

Relationships and Trends among Satellite NO_2 Columns and NO_x Emissions

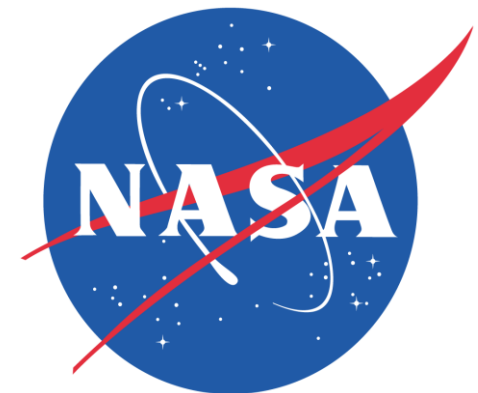
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Argonne National Laboratory

Presented at 2014 Midwest and Central States Air Quality Workshop
St. Louis, MO
April 22-24, 2014

... with the support of NASA AQAST and acknowledgment of the contributions of AQAST colleagues working on emissions:

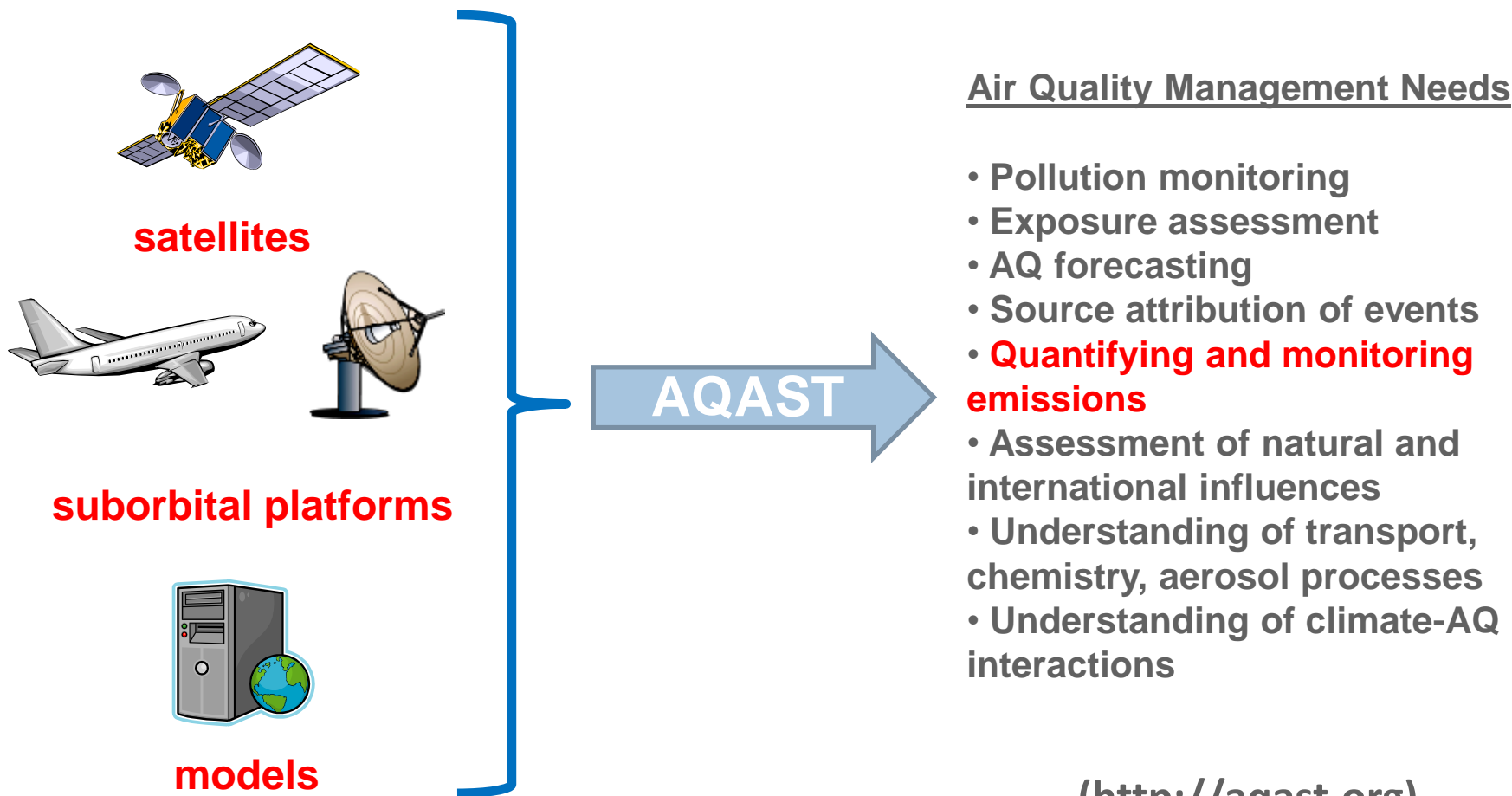
Greg Carmichael, U Iowa; Dan Cohan, Rice U; Ben de Foy, Saint Louis U; Bryan Duncan, NASA/GSFC; Arlene Fiore, Columbia U; Tracey Holloway, U Wisconsin; Lok Lamsal, NASA/GSFC; Can Li, NASA/GSFC

Daniel Jacob, Harvard University, AQAST team leader
John Haynes, NASA program manager



NASA Air Quality Applied Sciences Team (AQAST)

