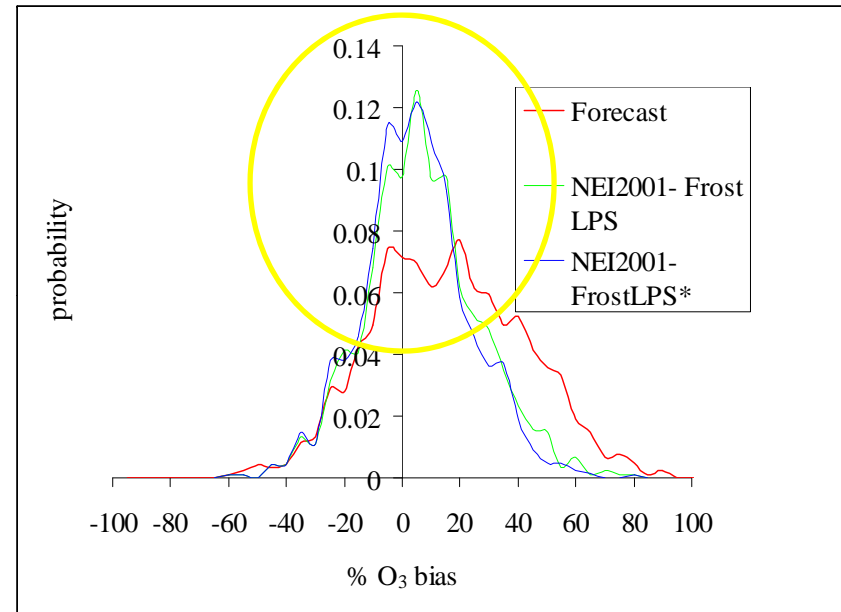
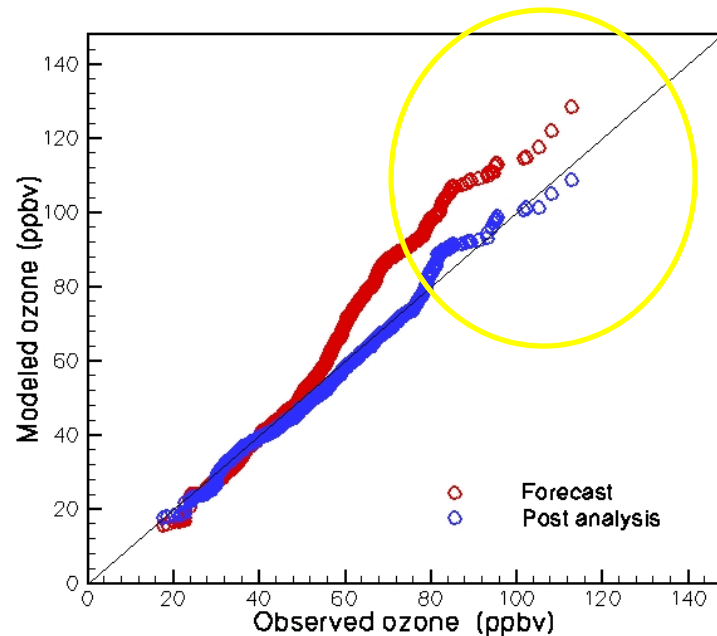
Characterization of Errors



Spatial errors of
JNO₂

The comprehensive set of observations allows analysis of cycles

Documenting improvement (ICARTT)



Left: Quantile-quantile plot of modeled ozone with observed ozone for DC-8 platform, data points collected at altitude less than 4000m, STEM-2K3, Forecast: NEI 1999, Post Analysis: NEI2001-Frost LPS*. MOZART-NCAR boundary conditions Right: Probability distribution of % ozone bias for Forecast (NEI 1999) and post analysis runs (NEI2001-FrostLPS and NEI2001-FrostLPS*) for DC-8 measurements under 4000m.

