

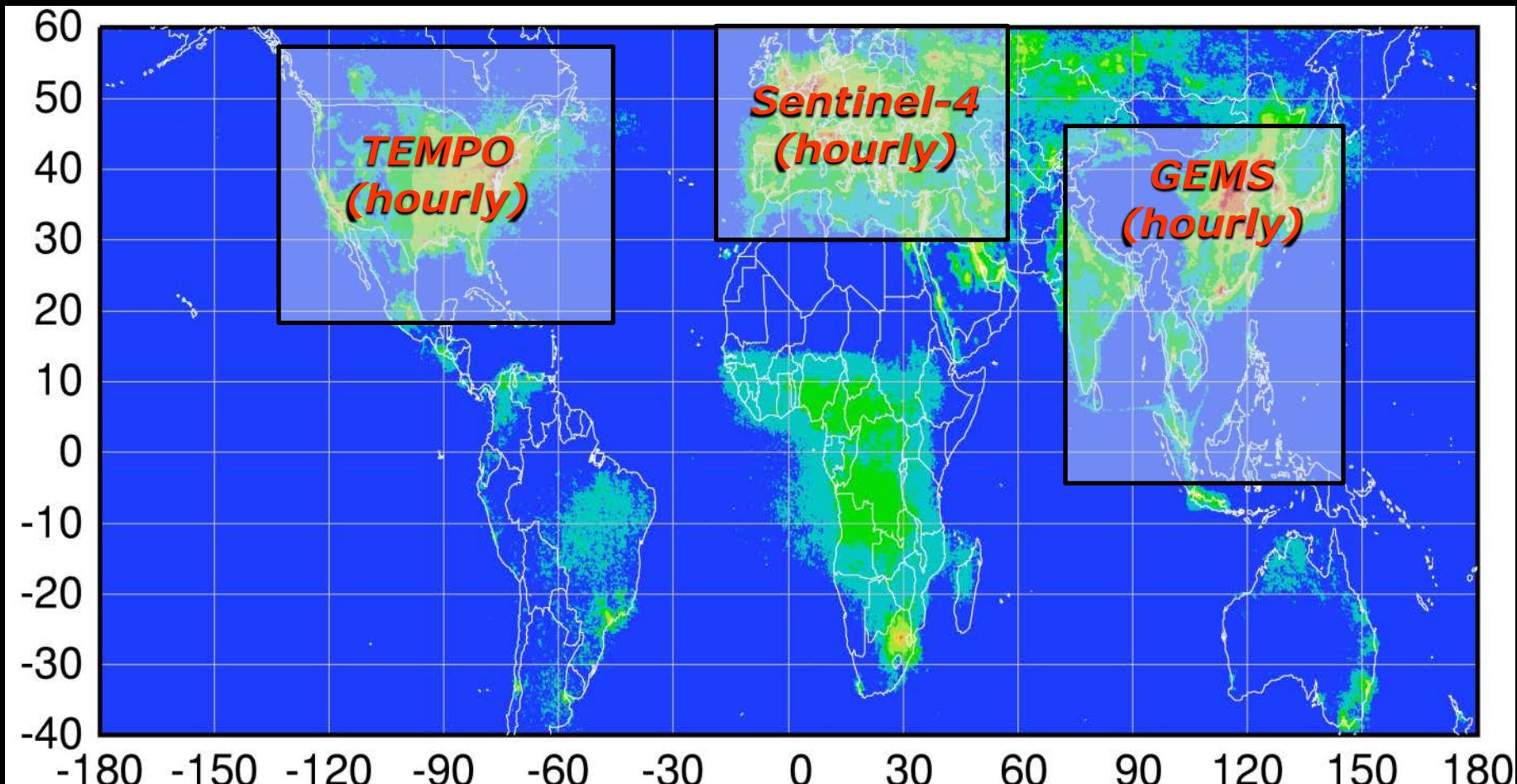
Winds and TEMPO



Ronald C. Cohen
UC Berkeley

\$ TEMPO, NASA ACMAP and GEOCAPE

High space and time resolution measurements of NO₂, H₂CO and O₃
will soon be routinely available



Preparing for new science opportunities using GEO NO₂

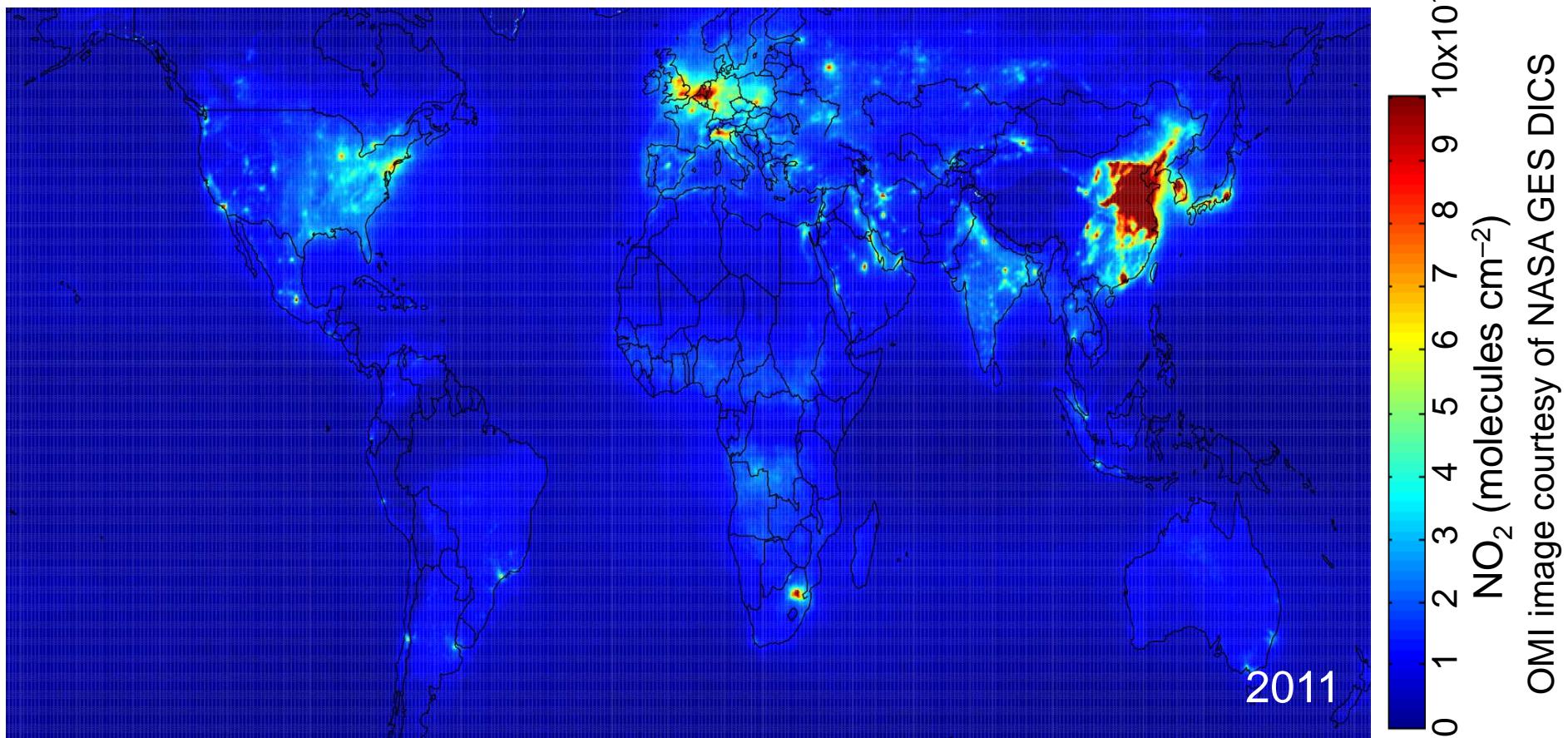
- Current State-of-the-Art
- Retrievals
- Data assimilation



Examples of Current State-of-the-Art

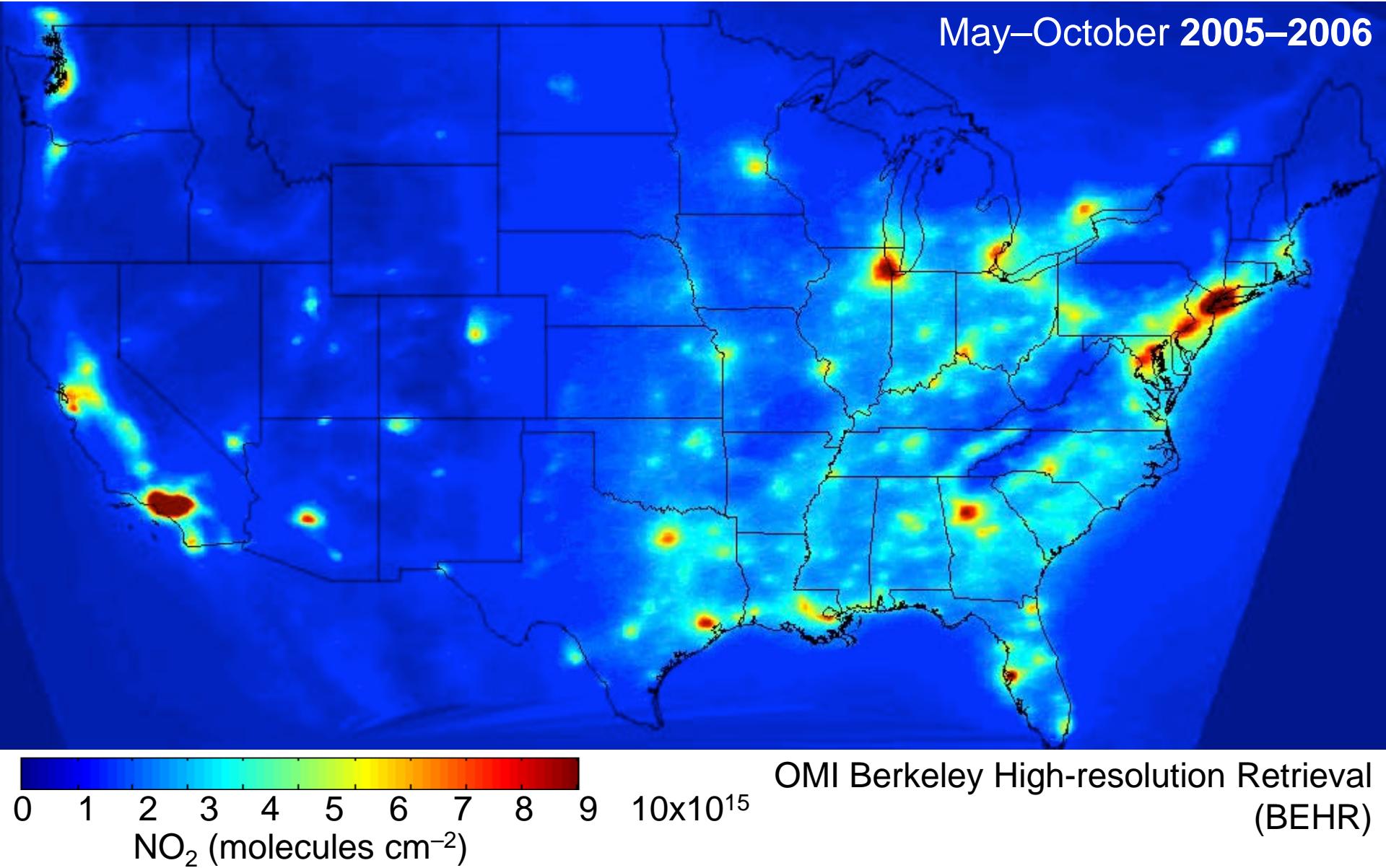


1) Beginning to understand how to quantitatively relate trends in slant columns to trends in emissions



