

# Chemical Data Assimilation and Air Quality Forecasting using CMAQ

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# Introduction

- Assimilating real-time observations is essential in weather forecasting
- AIRNOW observations are available in near real time. It has potential to be applied operationally for air quality forecasting
- Optimal interpolation (OI) is simple to apply and computation overhead is minimal
- In data assimilation, background error covariance ( $B$ ) is important
  - Determines the weighting between observations and a priori background
  - Determines the spread of the increment in space and between variables

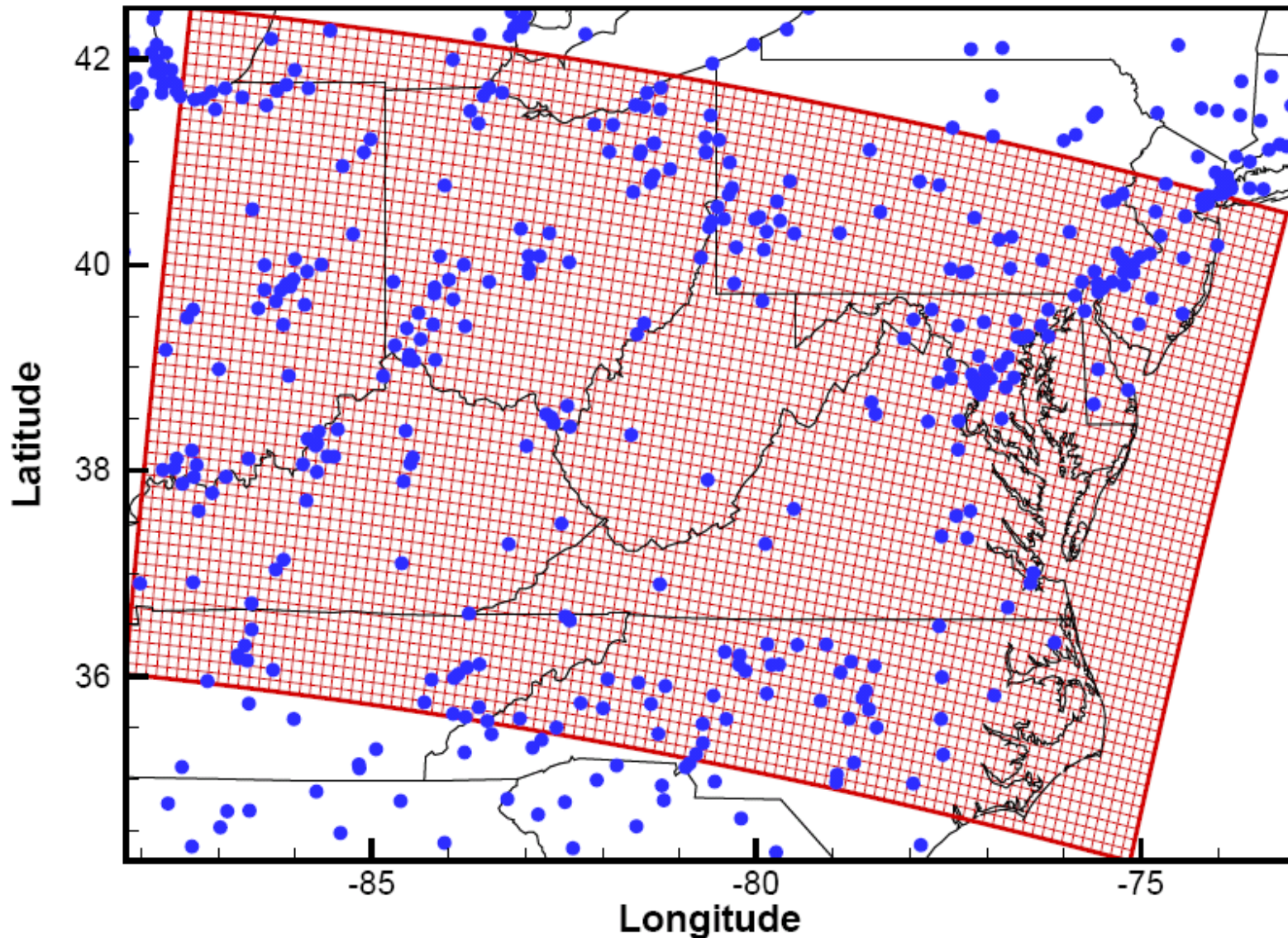
# Optimal Interpolation (OI)

- OI is a sequential data assimilation method. At each time step, we solve an analysis problem