Advanced Data Assimilation Techniques Provide Data Fusion and Optimal Analysis Frameworks

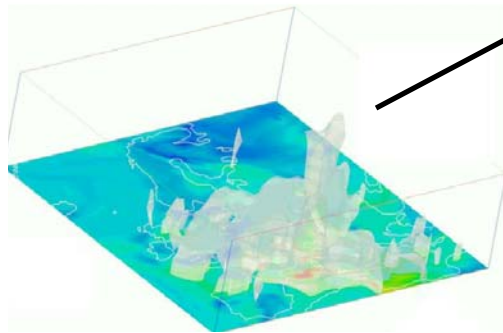
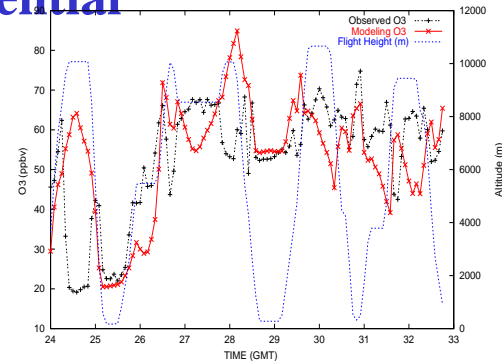
-- Treatment of Error is Essential

Model ~~vs~~ Observations
+

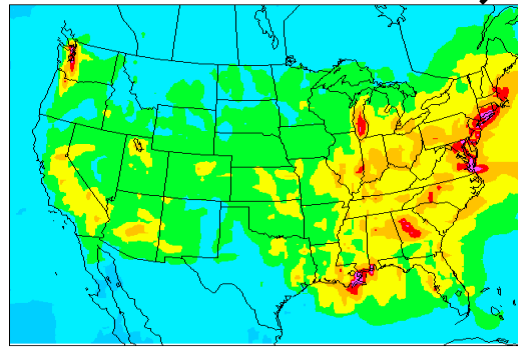
Example 4dVar:

Cost function

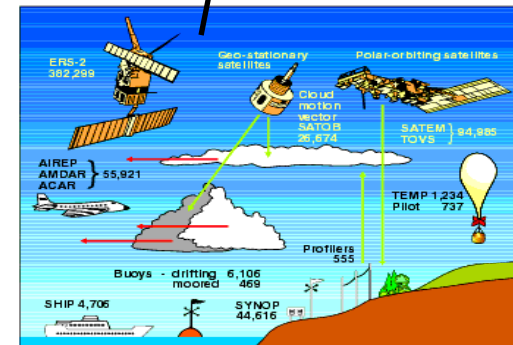
$$\min_y \psi(y) = \|\mathbf{y} - \mathbf{y}^b\|_{\mathbf{B}^{-1}}^2 + \|\mathbf{H} \cdot \mathbf{M}(y) - \mathbf{Q}\|_{\mathbf{R}^{-1}}^2$$