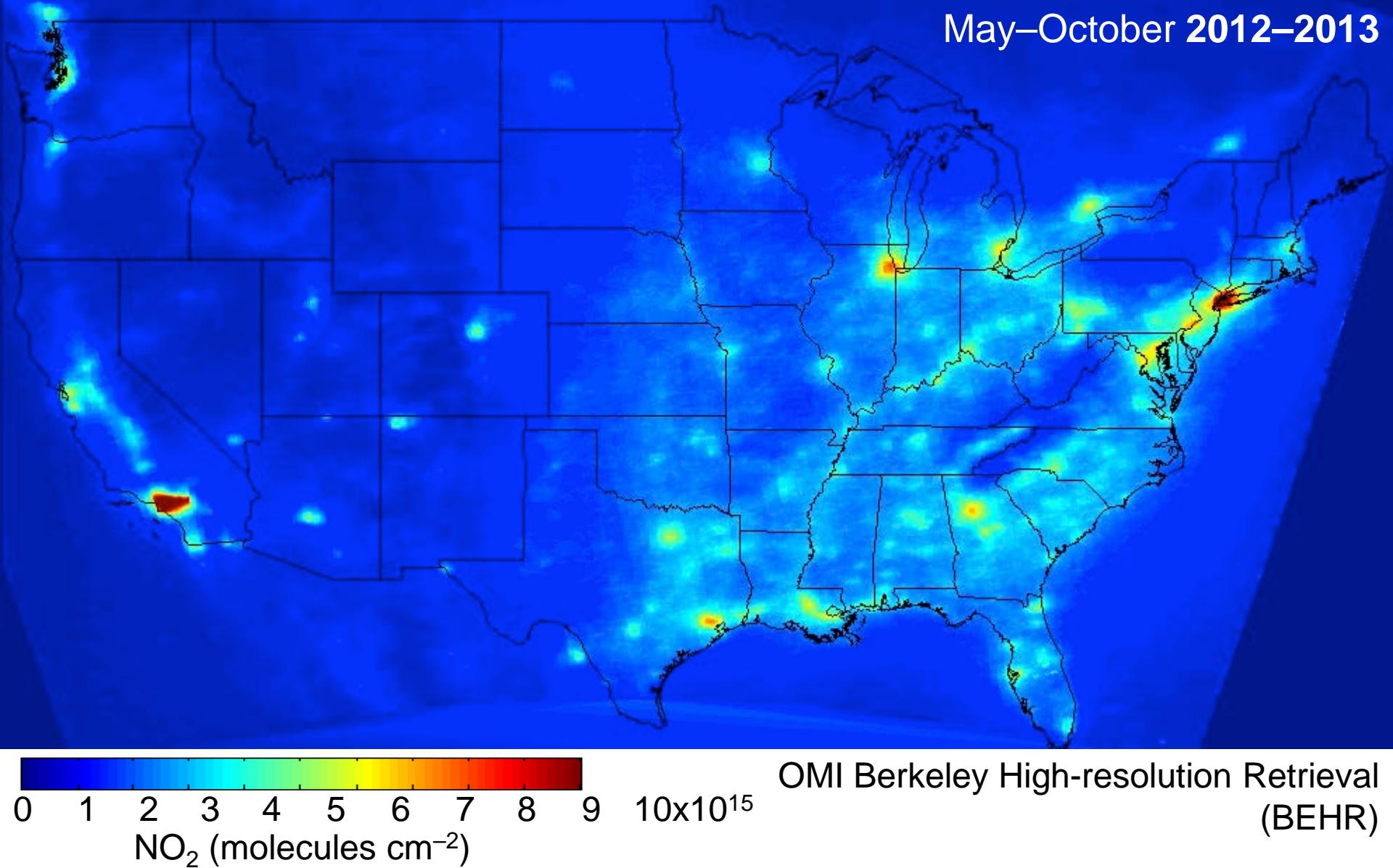
2005-2006

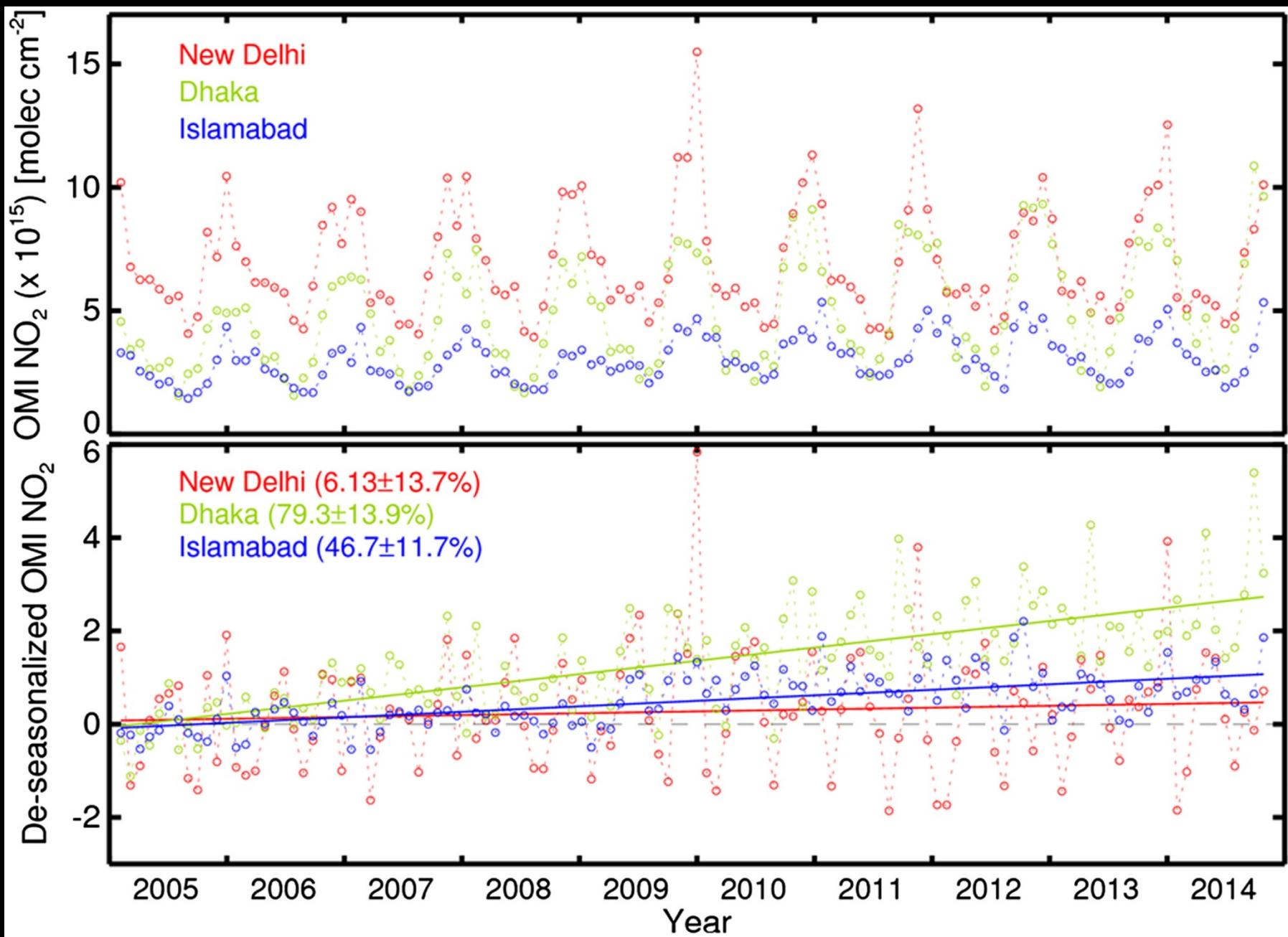
May–October 2005–2006



2012-2013

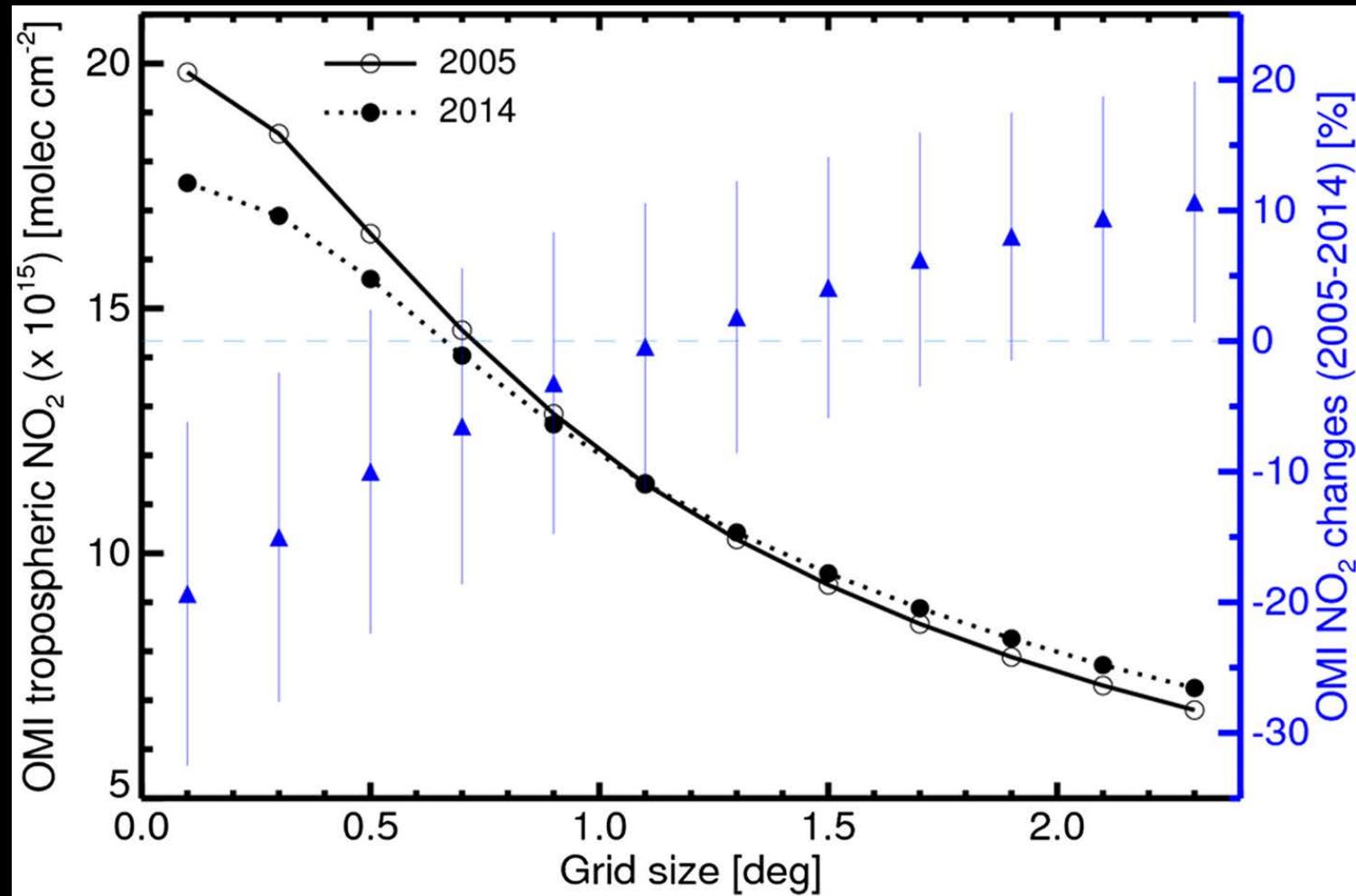
May–October 2012–2013





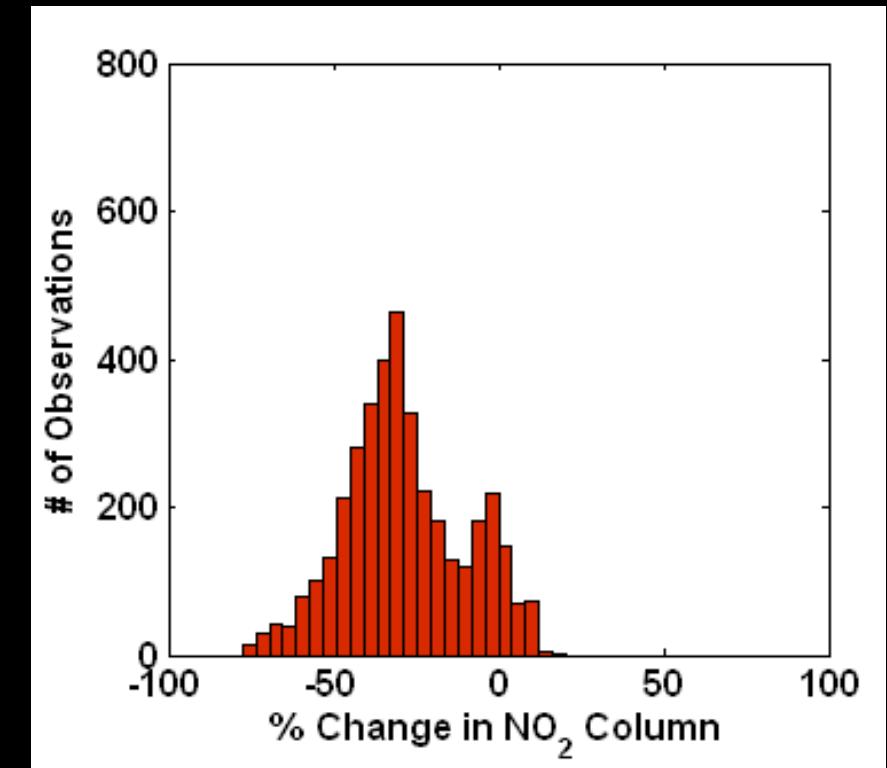
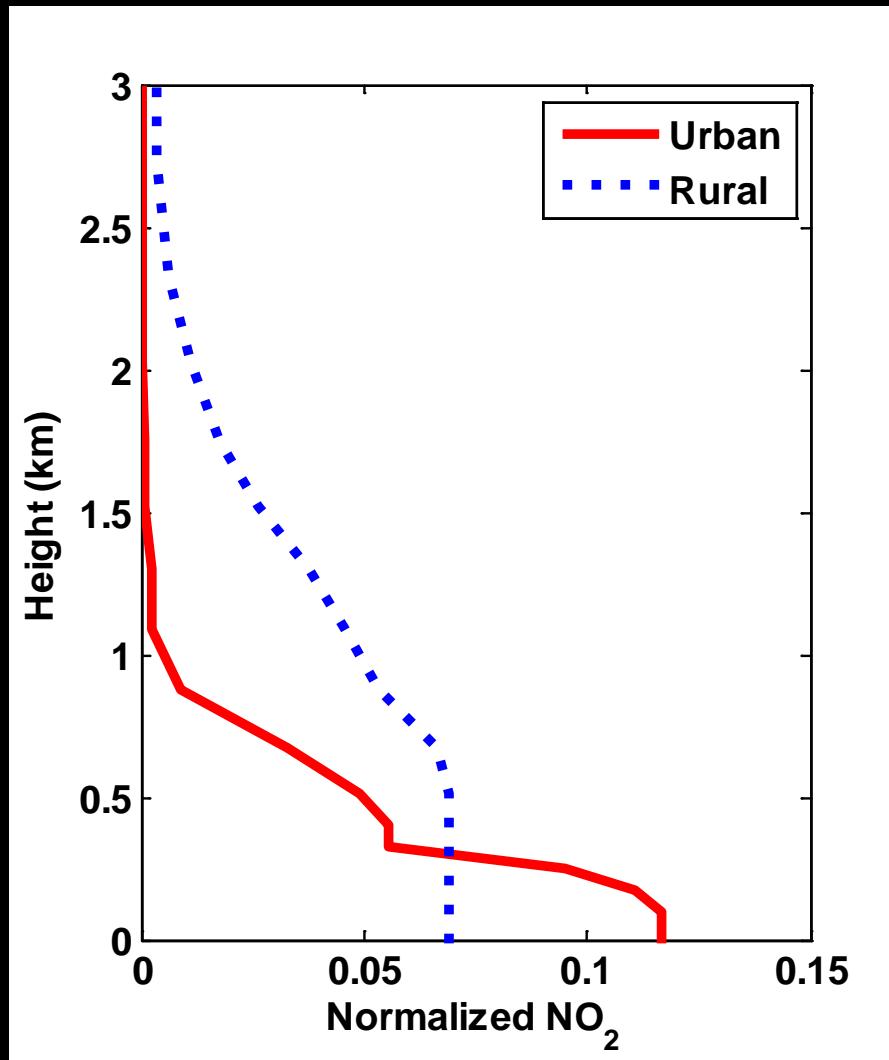
e.g. Duncan et al. JGR 2016

Resolution matters



e.g. Duncan et al. JGR 2016

Resolution matters (e.g. BEHR Retrieval)