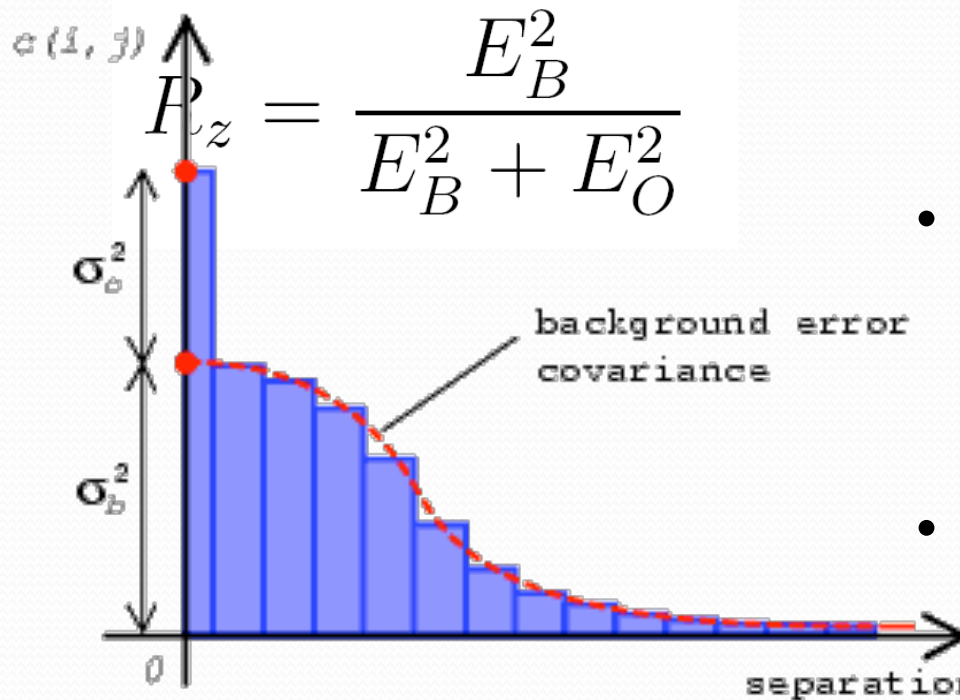
$$X^a = X^b + BH^T (HBH^T + O)^{-1} (Y - HX)$$

- We assume observations far away (beyond 1.5 background error correlation length scale) have no effect in the analysis
- In the current study, the analysis is solved in two steps: first in horizontal direction, then vertical