The AQAST goal is to transfer Earth Science knowledge to serve the needs of U.S. air quality management with focus on the use of NASA satellites, suborbital platforms, and models



(<http://aqast.org>)

Objective of the emissions component: Assessment of the applicability of current worldwide studies of satellite retrievals and emissions estimation to U.S. air quality management

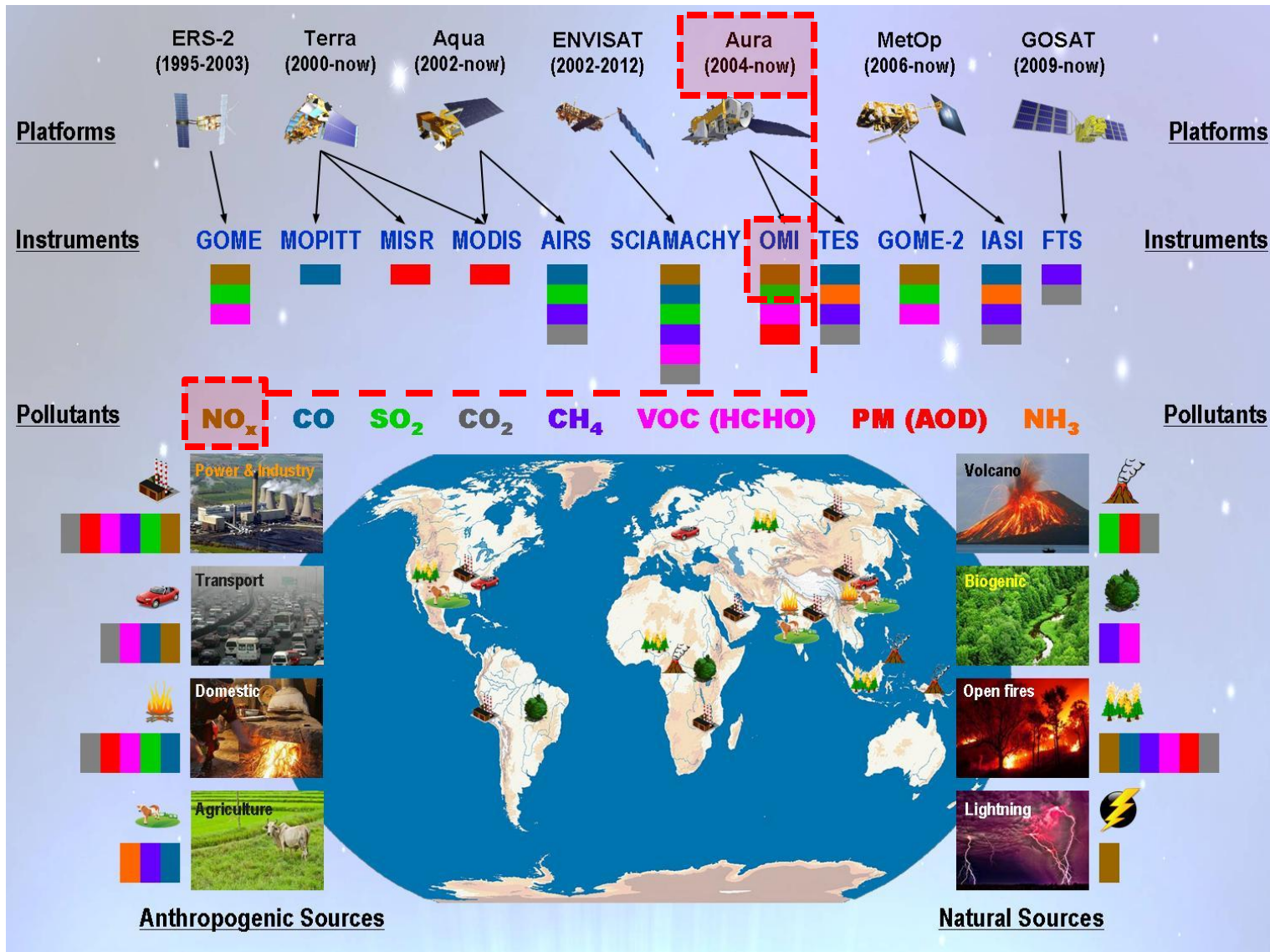
Issue: How can U.S. air quality managers make use of satellite retrievals to improve emission estimates, and what developments are needed to improve the usefulness of those retrievals?

Some potential applications in the U.S. (not exhaustive):