Current knowledge of the state



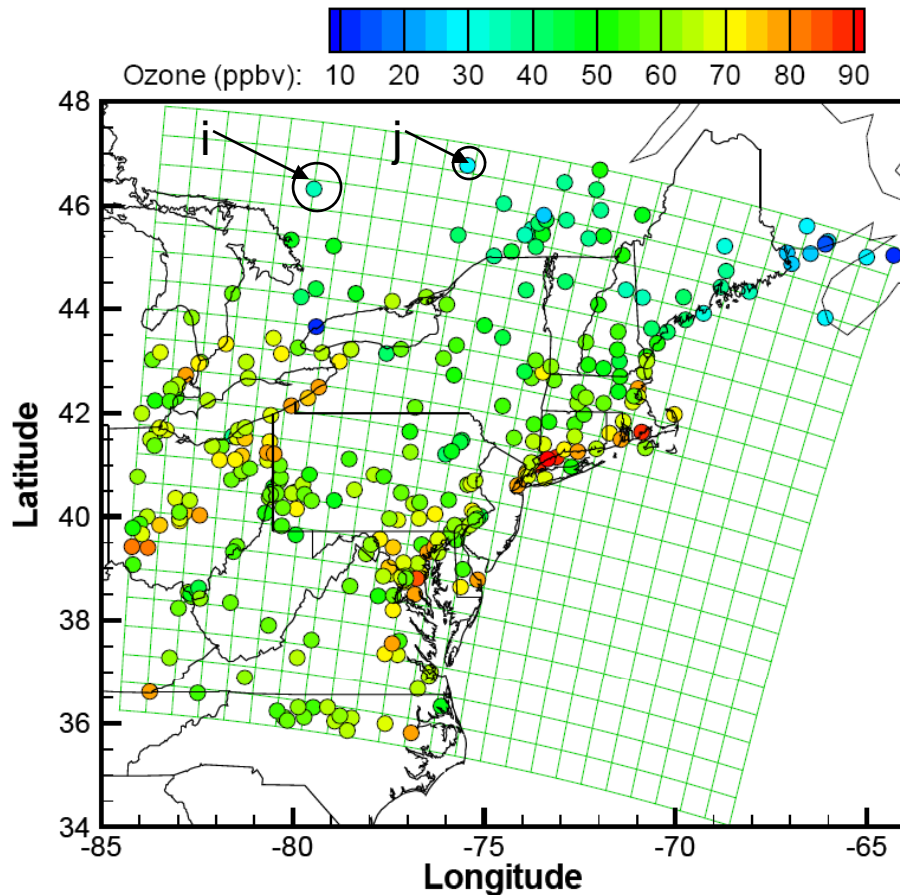
Model information consistent with physics/chemistry



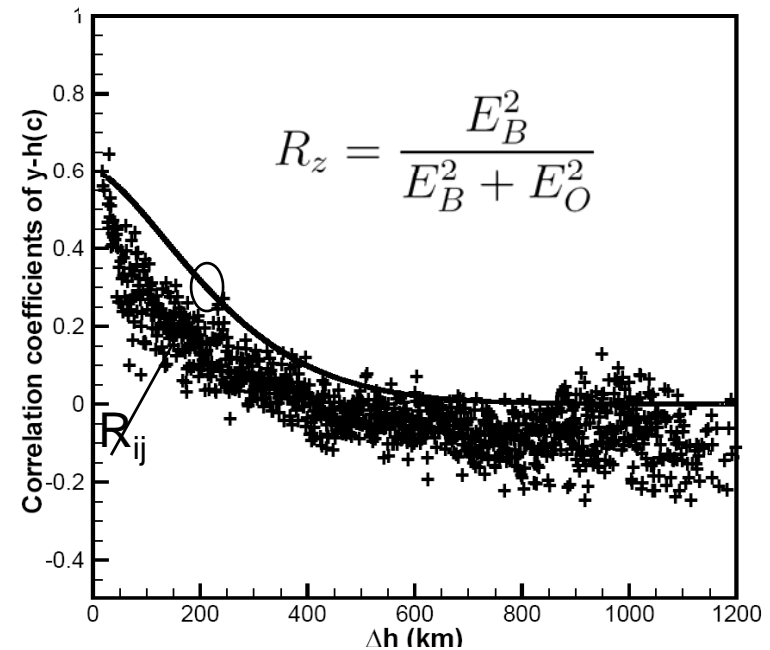
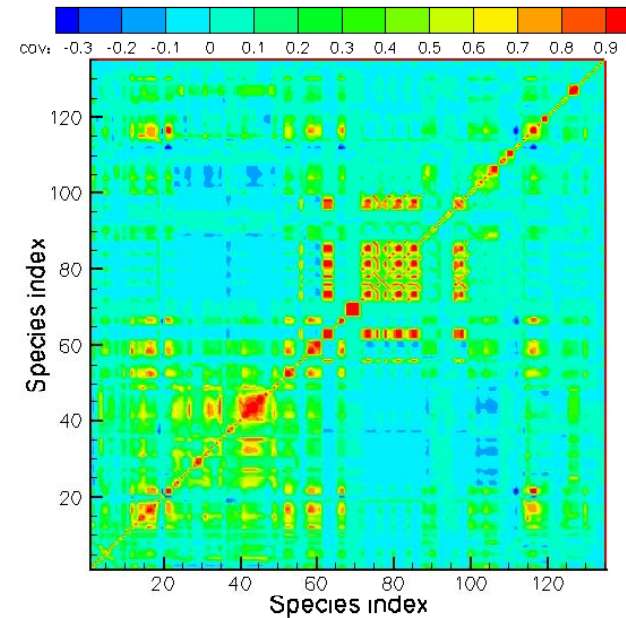
Observations information consistent with reality