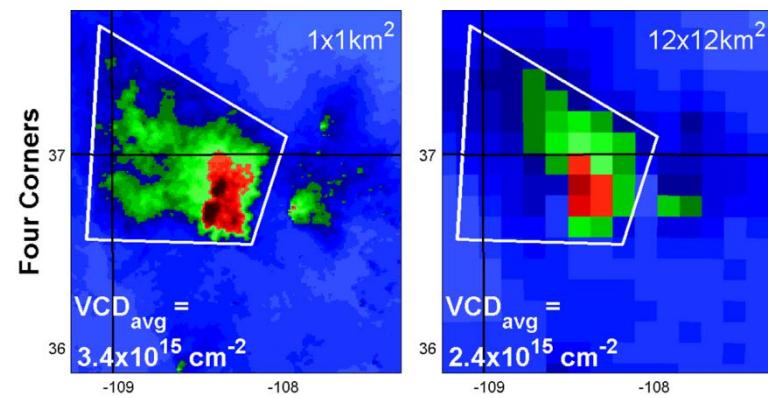
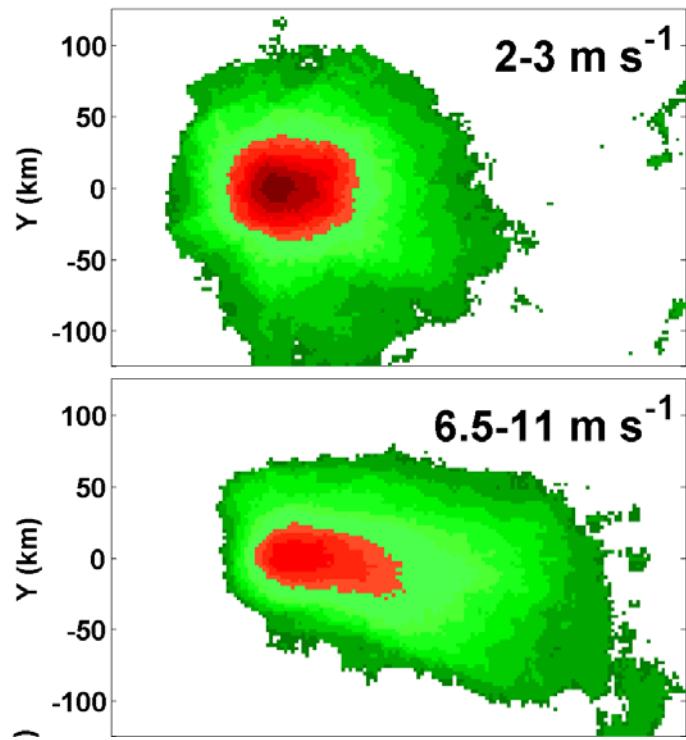
PDF of systematic errors

2) Beginning to understand how to quantitatively assess absolute emissions

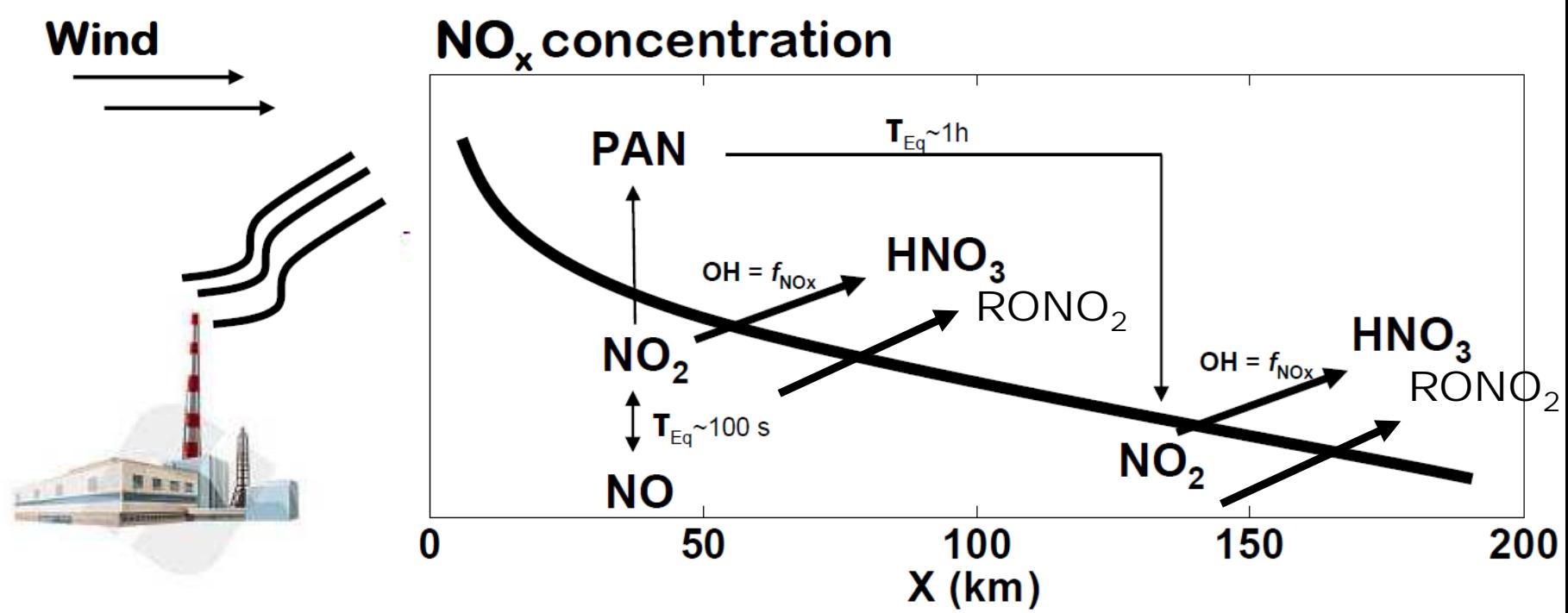
Two key ideas:

Winds are coupled to lifetime

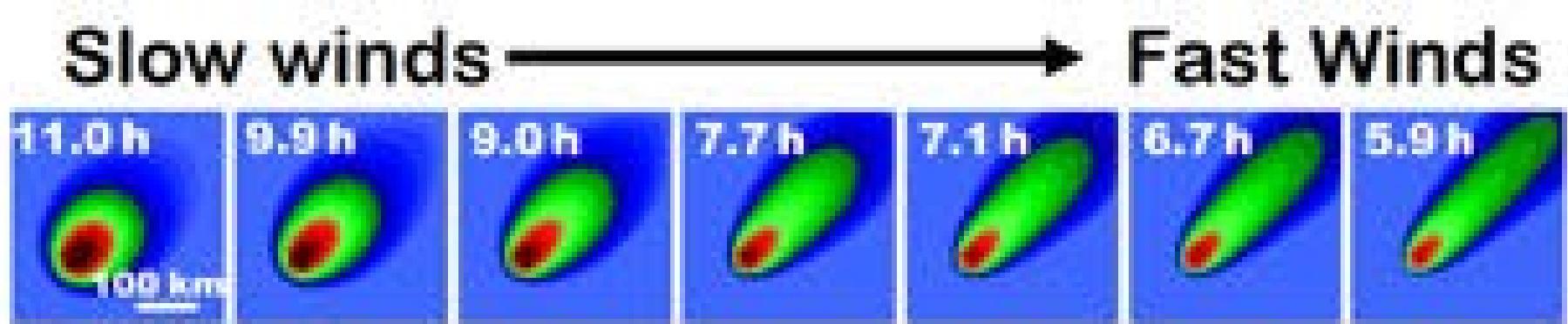
Resolution matters



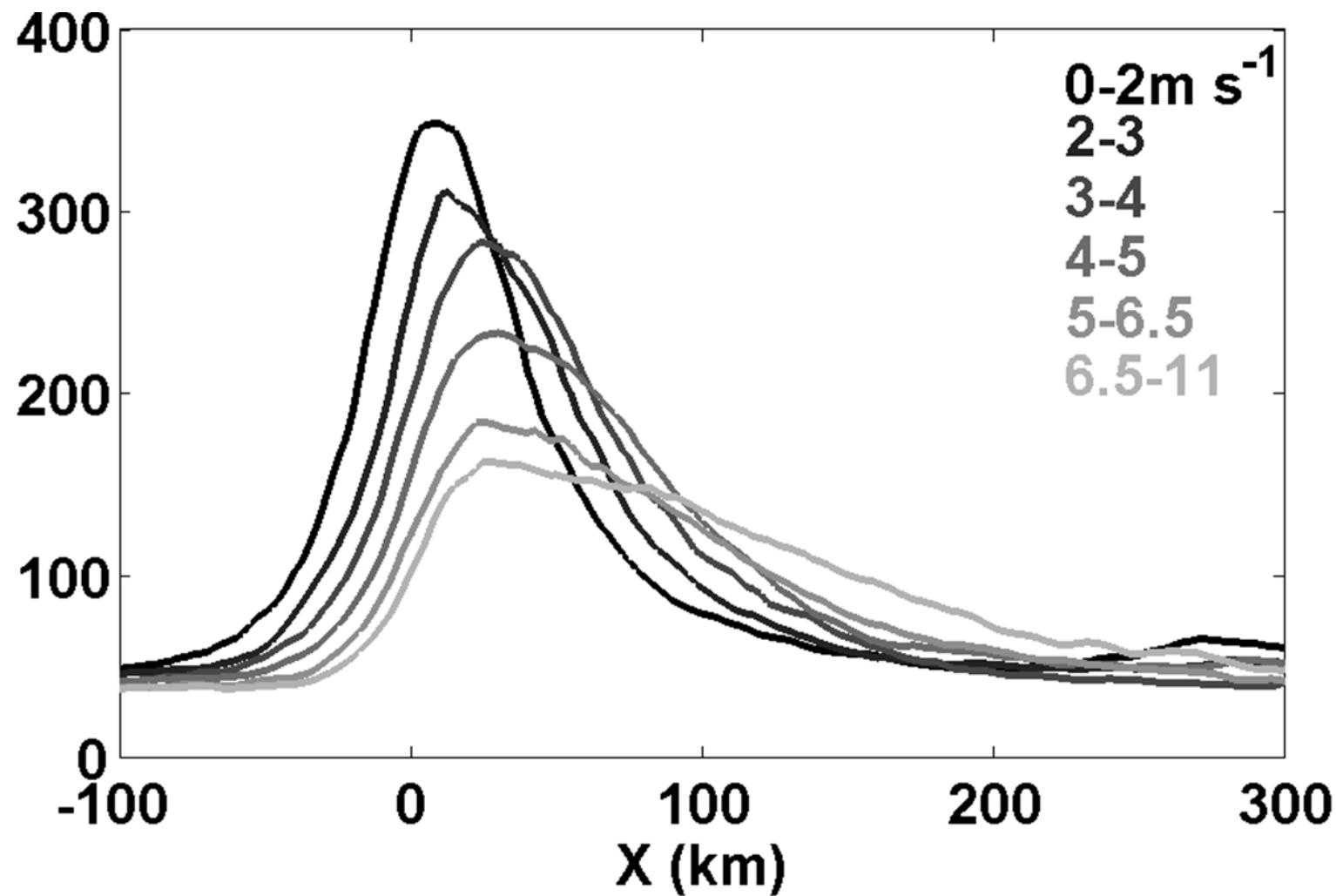
Daytime

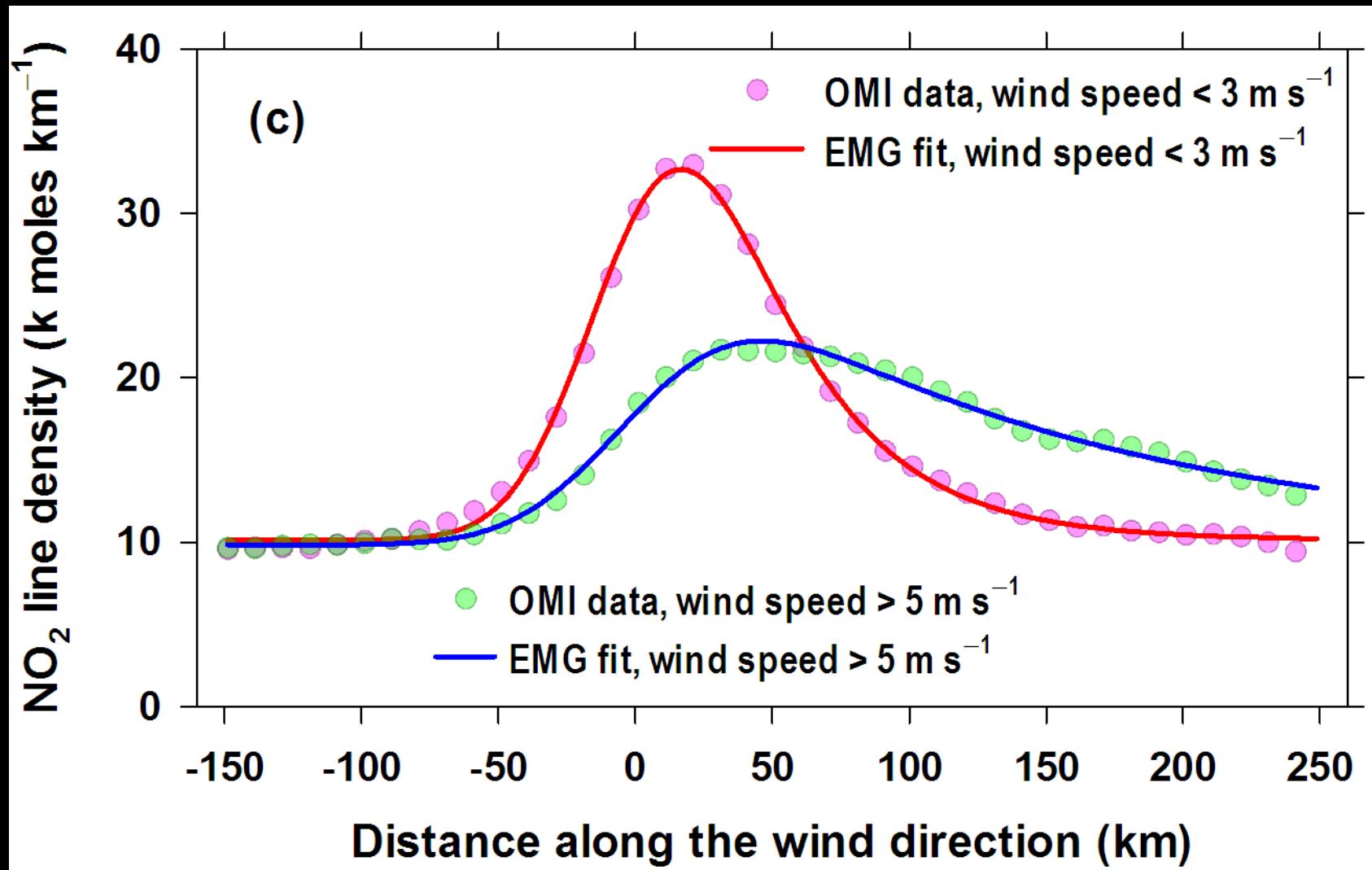


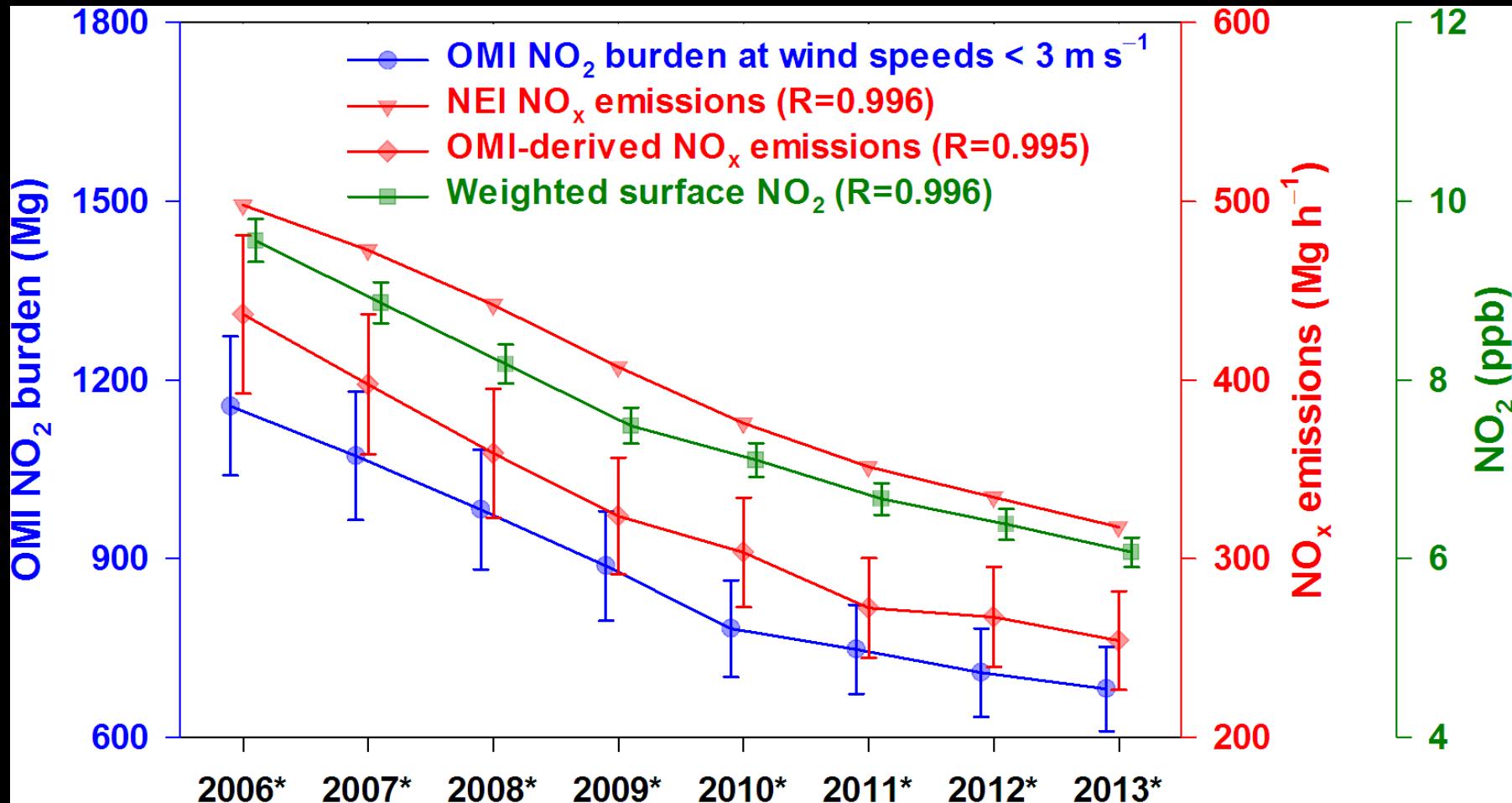
Model NO_x lifetime vs. wind speed.



Observations vs. wind speed.







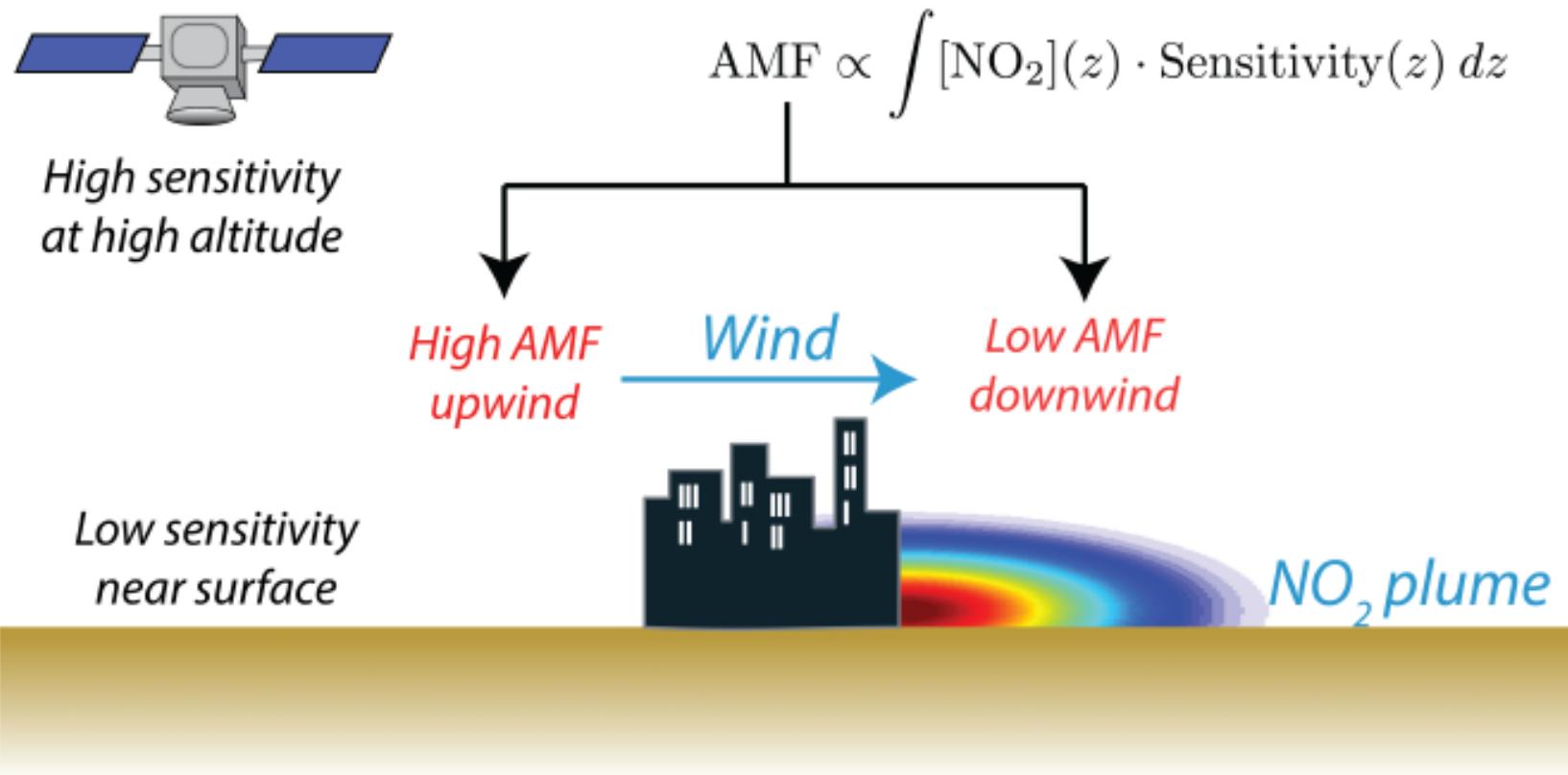
Retrievals



Josh Laughner



AMFs convert slant columns to vertical columns

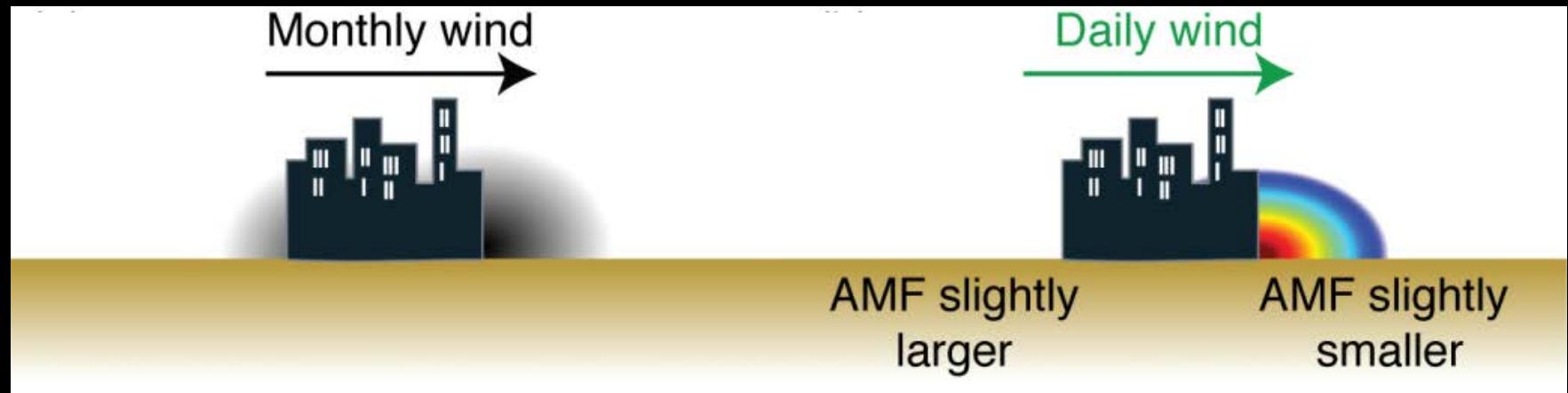


$$AMF = \frac{SCD}{VCD} \Rightarrow VCD = \frac{SCD}{AMF}$$

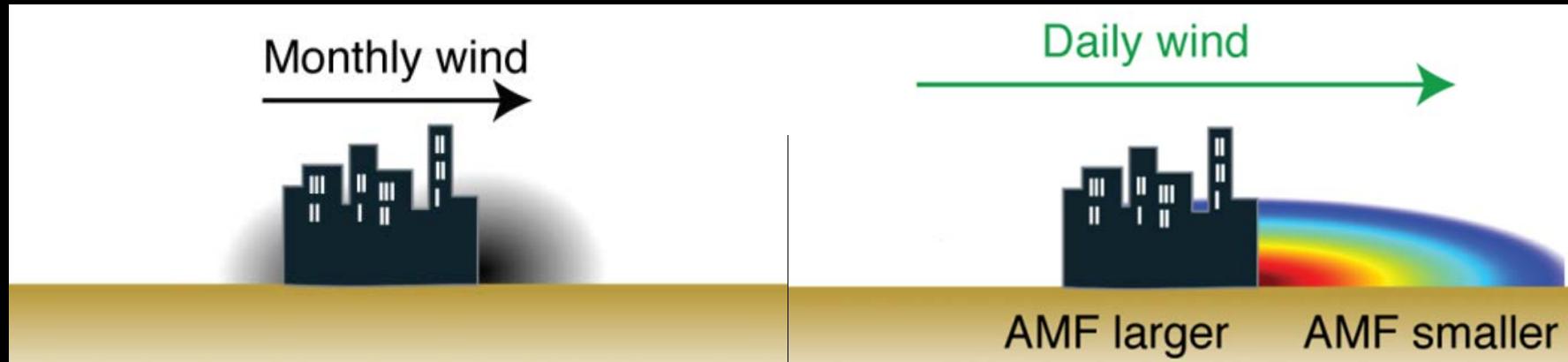
Winds and the a priori



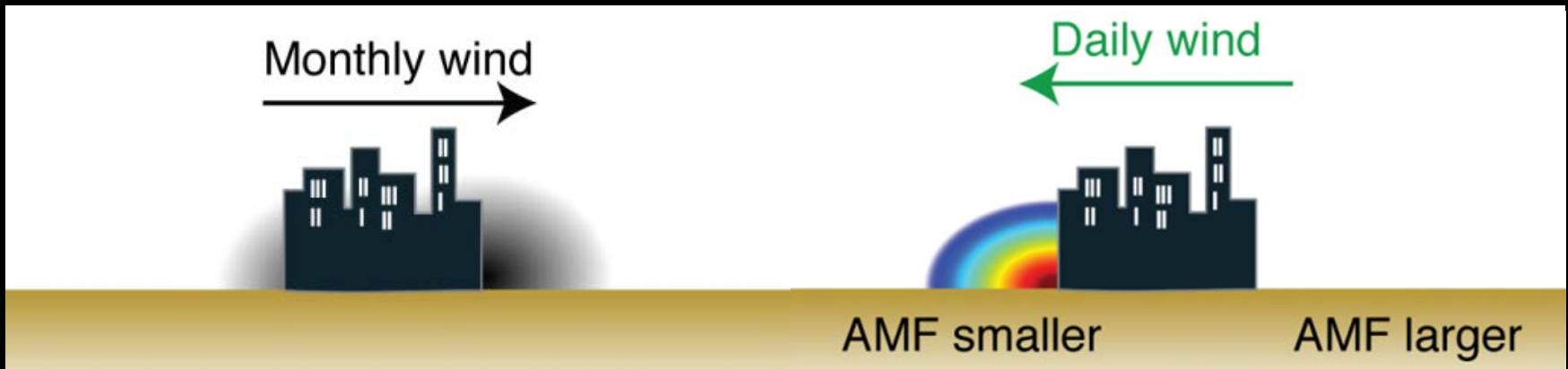
Monthly Average a priori and daily wind similar speed and direction