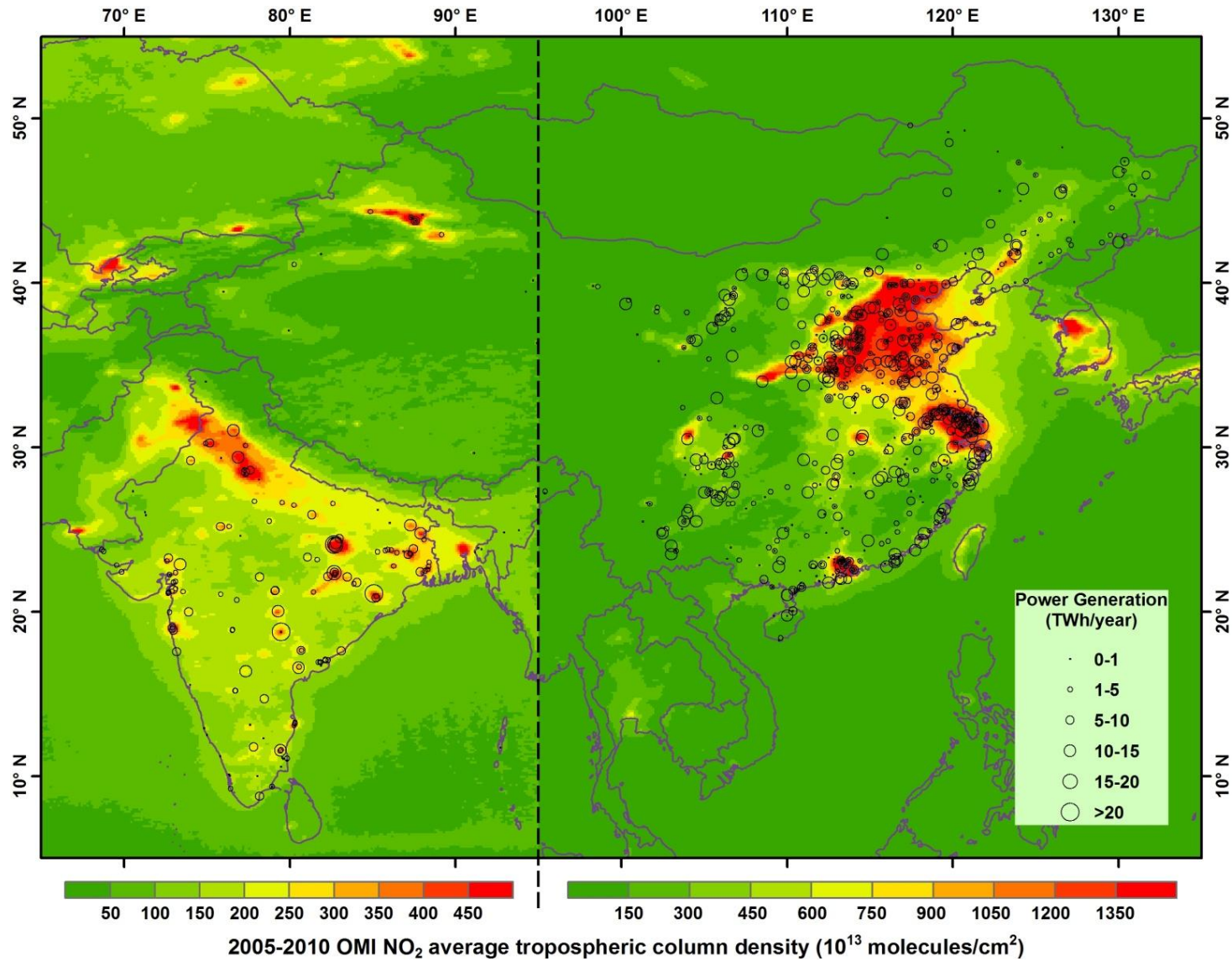
problematic industrial sources and industrial complexes, uncertain area sources (including biogenic), oil/gas extraction, verification of regional emission reductions and trends, quantification of atmospheric lifetimes, quantification of uncertain Mexican and Canadian emissions, etc., etc.



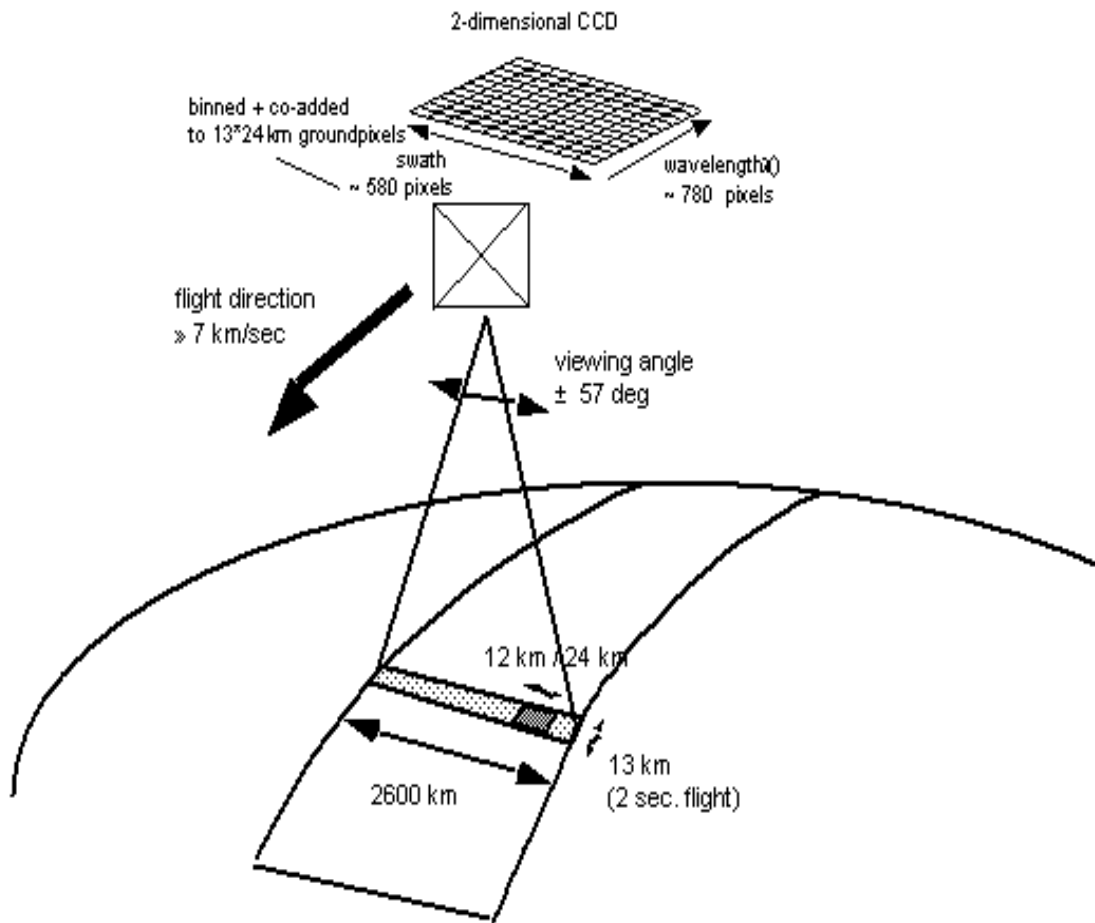
The complexity of satellite platforms, instruments, pollutants, sources, and world regions. How to process the information?