# Domain, Grid (100x60), and AIRNOW Stations

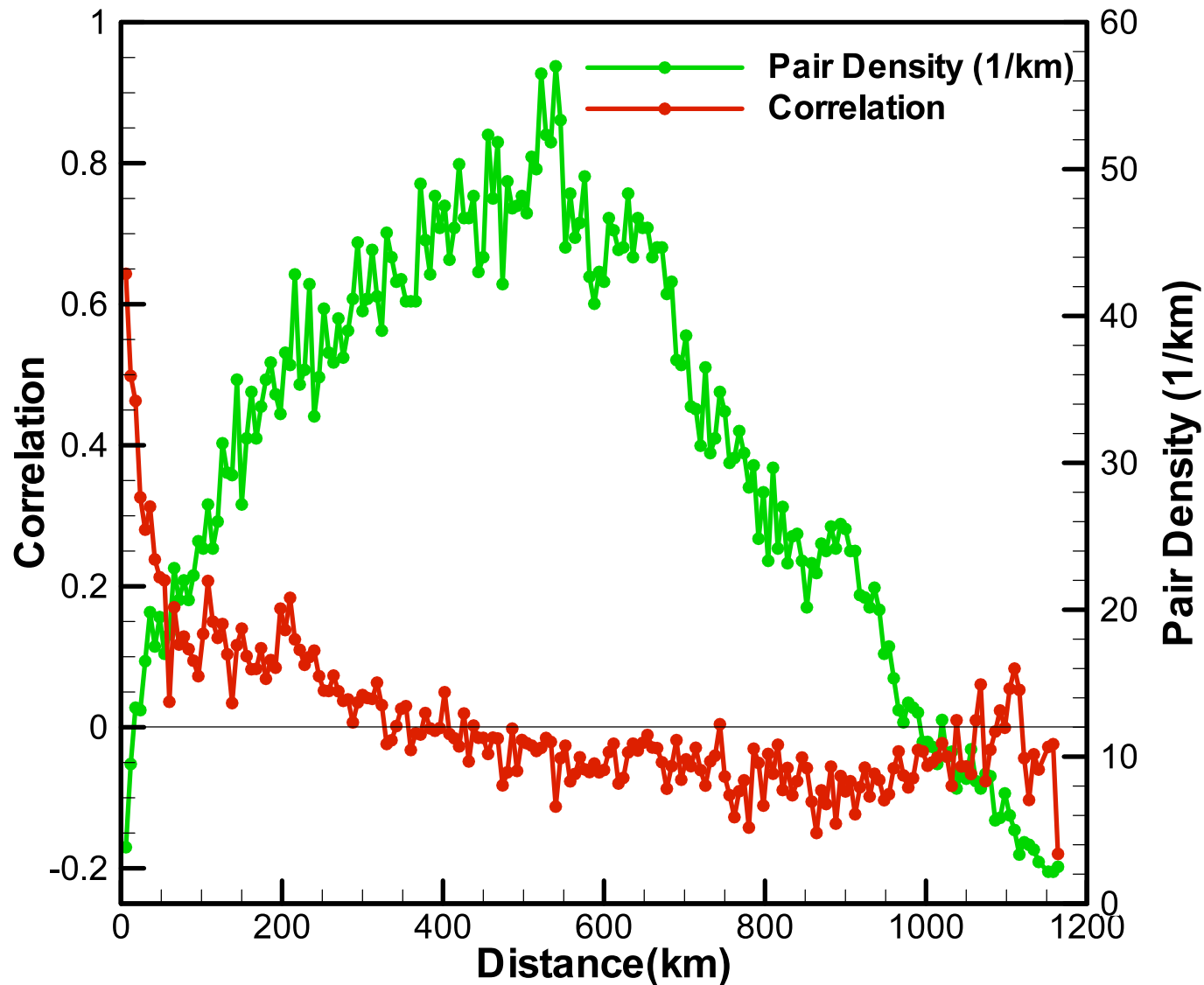


# Estimate Model Error Statistics with Hollingsworth-Lönnerberg Method



- At each station, calculate differences between forecasts (B) and observations (O)
- Pair up AIRNOW stations, and calculate the correlation coefficients between the two time series at the paired stations
- Plot the correlation as a function of the distance between the two stations,