The system is very under-determined – need to combine heterogeneous data sources with limited spatial/temporal information

Observational Method (Hollingsworth- Lönnerberg) for Background Errors



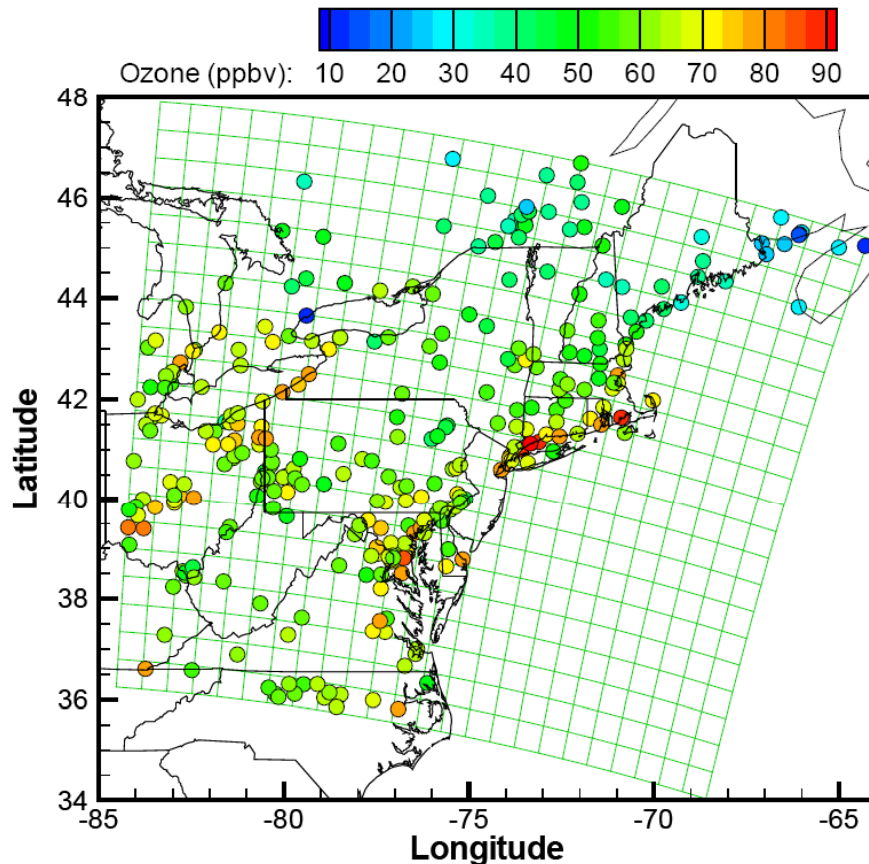
$$R_{ij} = \frac{(y^i - h^i(c))(y^j - h^j(c))}{\sqrt{(y^i - h^i(c)) \cdot (y^j - h^j(c))}}$$



$$E = \sqrt{E_B^2 + E_O^2} = \sqrt{10^2 + 8^2} \approx 12.8 \text{ ppbv}$$

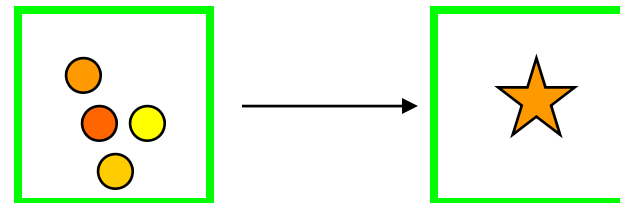
Observational error

$$J = \frac{1}{2} [c_0 - c_b]^T B^{-1} [c_0 - c_b] + \frac{1}{2} [y - h(c)]^T O^{-1} [y - h(c)]$$



Observational Error:

- Representative error
- Measurement error

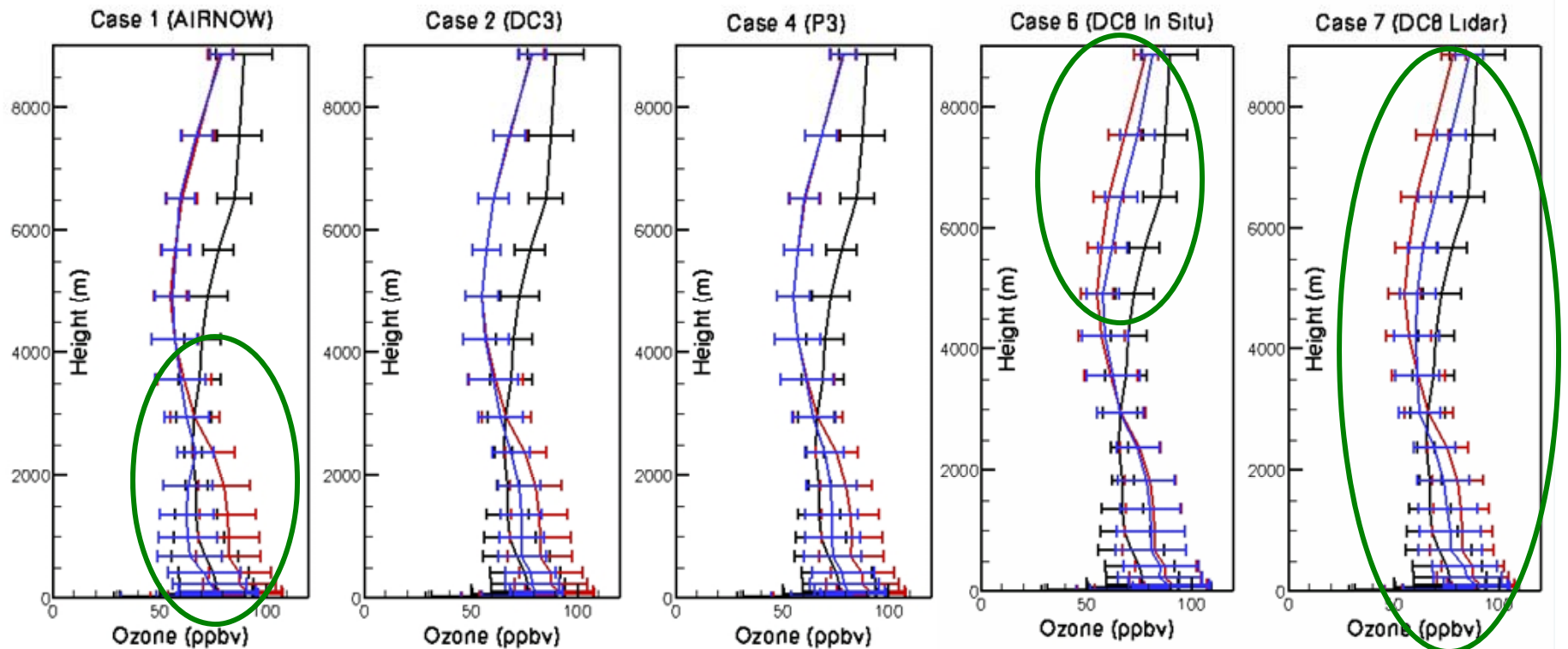


Observation Inputs

- Averaging inside 4-D grid cells
- Uniform error (8 ppbv)

Case	Assimilated Observations	Time	Number
1	AIRNOW	1300–2400 UT, hourly	2075
2	DC3	1852–2356 UT	412
3	MOZ-FN, MOZ-NF	1947–2007 UT, 2238–2252 UT	38
4	P3	1412–2207 UT	208
5	AIRMAP	1215–2400 UT	192
6	DC8-In	1416–2207 UT	138
7	DC8-Li	1429–2137 UT	465
8	RHODE, RONBR	1810–1822 UT, 1900–1921 UT	35
9	All above	1200–2400 UT	3563