Initial studies of high-NO_x source regions in Asia were promising



There is the potential to quantify point sources of NO_x , SO_2 , etc., from OMI, if pollutant transport and chemical conversion cooperate



TEMPO
Tropospheric Emissions: Monitoring of Pollution

TEMPO's concurrent high temporal (hourly) and spatial resolution measurements from geostationary orbit of tropospheric ozone, aerosols, their precursors, and clouds create a revolutionary dataset that provides understanding and improves prediction of air quality and climate forcing in Greater North America.

Hourly Measurement of Pollution

60 minutes

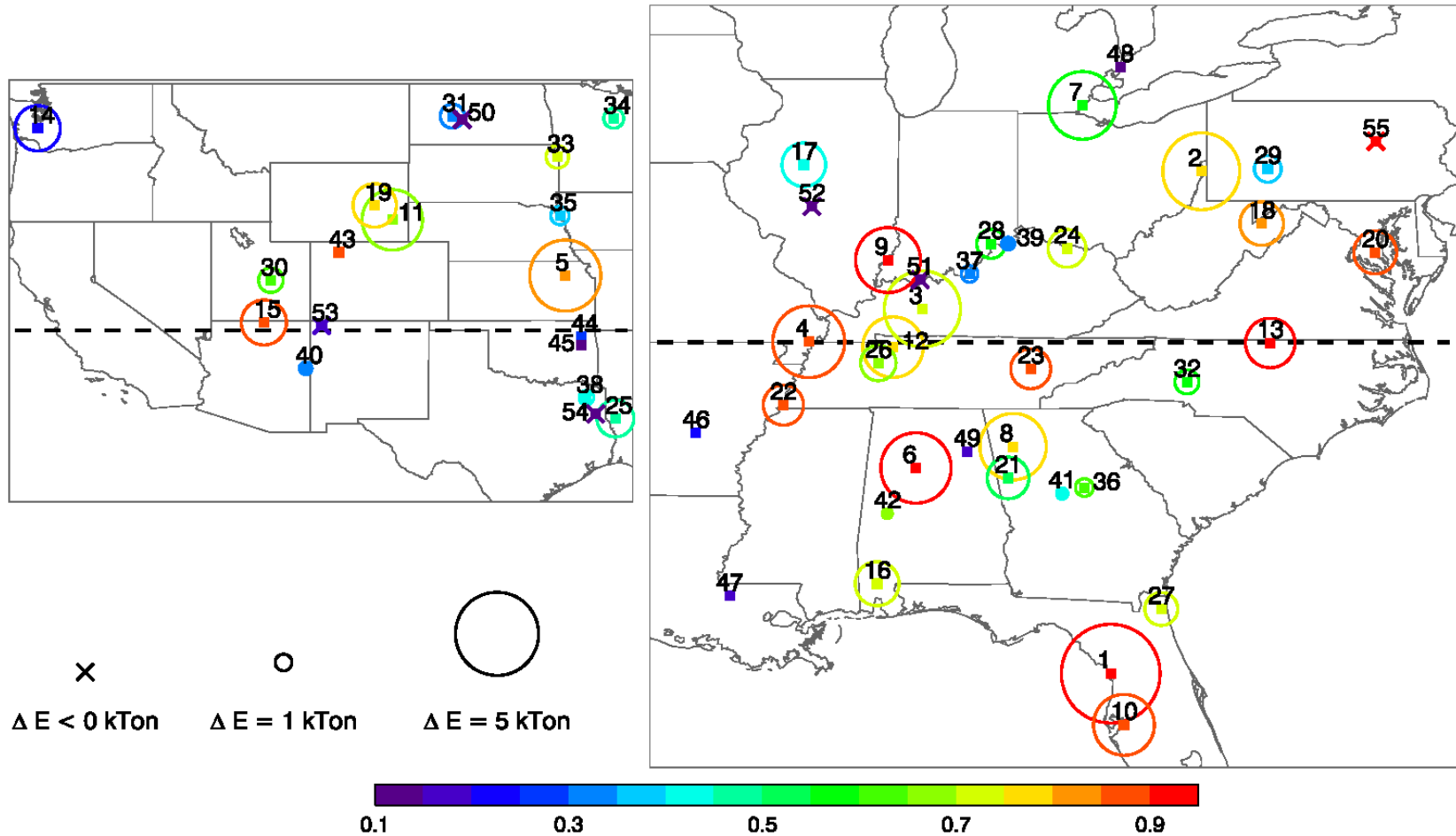
Houston, TX, June 2005

Hour of Day (Local Time)	TEMPO Data (ppb)	TEMPO Data (ppb)
00	10	10
01	10	10
02	10	10
03	10	10
04	10	10
05	10	10
06	10	10
07	10	10
08	10	10
09	10	10
10	10	10
11	10	10
12	10	10
13	10	10
14	10	10
15	10	10
16	10	10
17	10	10
18	10	10
19	10	10
20	10	10
21	10	10
22	10	10
23	10	10
24	10	10