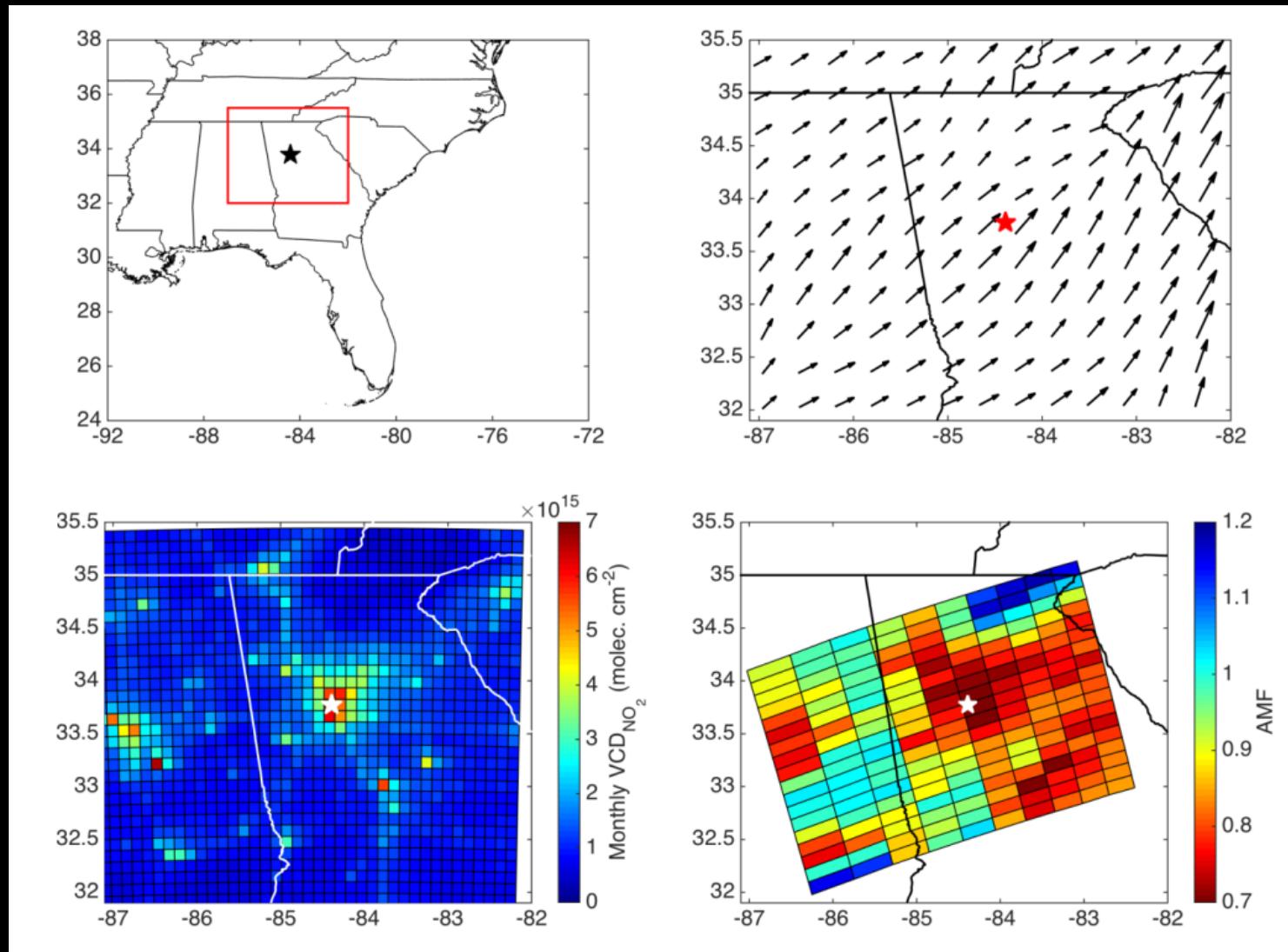
Monthly Average a priori with daily wind faster



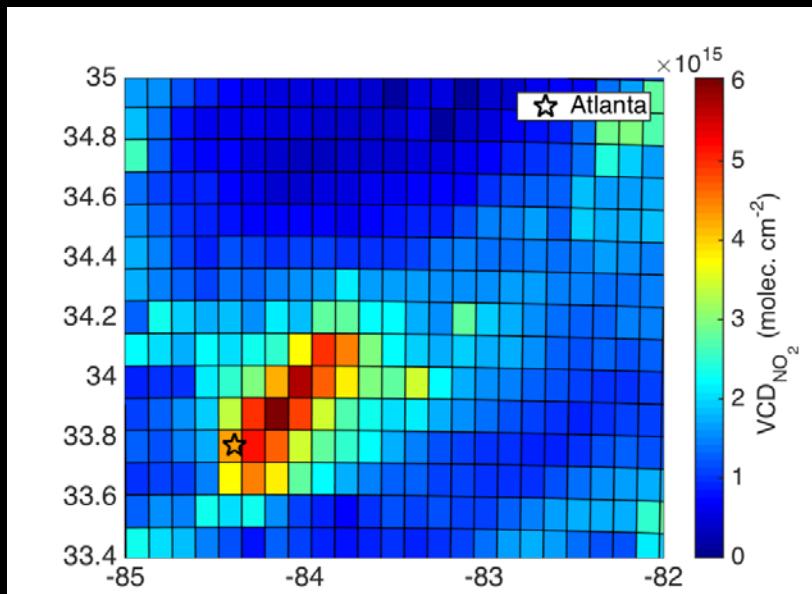
Monthly Average a priori and daily
wind same speed and opposite
direction



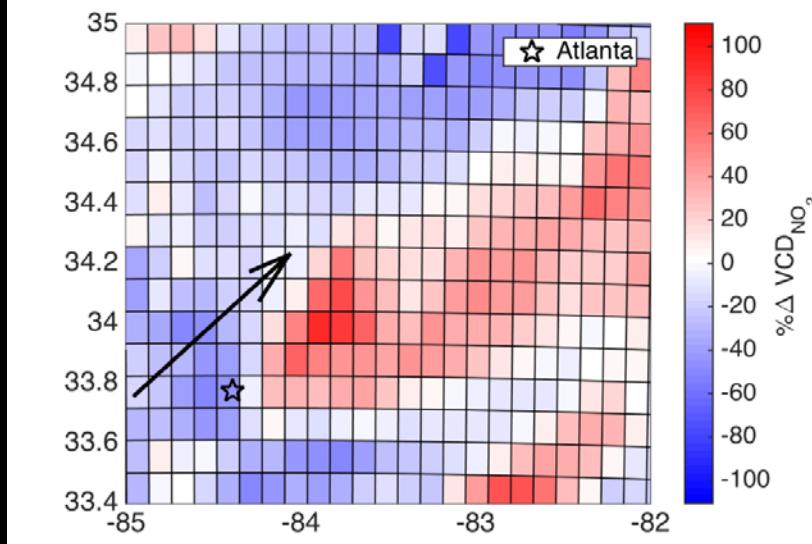
Theory says this should matter. Does it? Case study of Atlanta



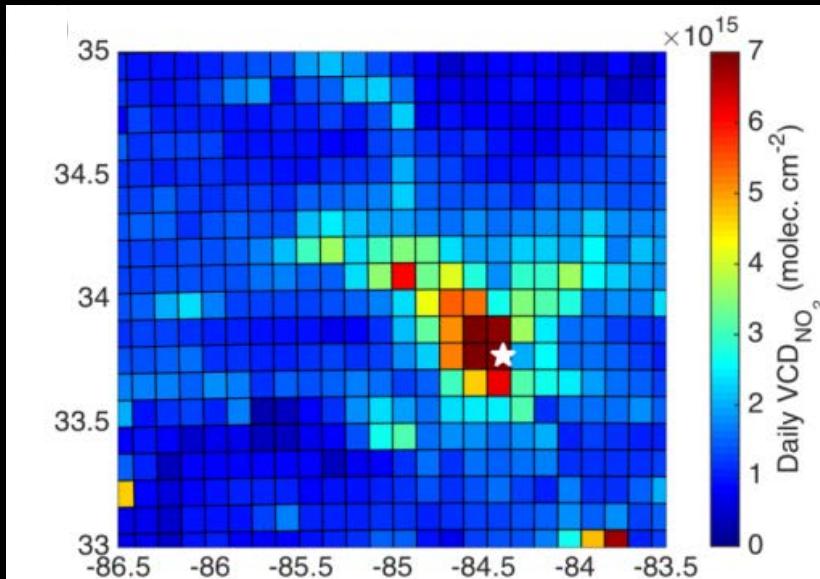
NO_2 VCD



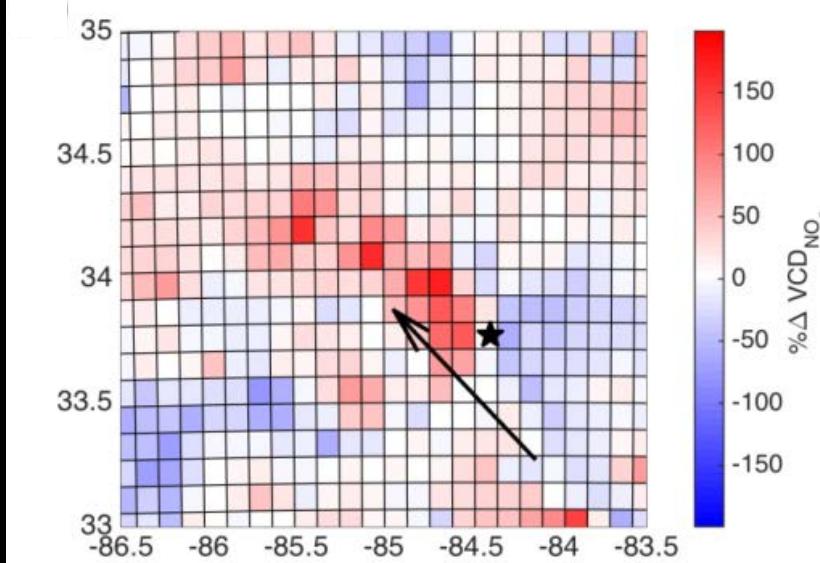
% Change
daily a priori



NO_2 VCD

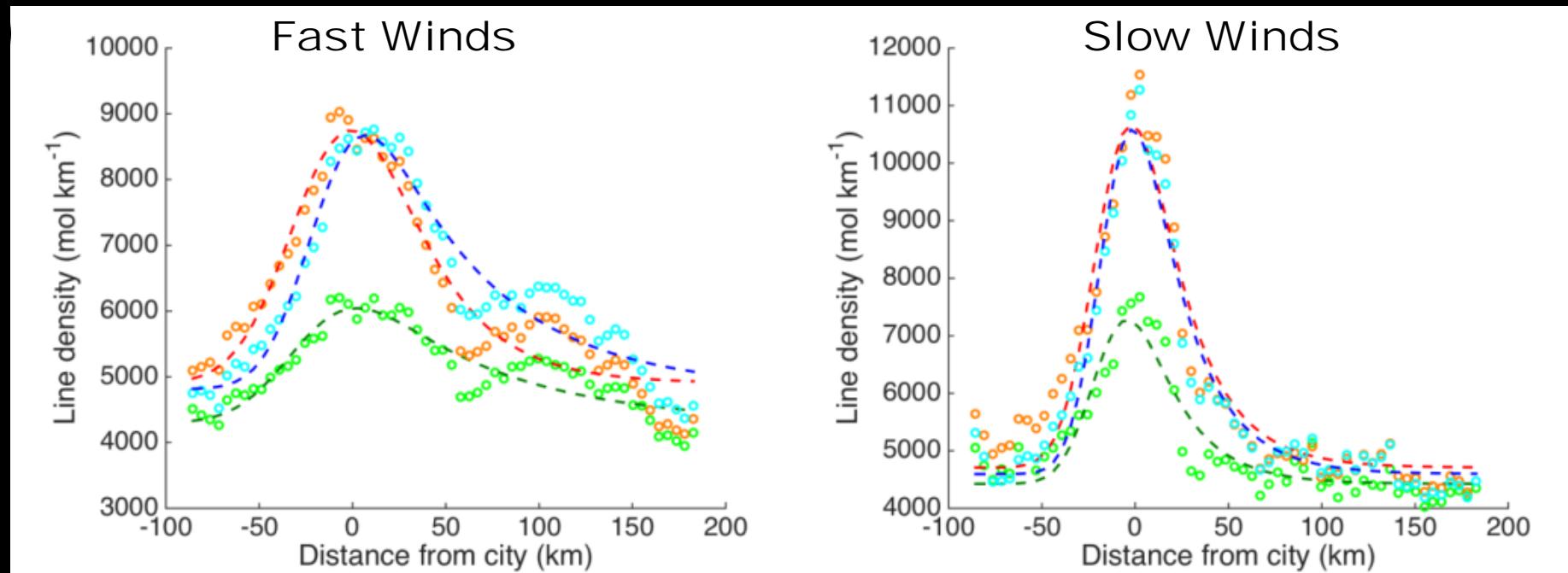


% Change
daily a priori



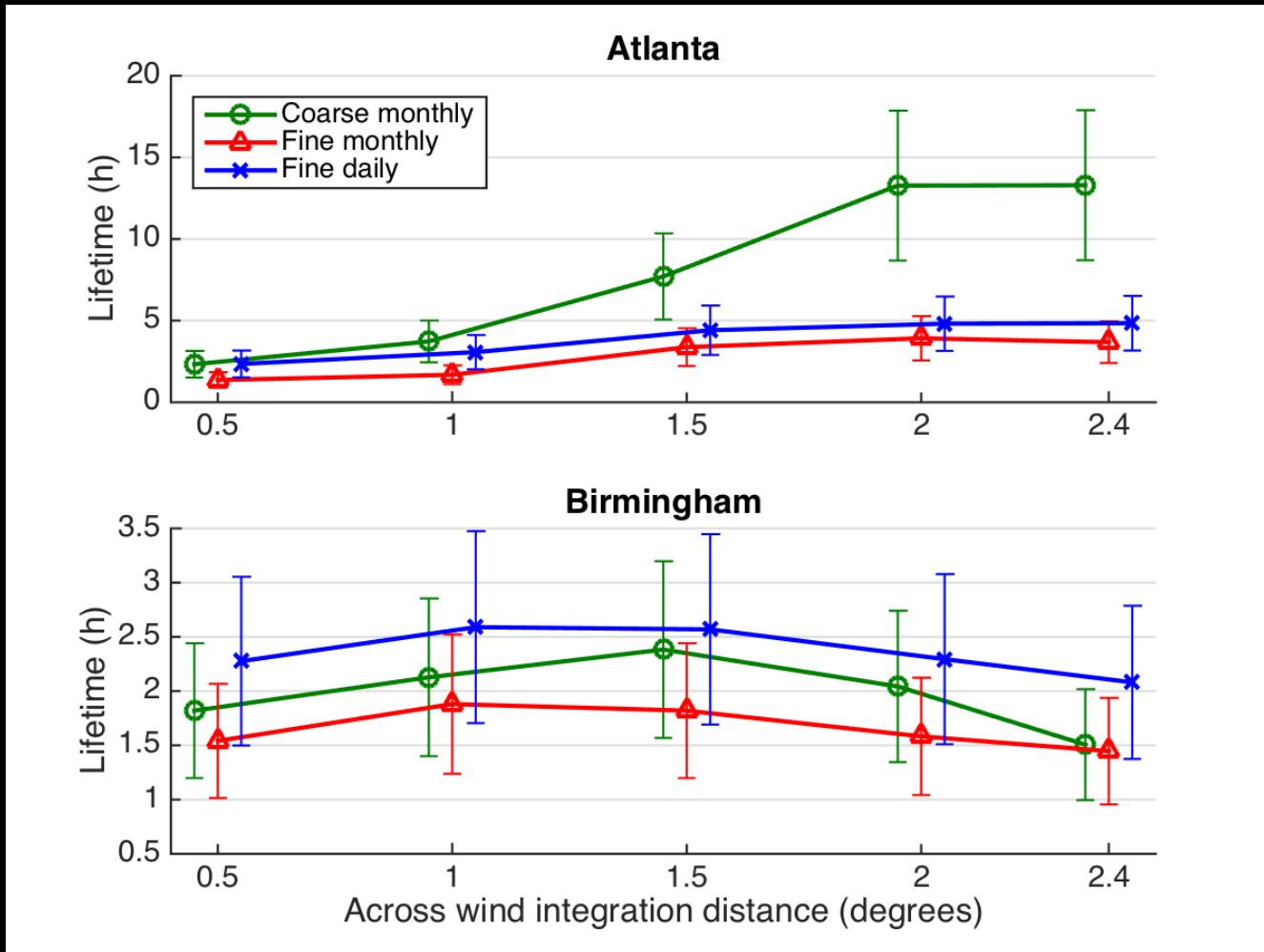
Fits to monthly average plumes

- Green: monthly a priori, low res
- Orange: monthly a priori, high res
- Blue: daily a priori, high res