# Horizontal Error Statistics

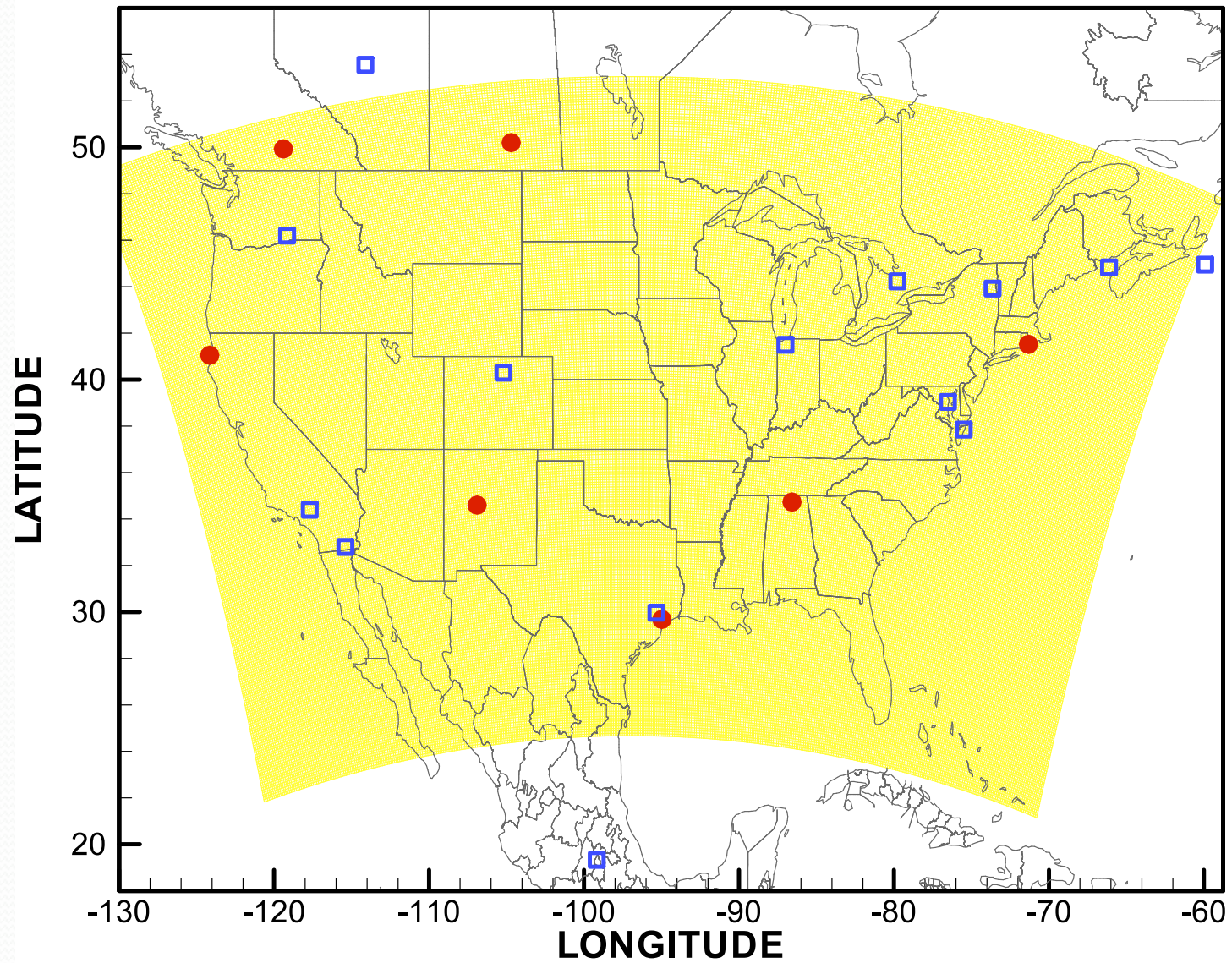


$E_B \sim$   
14.2 ppbv

$E_O \sim$   
3.3 ppbv

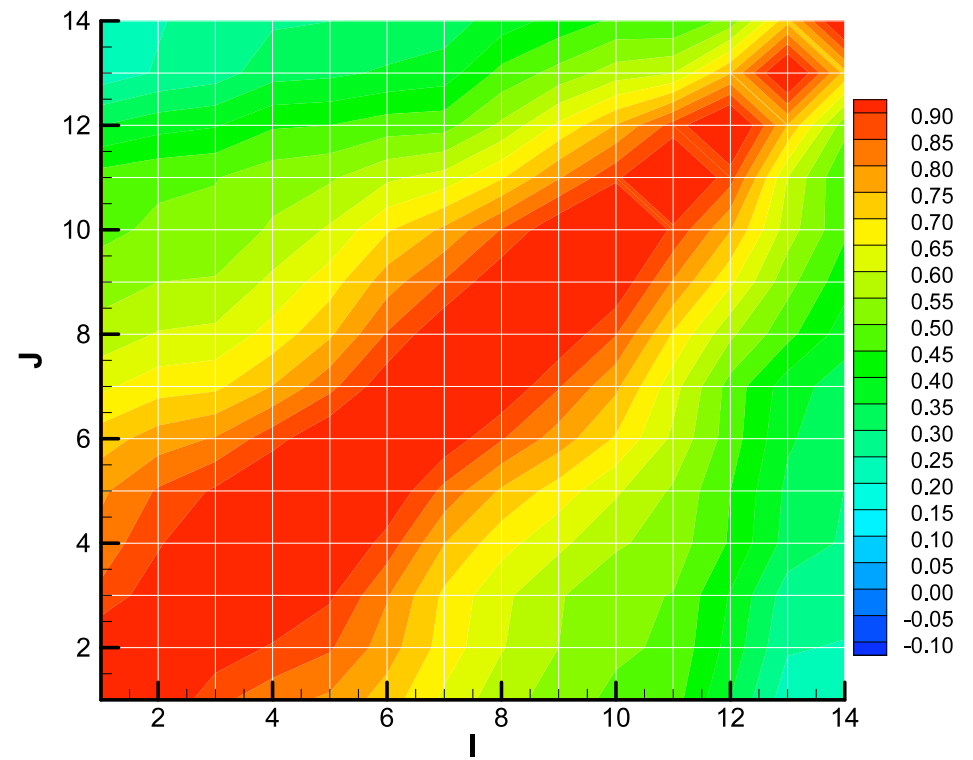
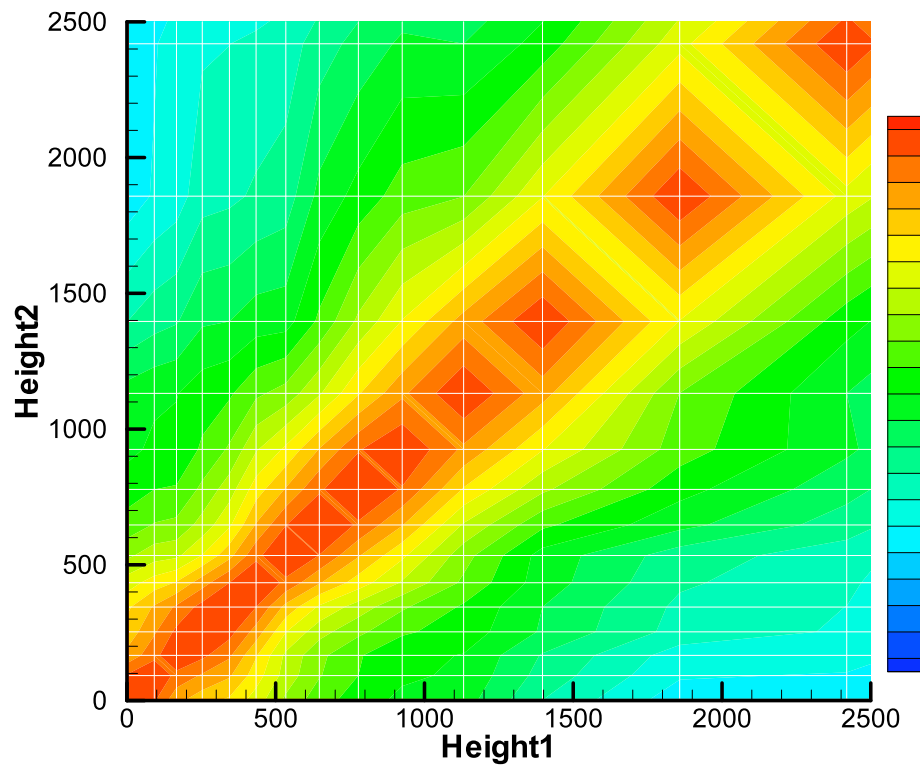
Correlation  
length:  
60 km