Logos: SunWise, NASA, Blue Cross of Texas

Coming in ~2019: TEMPO, in geosynchronous orbit over North America at 2×4.5 km, 1-hour resolution!

NO_x emissions from U.S. power plants [Duncan et al., 2013]



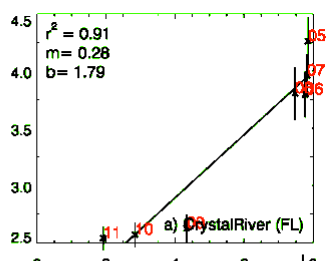
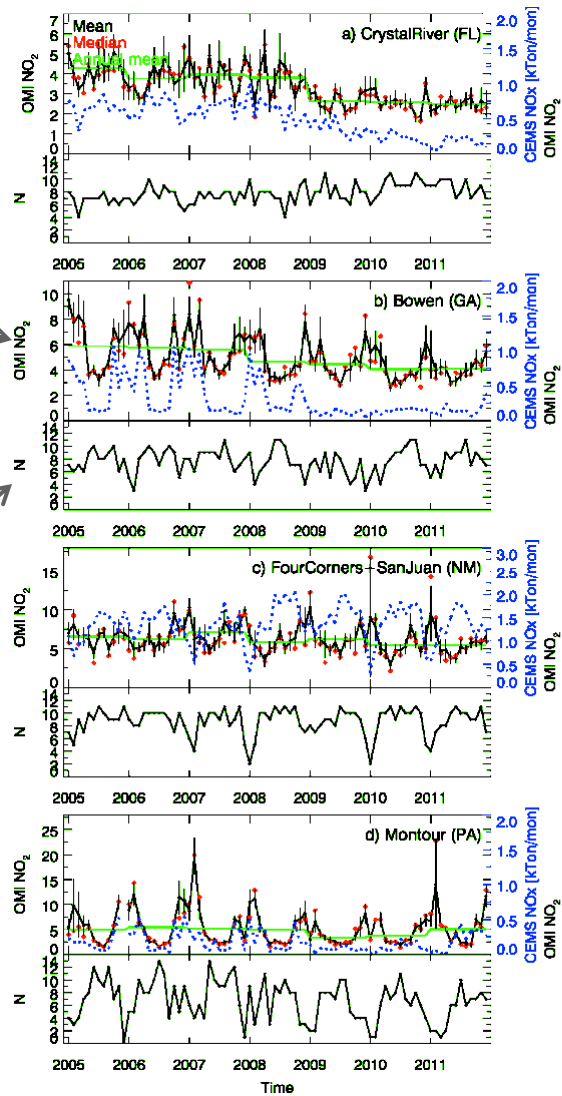
(size of circle represents change in emissions between 2005 and 2011; color of circle represents r^2 correlation between annual-average OMI and CEMS data)



Sample measurement data sets for four power plants in FL, GA, NM, and PA over a seven-year period, 2005-2011 [Duncan et al., 2013]

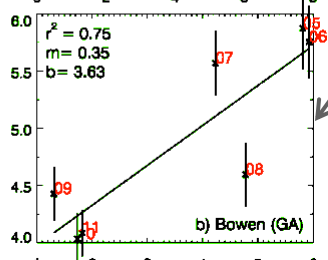
monthly OMI NO₂
 daily CEMS NO_x
 annual mean NO₂

sample size (days per month)