Derived Lifetime and a priori

Blue: most near to truth

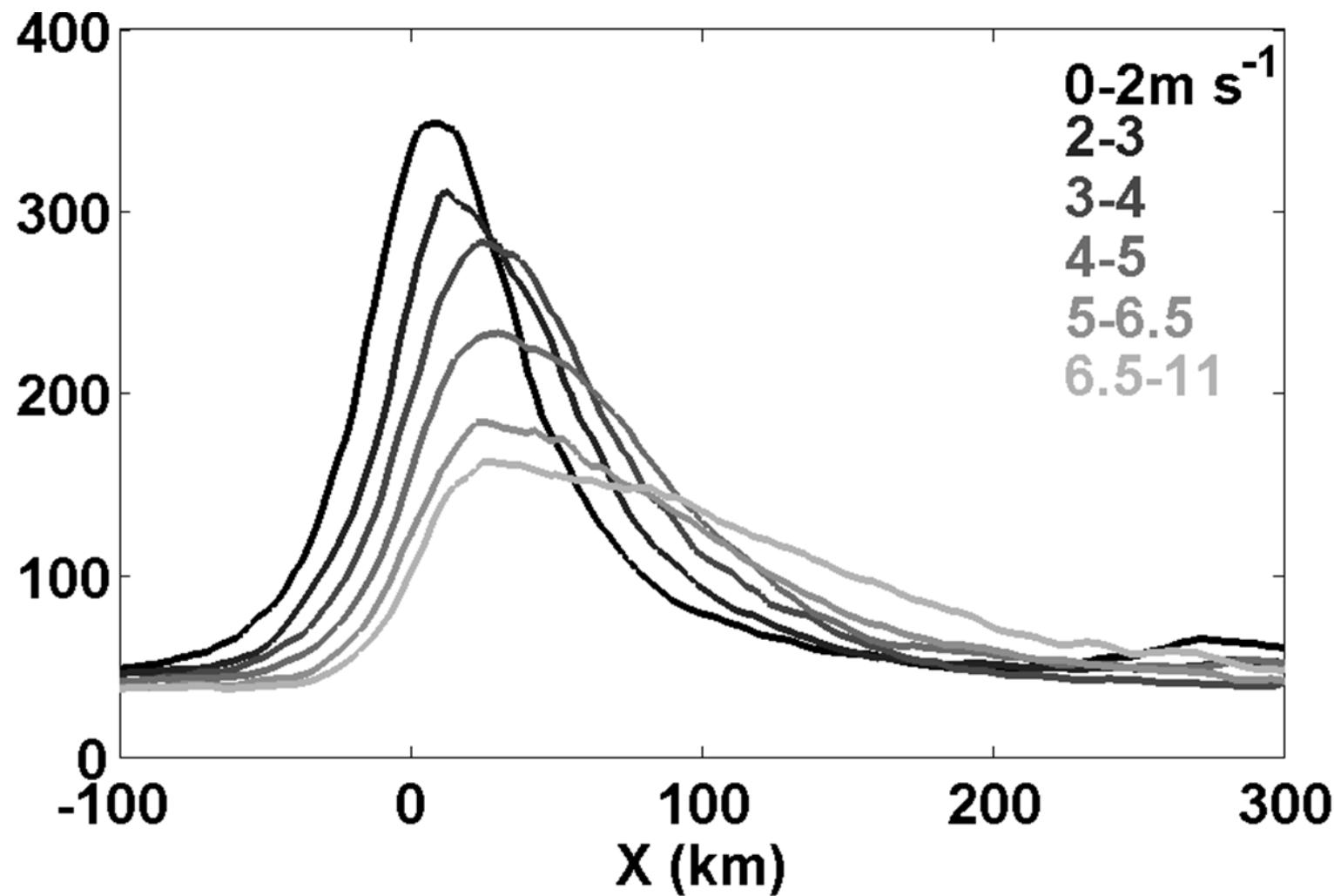


WRF-NEI Emissions ($\text{Mg NO}_x \text{ h}^{-1}$) 13.74

	Wind Speed	Monthly/108 km	Monthly/12km	Daily/12km
Emissions ($\text{Mg NO}_x \text{ h}^{-1}$)	>3m/s	3.39	11.66	7.55
	>5m/s	7.8	13.41	12.13
Lifetime (h)	>3m/s	3.73	1.68	3.07
	>5m/s	2.44	2.18	2.48

Variation in results arises from combining wind speeds. Emissions are almost certainly constant but columns vary with wind speed Substantially in narrow windows of speed.

Observations vs. wind speed.



Summary retrievals:

Emissions and lifetimes to even ~20% accuracy from remote sensing of NO₂ will require:

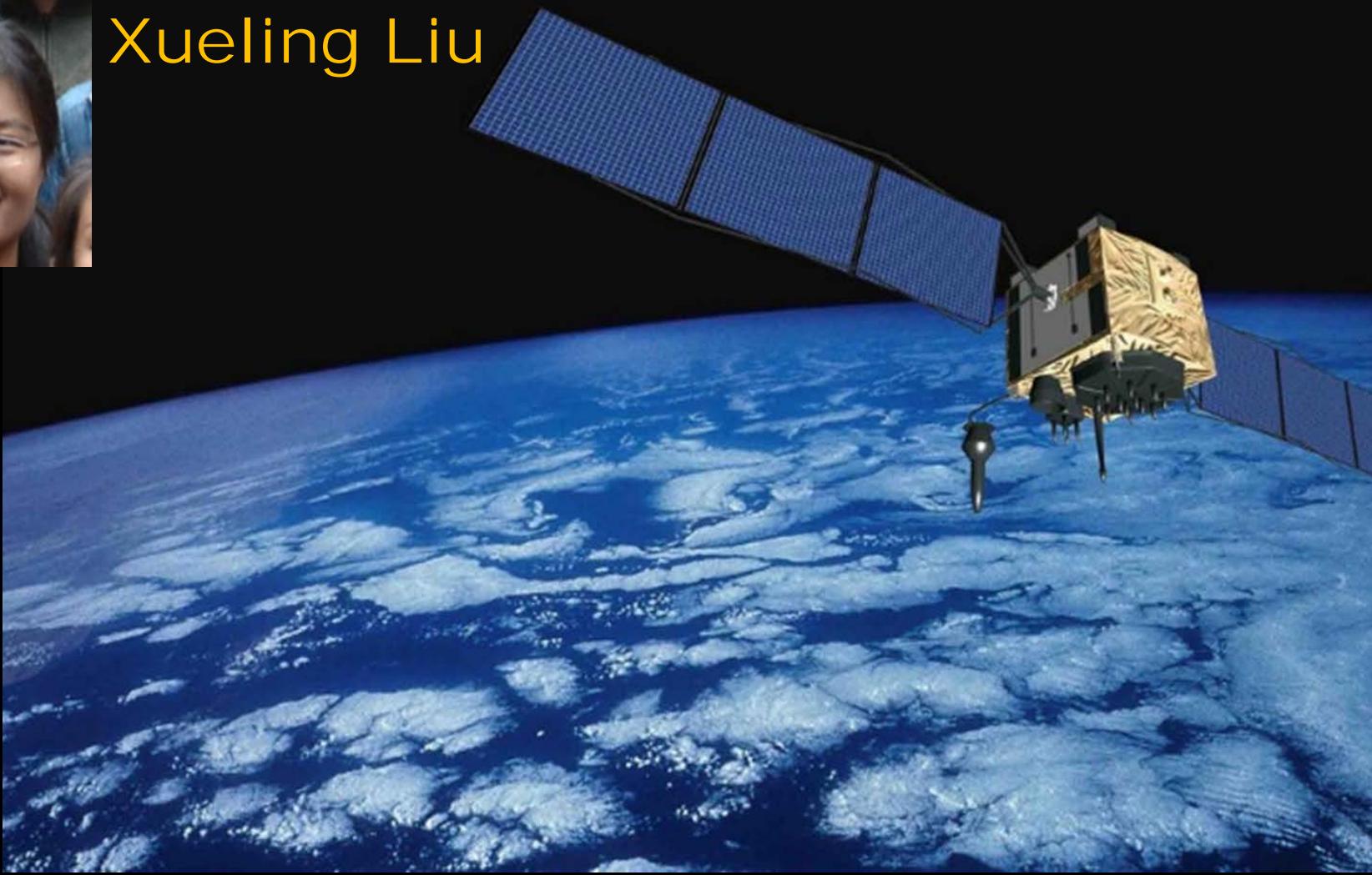
high resolution *a priori* using daily wind fields

And new strategies for analysis that allow use of narrow wind ranges

Emissions and Winds from Data Assimilation



Xueling Liu



Ensemble Kalman Filter-Data Assimilation



Xueling Liu

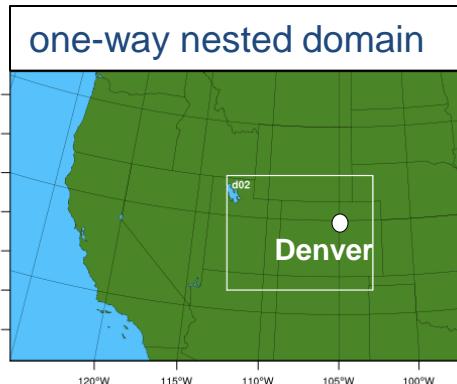
Also: A. P. Mizzi, J. L. Anderson, NCAR
I. Fung, UC Berkeley

Data Assimilation of NO₂ column observations

Xueling Liu, RC Cohen, A. Mizzi, I. Fung, J. Anderson

- Forecast model: WRF-Chem CTM
 - d01 (coarse domain) dx=12km, Grids=220*140 : Met Initialization and lateral boundary conditions from NARR.
 - d02 (fine domain) dx=3km, Grids=281*221: Both met and chemistry initialized and forced by d01.

- Anthr: NEI 2011
- Bio: MEGAN
- Met input: NARR
- PBL scheme:YSU
- Chem mechanism: RADM2



• Observation Simulator for TEMPO

Frequency: daytime hourly; resolution: 2×4.5km²

- Generation of scene-dependent AKs

Assuming cloud-free scenes, calculate AK based on the parameters--terrain pressure, albedo, solar zenith angle, view zenith angle and relative azimuth angle.

- Capture the spatial and temporal variation of AKs.

- Data Assimilation Engine: NCAR Data Assimilation Research Testbed
 - Ensemble Adjustment Kalman Filter
 - Updated chem variables: NO_x Emission, NO, NO₂, HNO₃, and O₃

$$\mathbf{x}_i^b = M\mathbf{x}_{i-1}^a$$

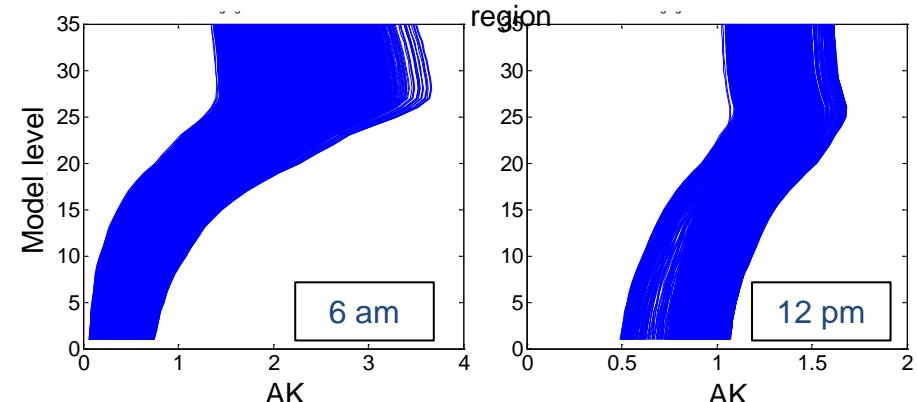
$$\overline{\mathbf{x}}_i^b = \overline{\mathbf{x}}_i^b + \mathbf{K}_i(\mathbf{y}_i^o - H\overline{\mathbf{x}}_i^b)$$

$$\mathbf{K}_i = \mathbf{P}_i^b \mathbf{H}^T [\mathbf{H}\mathbf{P}_i^b \mathbf{H}^T + \mathbf{R}]^{-1}$$

$$\mathbf{X}_i^a = \mathbf{T}_i \mathbf{X}_i^b$$

$$\mathbf{x}_i^a = \overline{\mathbf{x}}_i^a + \mathbf{X}^a$$

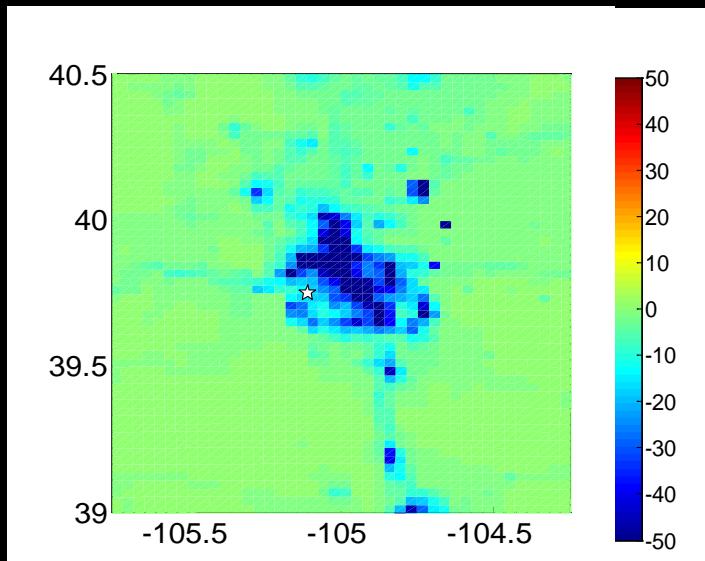
Averaging Kernel Profiles at 2230 locations in Denver region