# Ozonesonde: Vertical Error Statistics





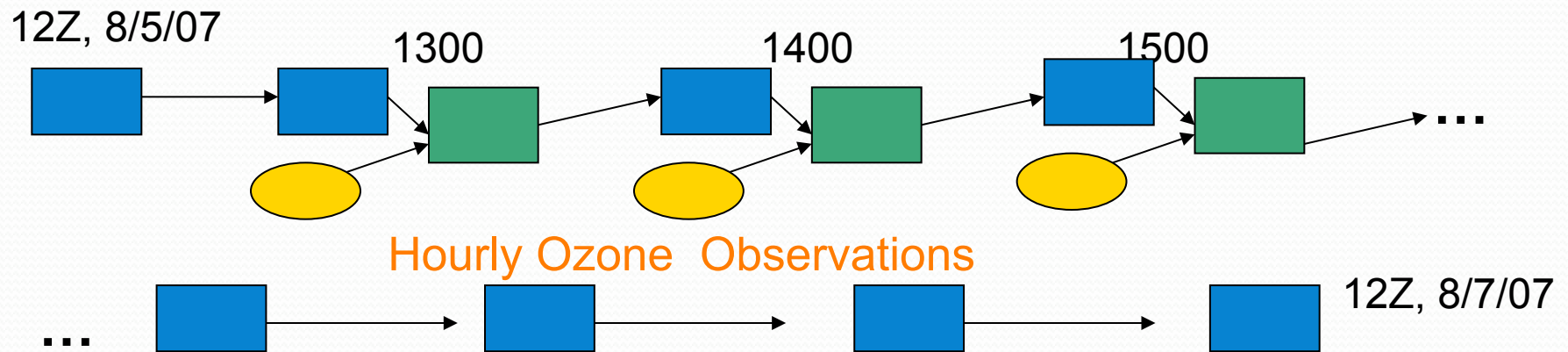
# Vertical Correlation



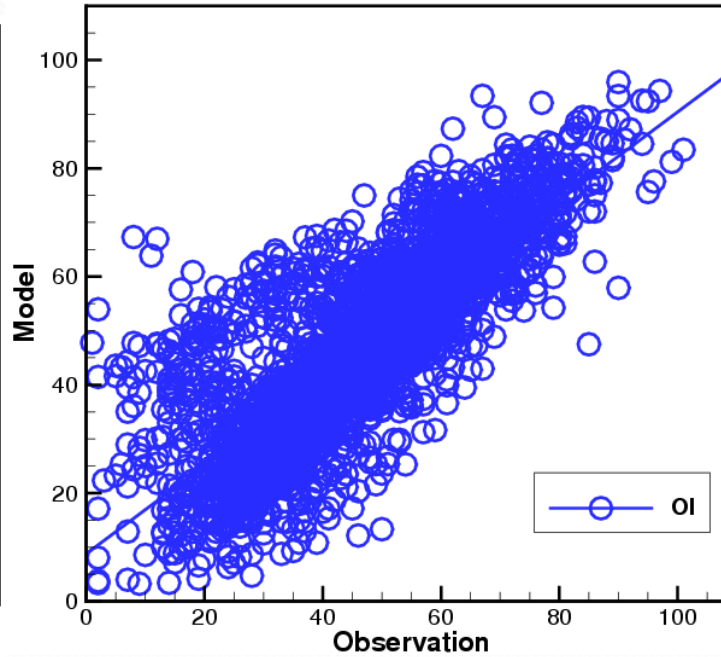
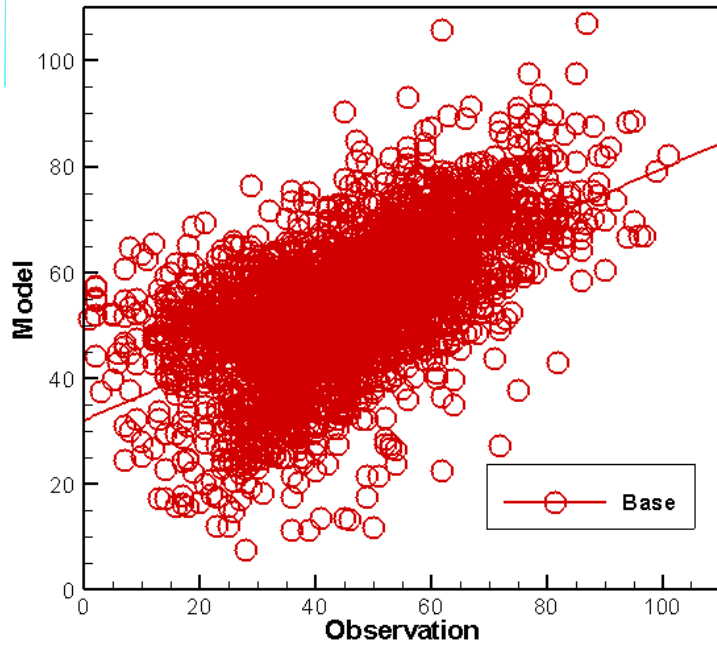


# Setup of OI Assimilation Tests

- CMAQ 4.6, Model starts at 12Z, 8/5/07
- Hourly AIRNOW observations assimilated in first day
- Model continues to run another day without observations