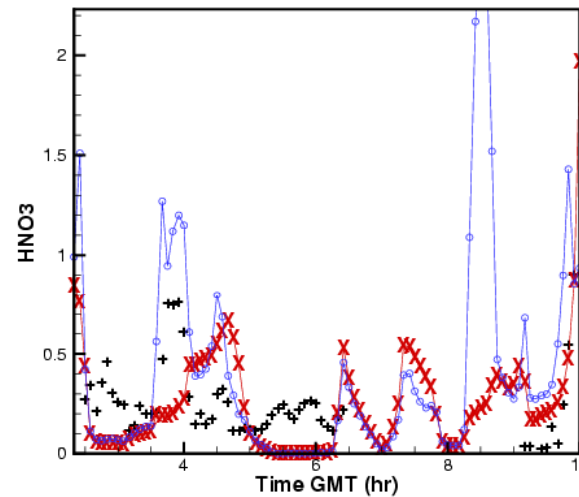
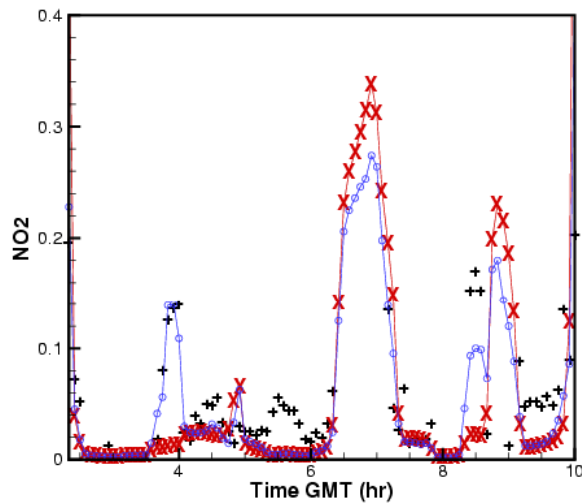
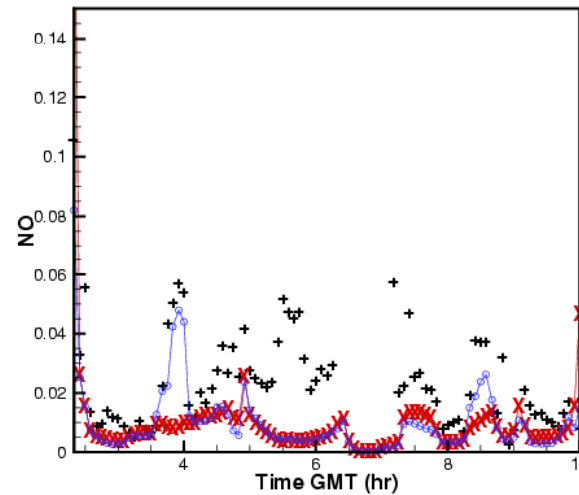
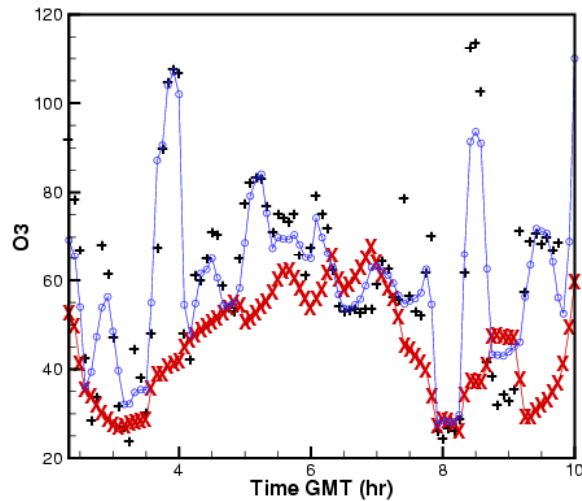
Information content of various observations evaluated by different combinations of data sets assimilated –
the importance of measurements above the surface.