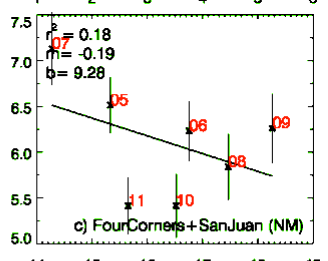


$r^2 = 0.91$

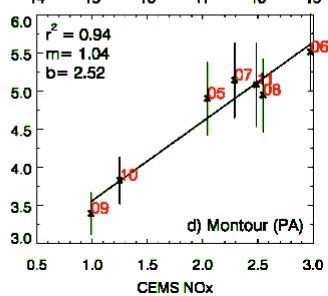
Annual mean OMI NO₂ vs annual CEMS NO_x



$r^2 = 0.75$



$r^2 = 0.18$

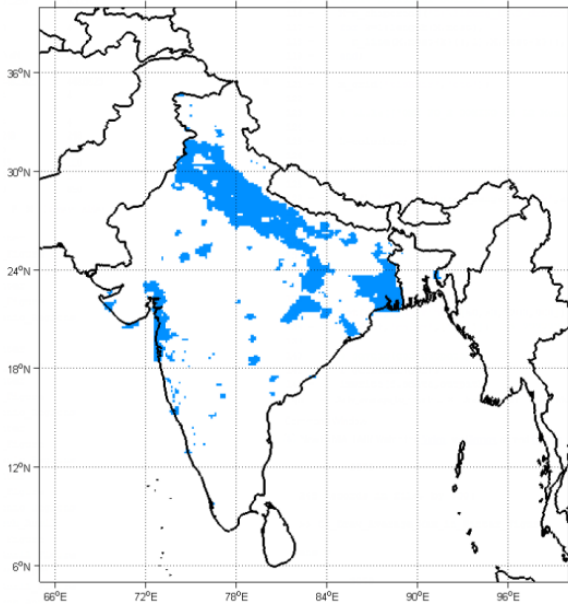


$r^2 = 0.94$

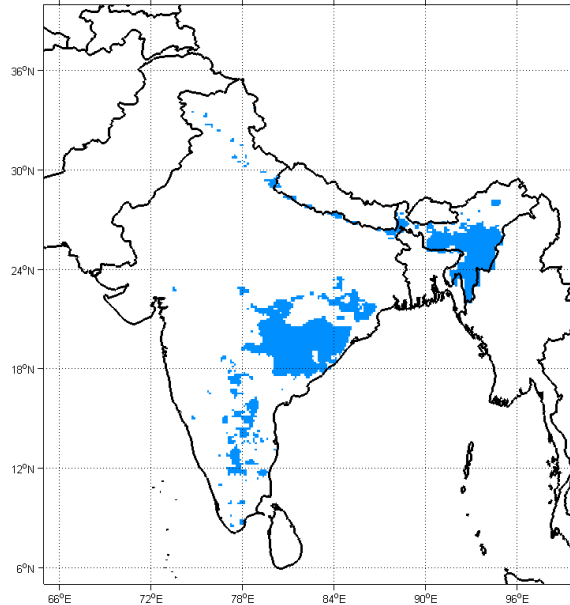
Identification of NO_x source regions using seasonality [Lu et al., 2013]

Example: OMI NO₂ (2005-2011)

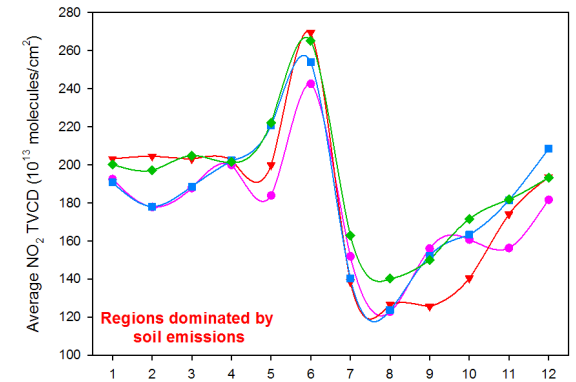
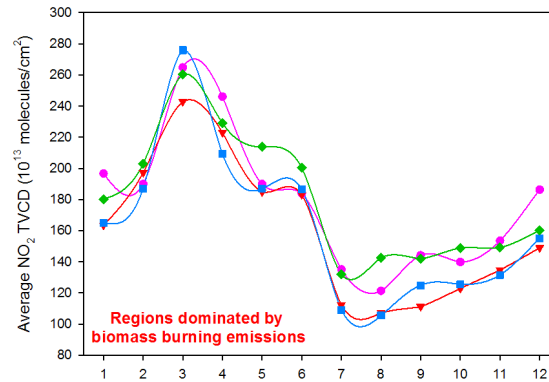
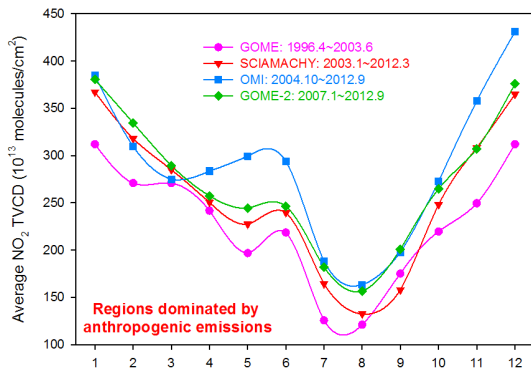
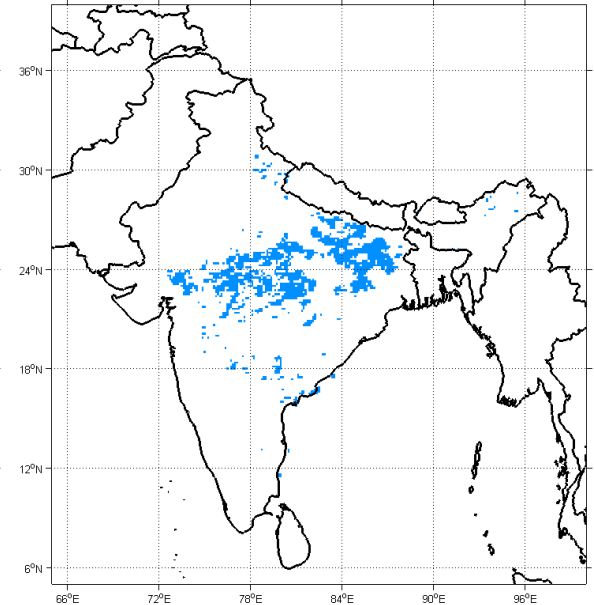
Anthropogenic emissions dominant (Dec/Jan peak)



Biomass burning emissions dominant (Mar/Apr peak)