# Observation-Prediction (in ppbv)

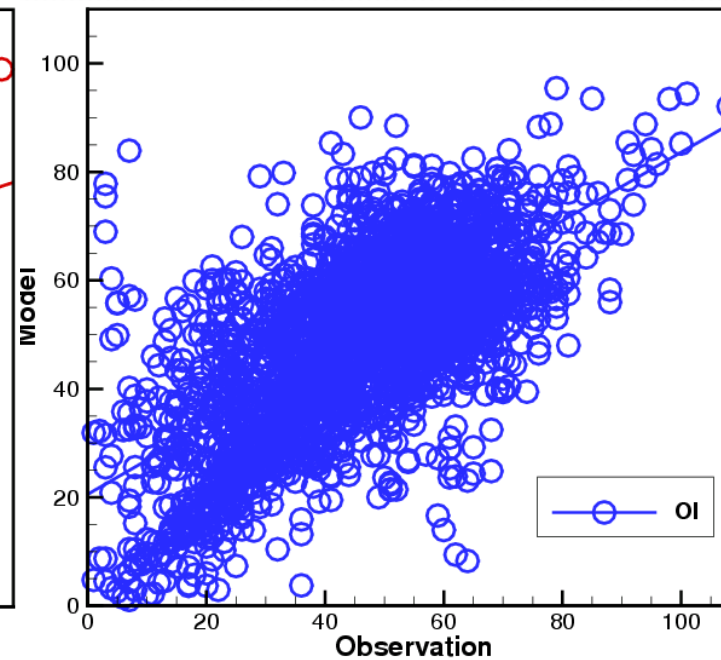
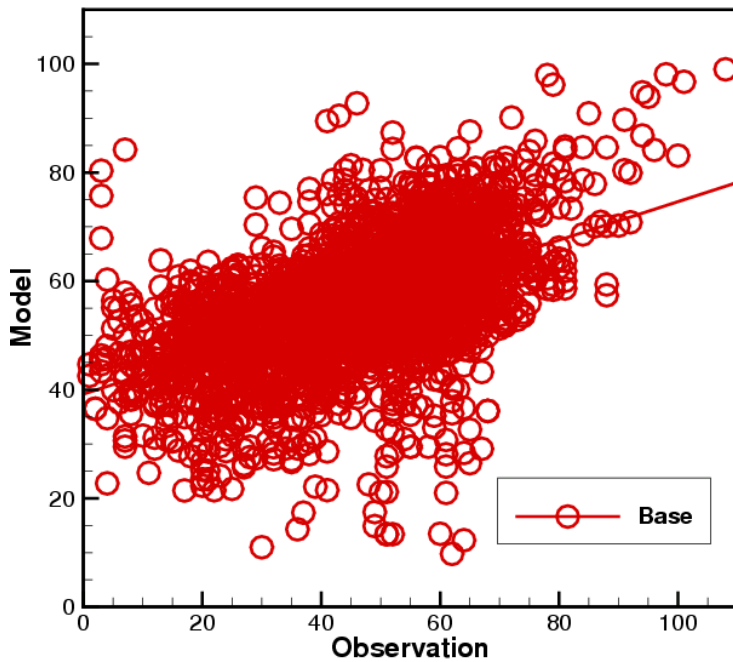