Surface-only

Lidar-DC8

Assimilating multiple species



Measurement uncertainties:

O3: 8%

NO: 20%

NO2: 20%

HNO3: 100%

PAN: 100%

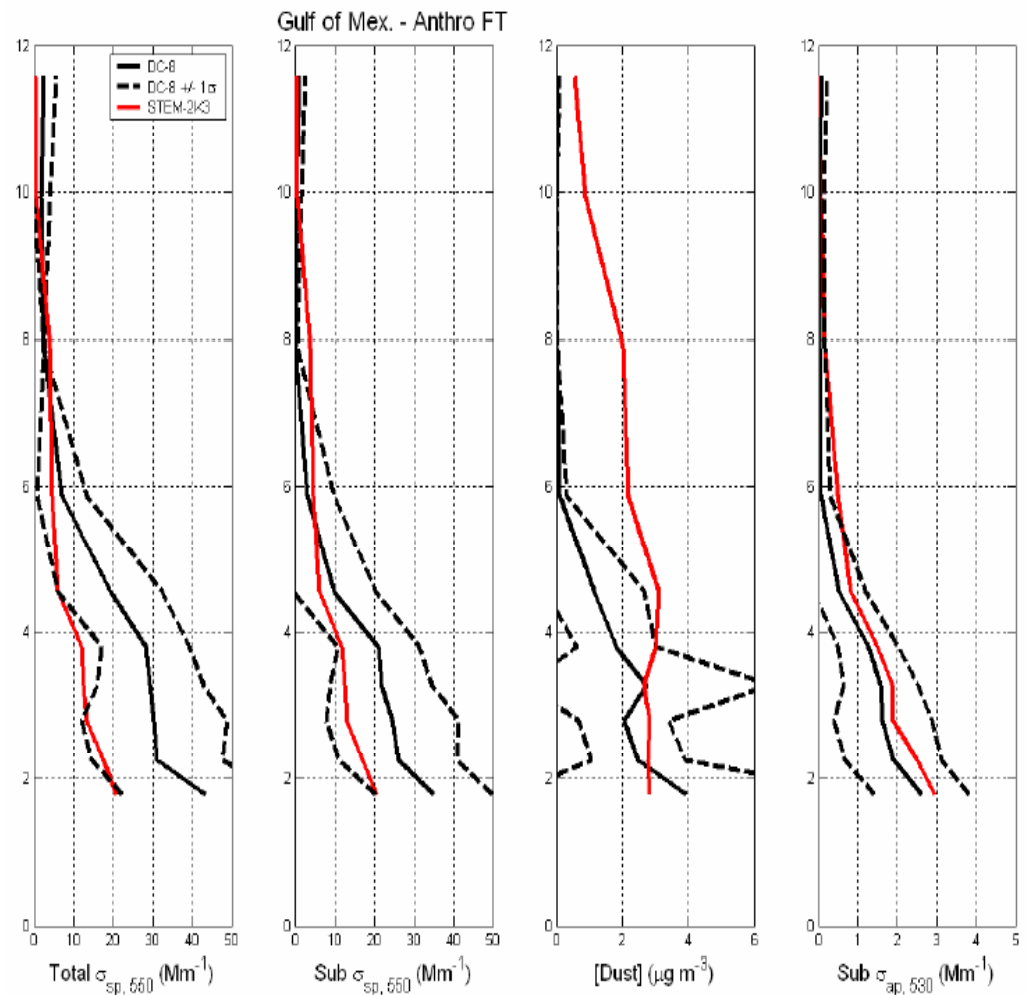
RNO3: 100%

Aerosol Issues

- 1) How well do models replicate vertical structure of anthropogenic aerosols?
- 2) How well do models predict column integrated aerosol optical properties?

Approach: Observations compared to size distributions and optical properties prescribed and/or generated by chemical transport models in order to evaluate the fidelity of the model's representation of the atmospheric aerosol.

There are many challenges: matching size distributions, partitioning, number distributions, etc.



Cam's Thesis (2008)

**Models can Also Add Value to the Observations:
e.g., 4-d context, trajectory analysis, observation “filling” using
trajectories, etc.**

How Representative are the Aircraft Observations?