Soil emissions dominant (Jun peak)



Application of satellite observations for timely updates to NO_x emission inventories (Randall Martin group)

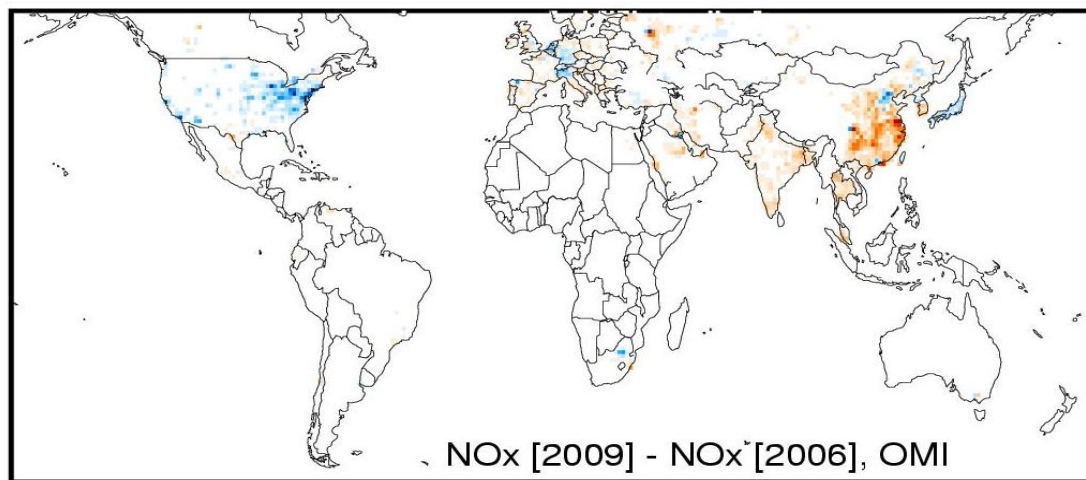
Use CTM to calculate local sensitivity of changes in trace-gas column to changes in emissions

Fractional Change
in Emissions

$$\Delta E = \beta \times \Delta \Omega$$

Fractional Change in
Trace-Gas Column

Local Sensitivity of Column Changes
to Emission Changes



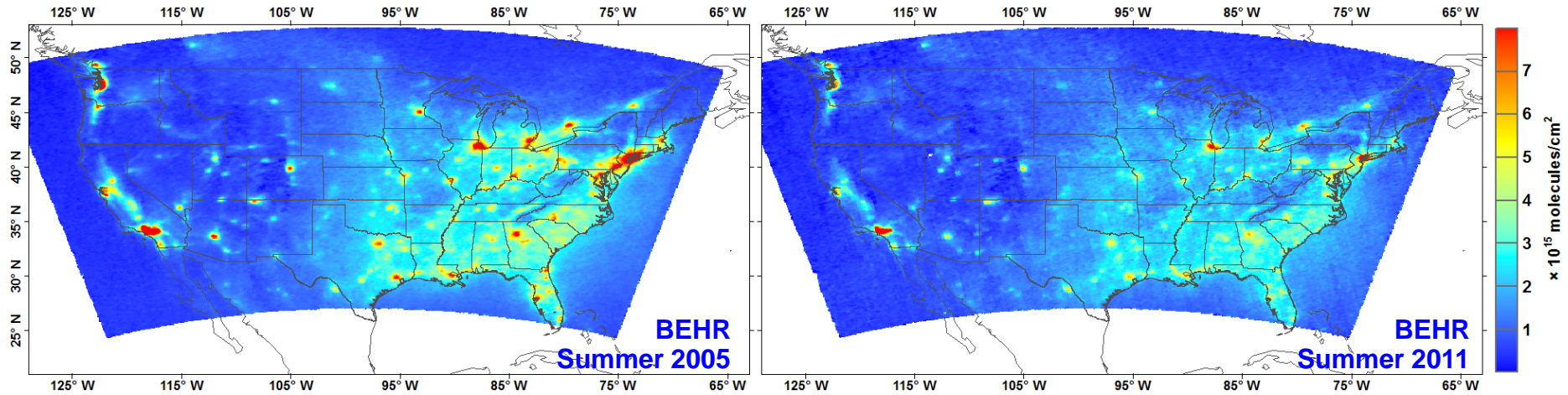
Forecast global inventory for 2009, based on bottom-up inventory for 2006 and monthly OMI NO₂ for 2006-2009