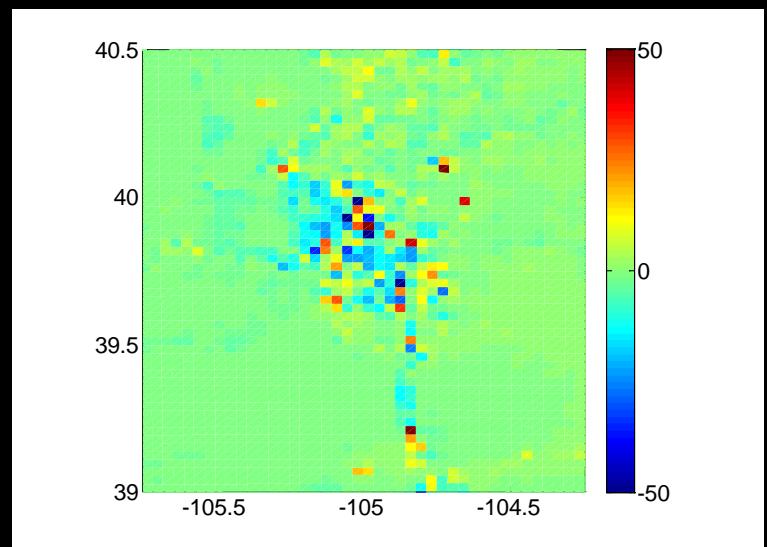
Data Assimilation—another talk

Data assimilation will be effective for constraining emissions—possibly on hourly/daily timescales

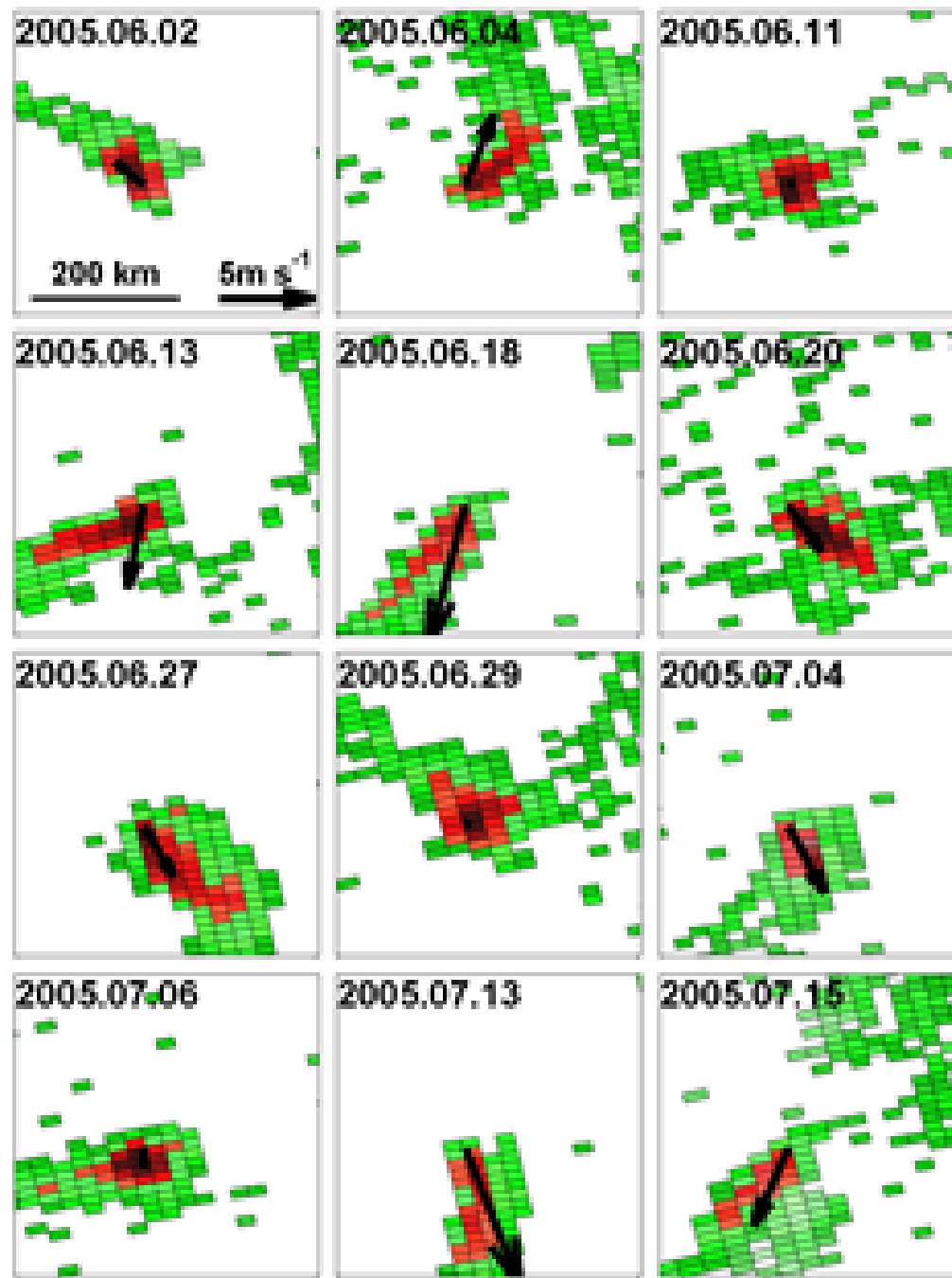
Emissions: Prior – truth
(mol/km²/hr)

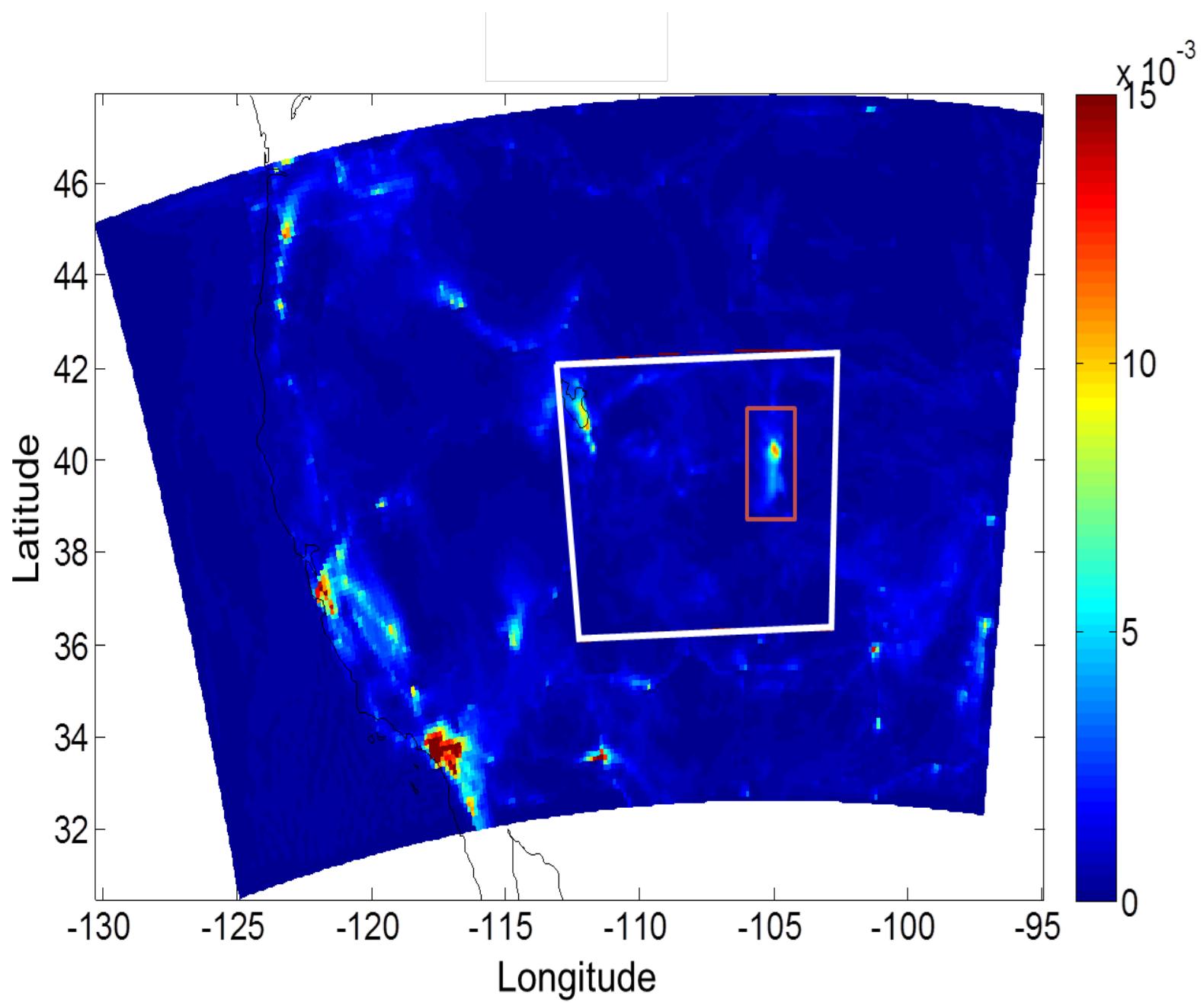


Emissions: Posterior – truth
(mol/km²/hr)

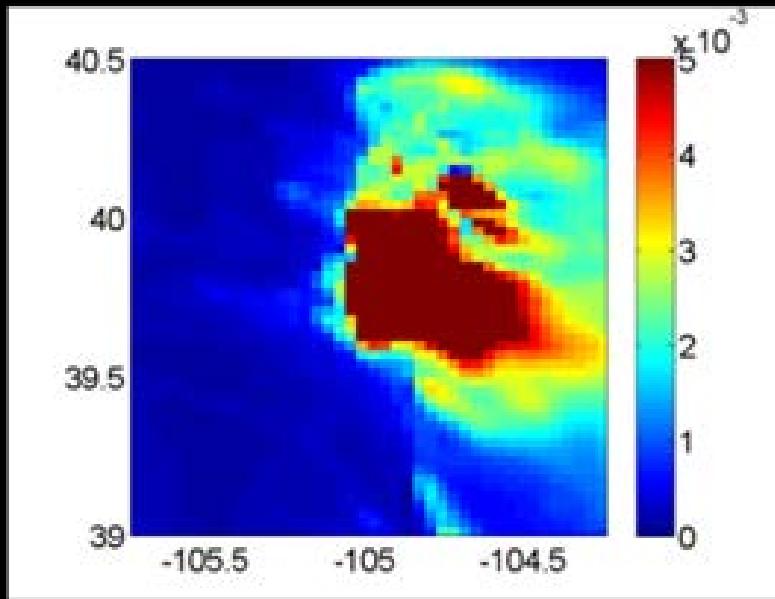


Riyadh

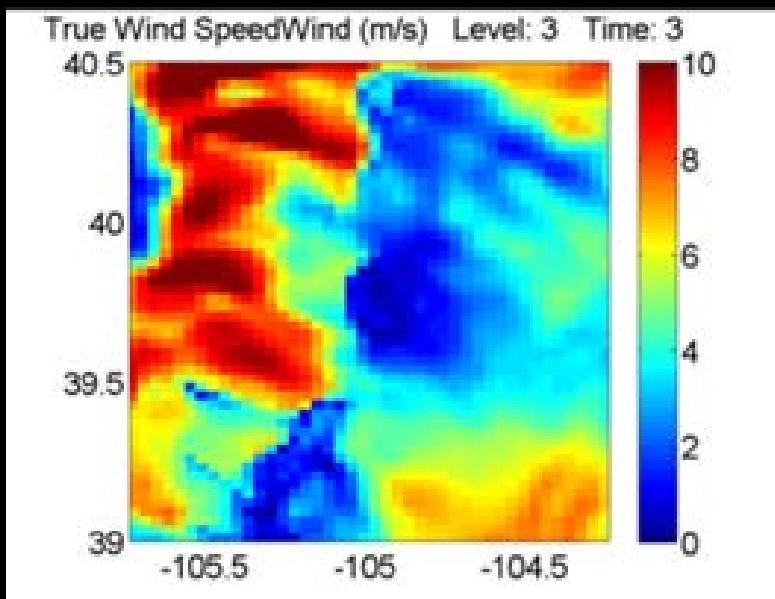




Truth (Denver)