Day 1

$R=0.59$

$R=0.81$

1300 - 2400 Z



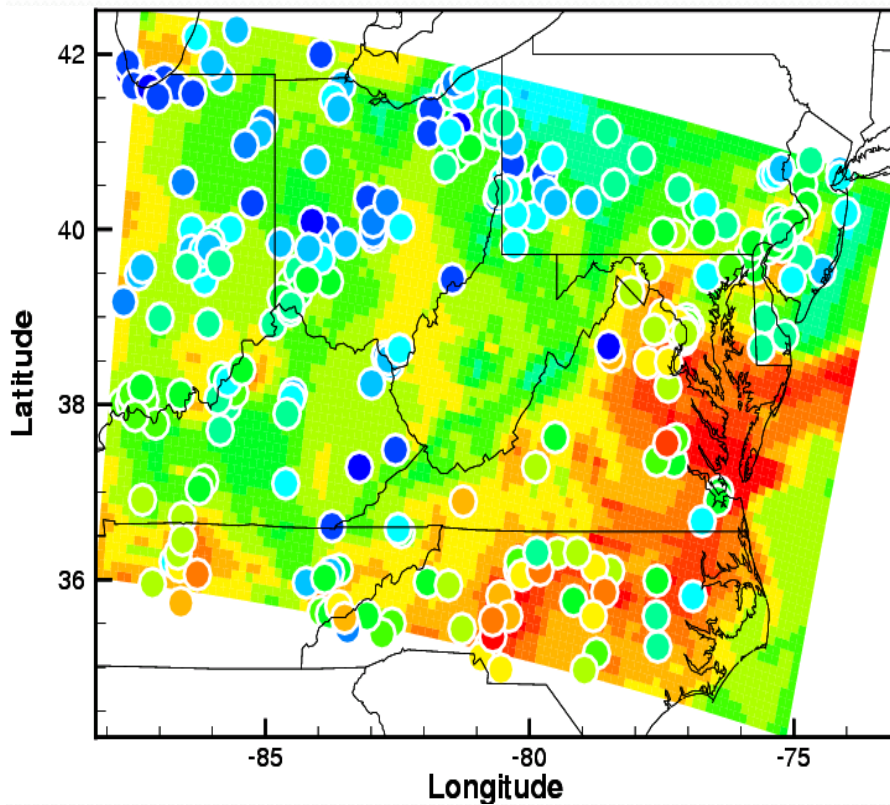
$R=0.56$

$R=0.68$

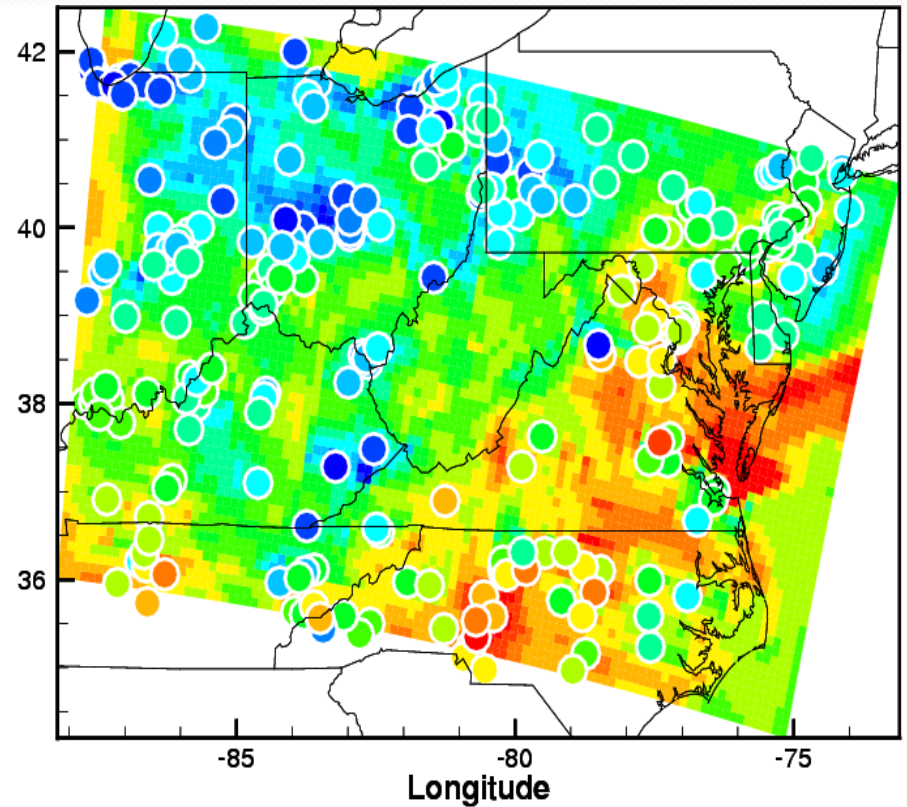
Day 2

# Surface O<sub>3</sub> at 1800Z, 8/5/07

## Base Case

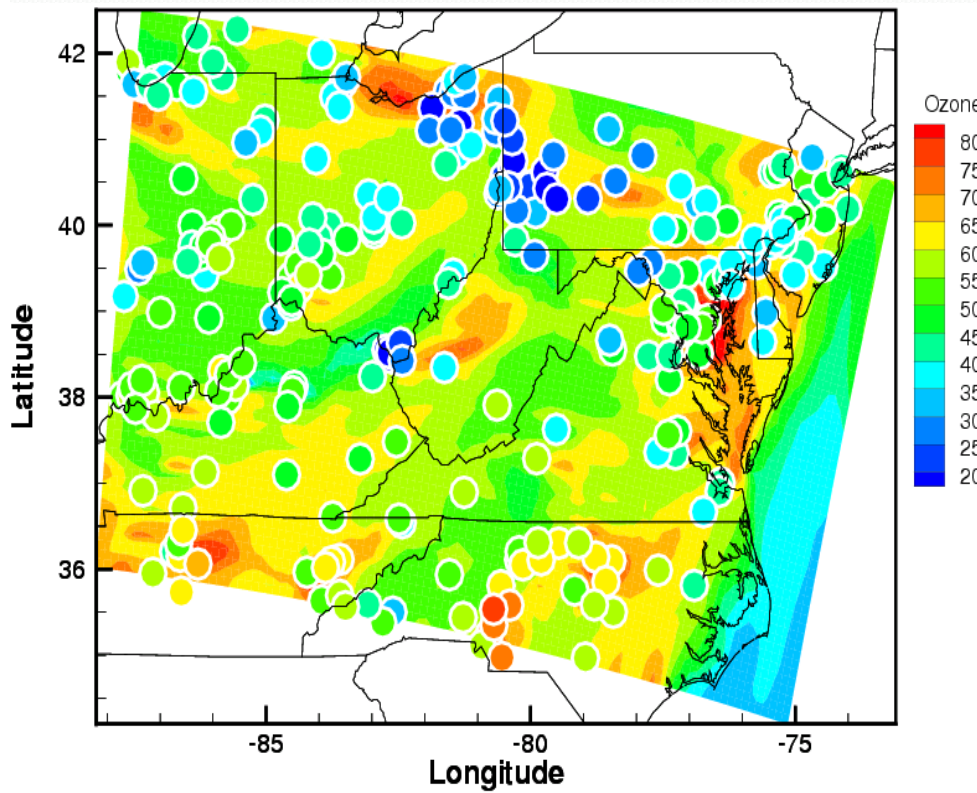


## OI (Analysis)

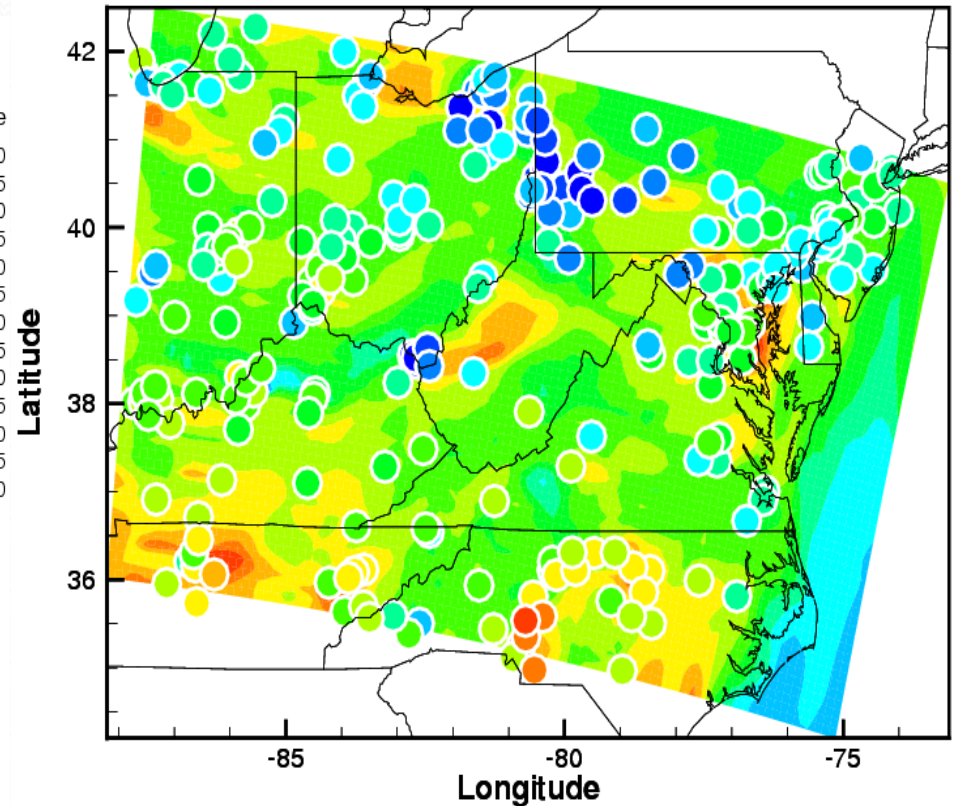


# Surface O<sub>3</sub> at 1800Z, 8/6/07 (6 hours after OI)

## Base Case



## OI (Forecast)



# 4D-Var Data Assimilation

1. CMAQ 4.5 Adjoint was developed by A. Sandu *et al* (adjoint available for transport and chemistry)
2. Assimilation time window is reduced to 15 hours
3. Only initial O<sub>3</sub> are adjusted to minimize the cost functional (using L-BFGS-B routine)

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

## 4D-Var .vs. OI

	DA method	<b>B</b>	Bias (Day1)	RMS (Day1)	Bias (Day2)	RMS (Day2)
1	n/a	n/a	8.3	15.9	8.7	16.3
2	4D-Var	Diagonal	-0.8	11.0	7.6	15.6
3	OI	Diagonal	2.6	12.7	7.5	15.8
4	OI	$H \begin{bmatrix} \times \\ \times \\ \times \end{bmatrix} V$	-1.3	13.2	3.1	12.8
5	4D-Var	$H \begin{bmatrix} \times \\ \times \\ \times \end{bmatrix} V$	?	?	?	?

Biases and RMS errors are calculated at day time (8am-8pm LT) . Units : ppbv.

1. 4D-Var (V4.5) gives slightly better results than OI (V4.6)
2. “ $\begin{bmatrix} \times \\ \times \\ \times \end{bmatrix}$ ” denotes Kronecker product and “TSVD” method is used for the inverse of B matrix (Chai, *et al* , Four dimensional data assimilation experiments with ICARTT ozone measurements, JGR, 2007)

# Summary

- CMAQ model error statistics has been estimated using Hollingsworth-Lönnerberg method in both horizontal and vertical directions
- Assimilating AIRNOW observations into CMAQ model using Optimal Interpolation proves to be beneficial for the next-day ozone forecasting
- A 4D-Var data assimilation test shows slightly better results than OI with same diagonal B
- Model error covariance results are utilized to assimilate AIRNOW observations with OI approach ( $B=H\begin{bmatrix} W \\ W \end{bmatrix}V$ , implemented using TSVD method), it shows significant improvement over OI with diagonal B