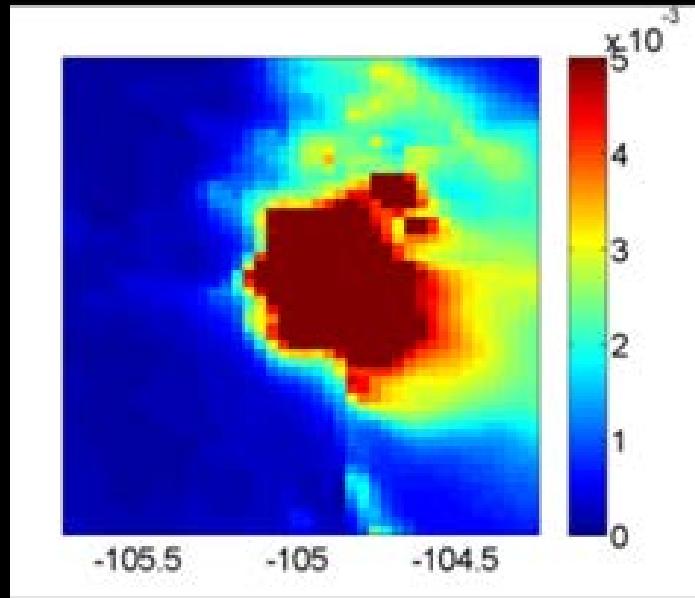


NO₂ Column

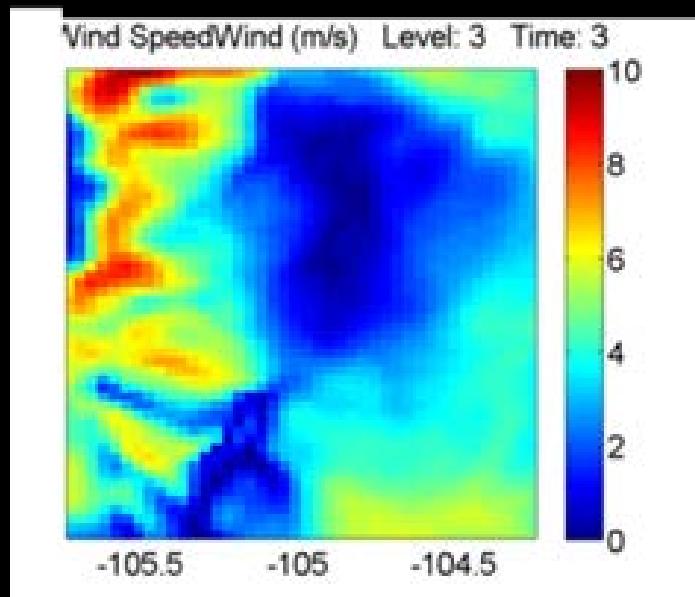


Winds

Prior with biased winds

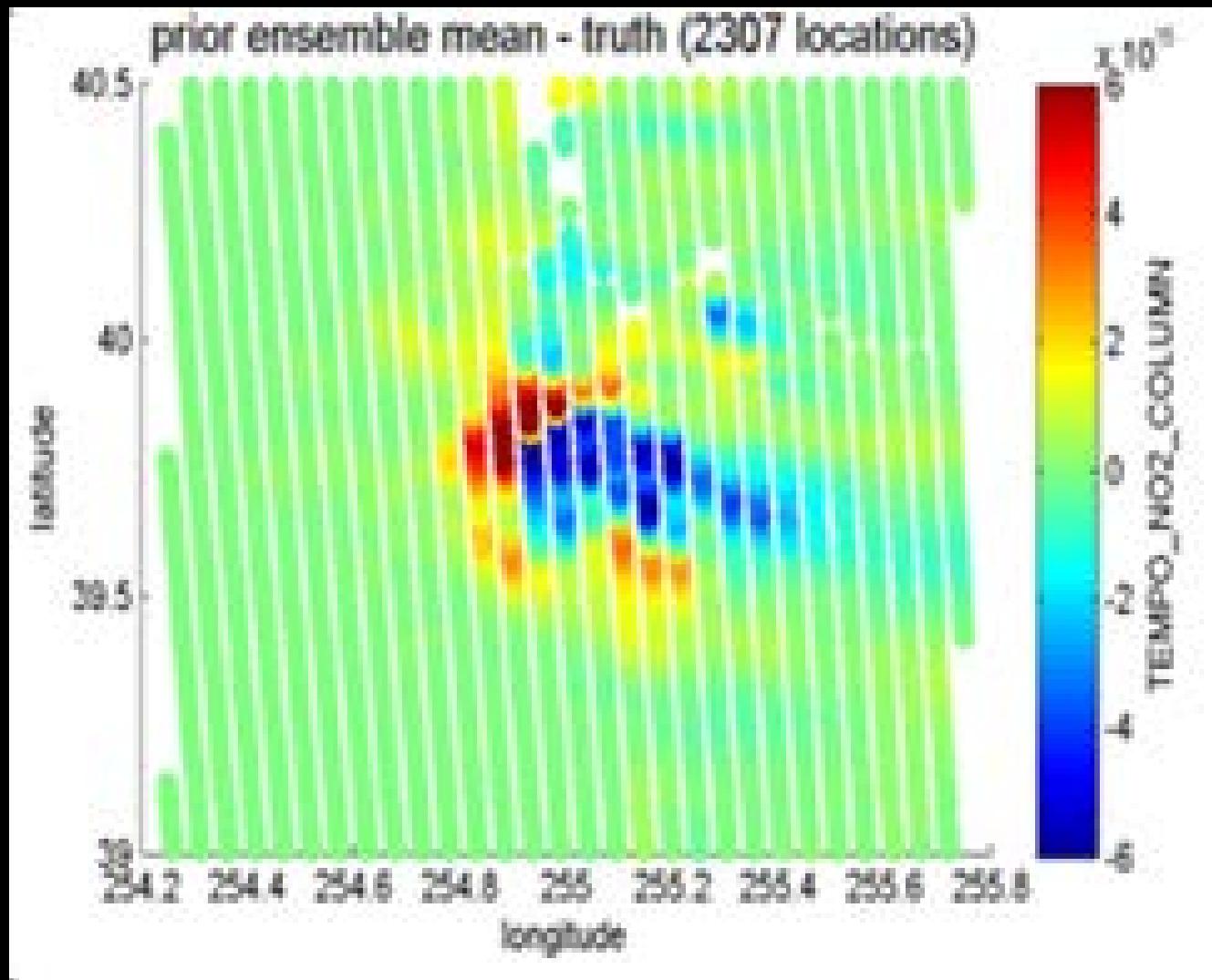


NO₂ Column

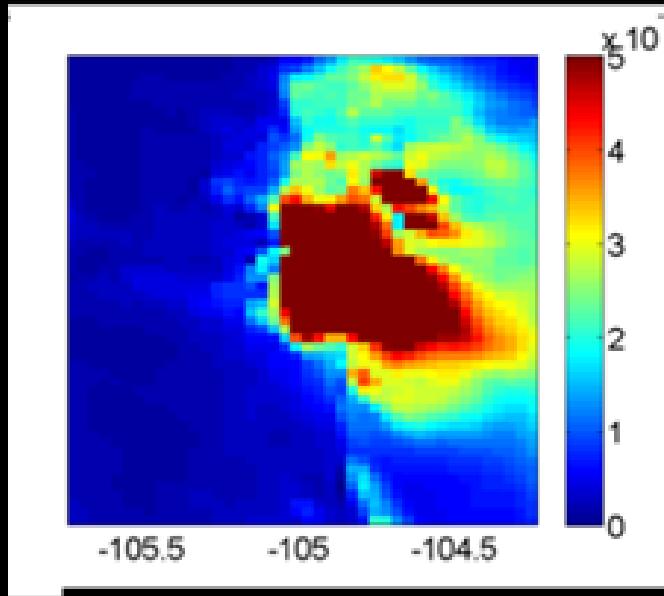


Winds

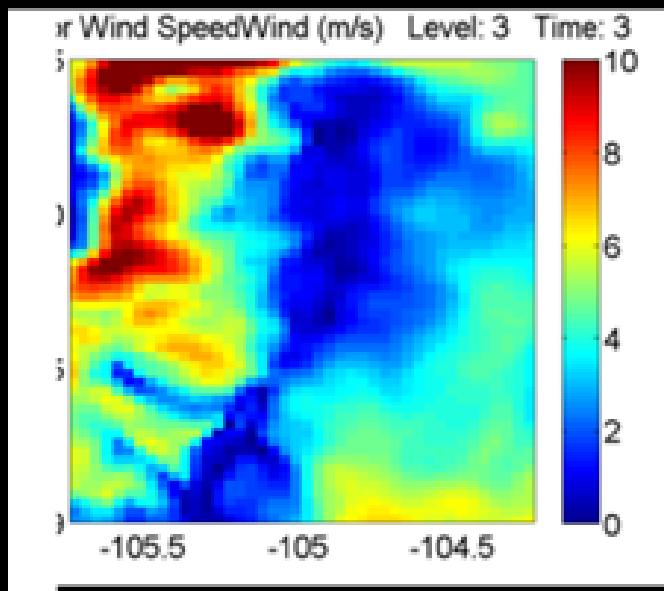
Truth - Prior (Denver) Sampled with TEMPO



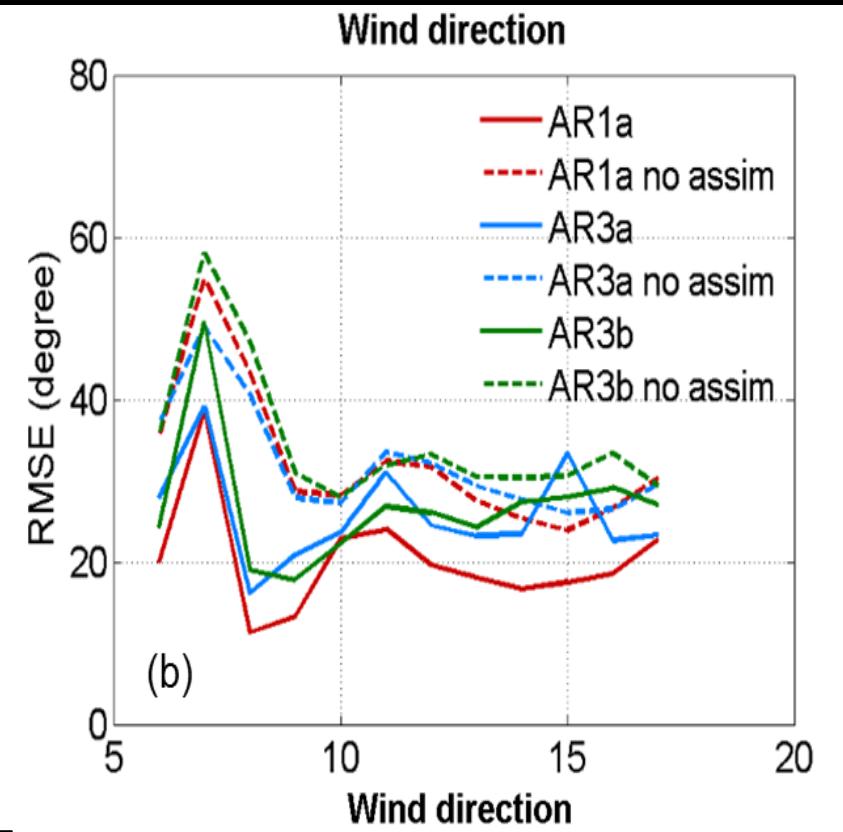
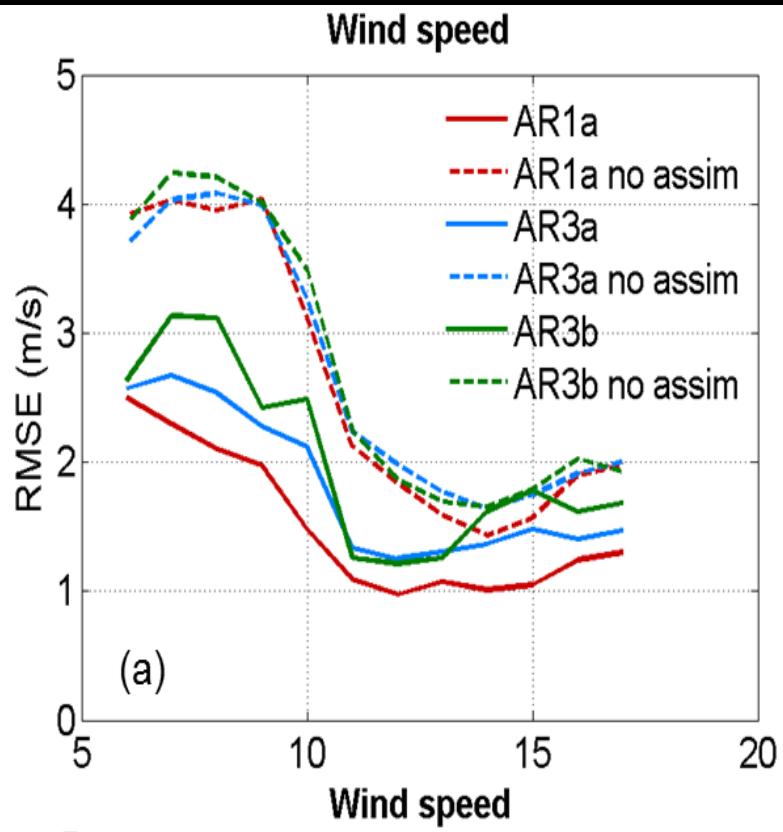
Posterior After Assimilation



NO₂ Column

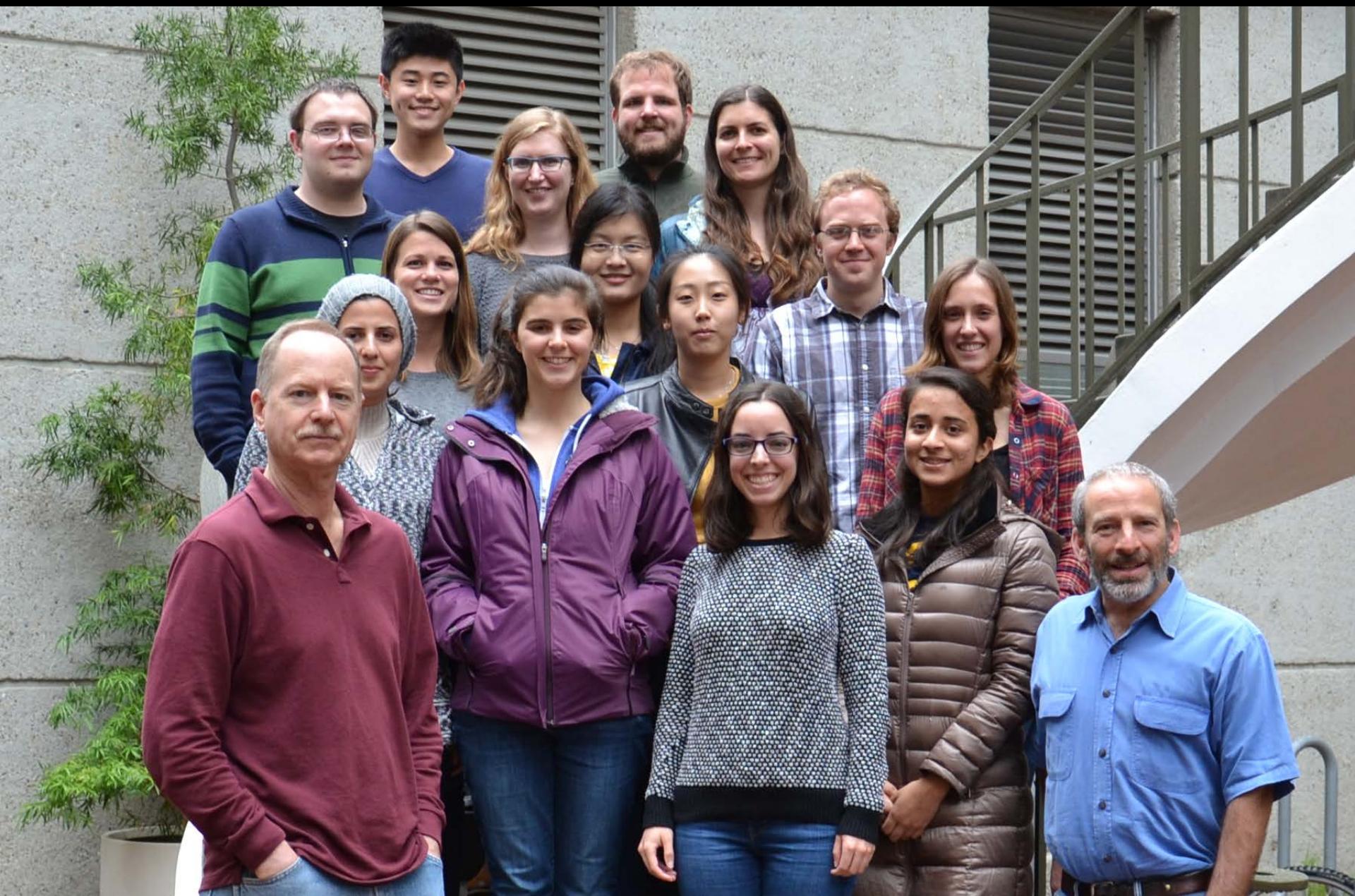


Winds



Winds and TEMPO

- 1 *Winds affect the NO₂ lifetime in interesting and important ways*
- 2 *Retrievals that account for daily wind variations are needed for quantitative accuracy in our inferences of emissions and lifetime*
- 3 *Data assimilation will be effective for constraining emissions—possibly on hourly/daily timescales*
- 4 *Data assimilation will also be able to constrain meteorological fields e.g. winds*



Thank you!