-1.8 -1.5 -1.1 -0.7 -0.3 0.0 0.3 0.7 1.1 1.5 1.8
x 10¹¹ atoms N cm⁻² s⁻¹

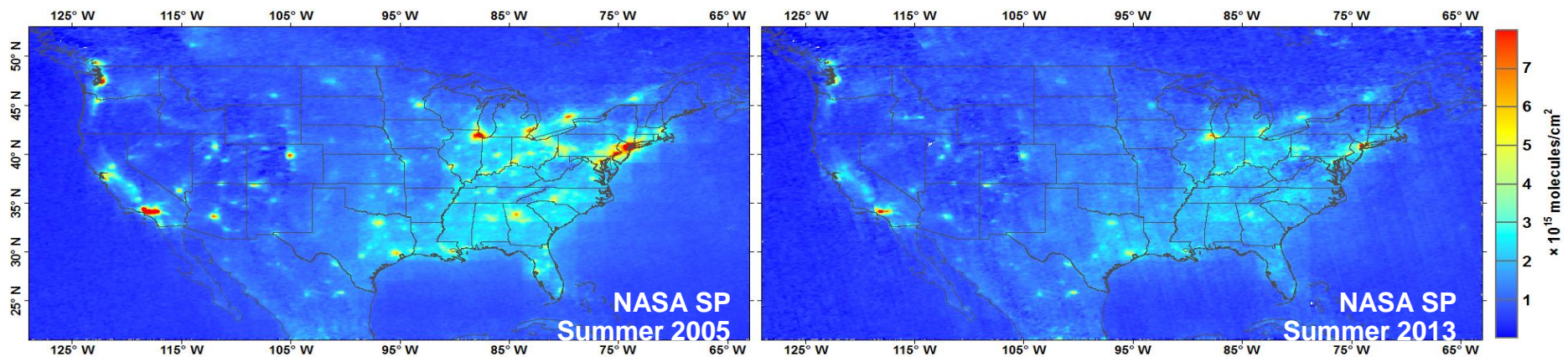
Lamsal et al., 2011

General decrease of OMI NO₂ over U.S. since 2005

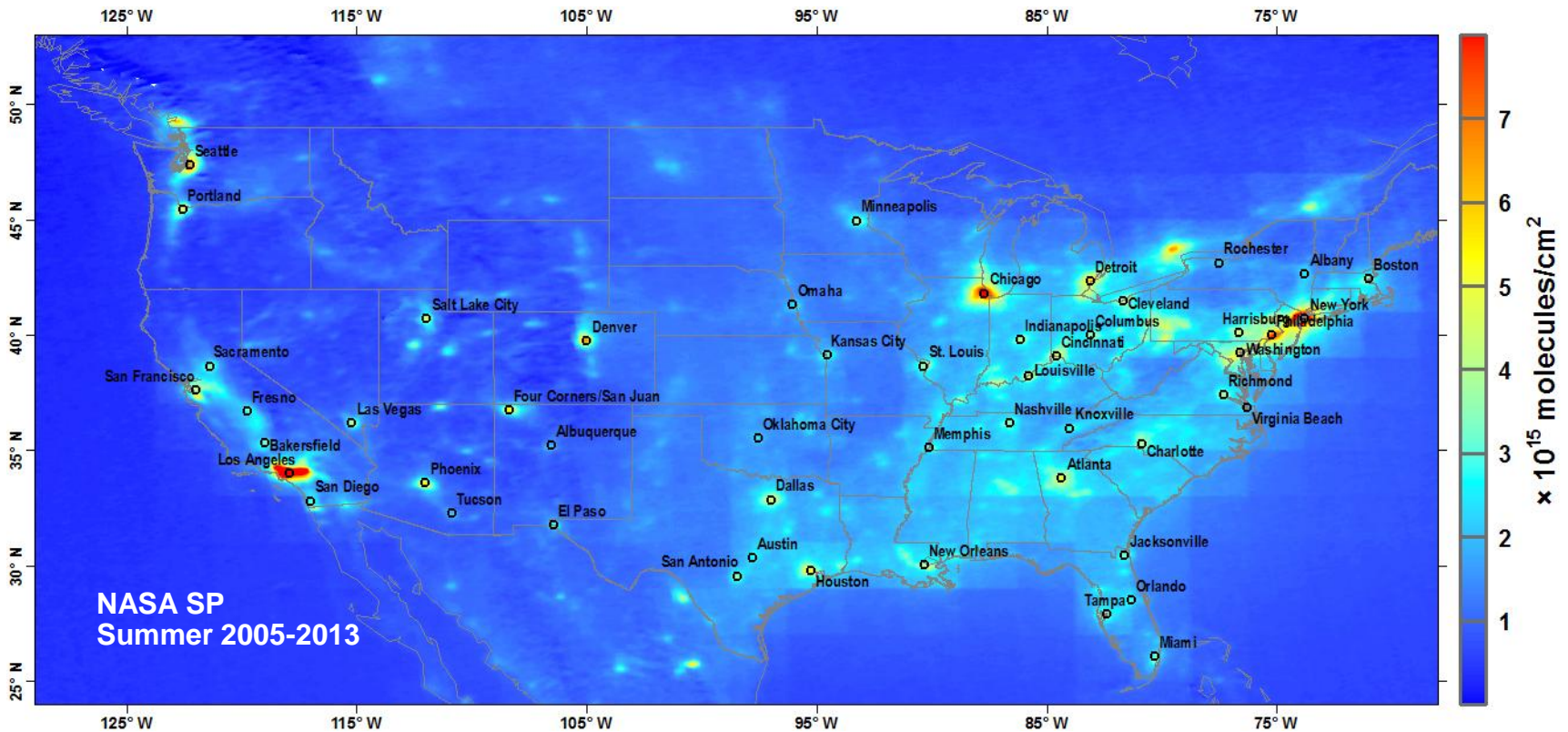
Summertime BEHR OMI NO₂ (2005 vs. 2011)



Summertime NASA SP OMI NO₂ (2005 vs. 2013)



New examination of urban NO_x emissions using OMI



- Examined the top 80 urban areas on the basis of population
- Combined adjacent urban areas that share the same NO₂ hotspot
- Excluded some urban areas, the NO₂ signals of which are not isolated

51 urban areas selected for further examination

- Represent about 40% of total NO_x emissions in the U.S.



Example: Chicago

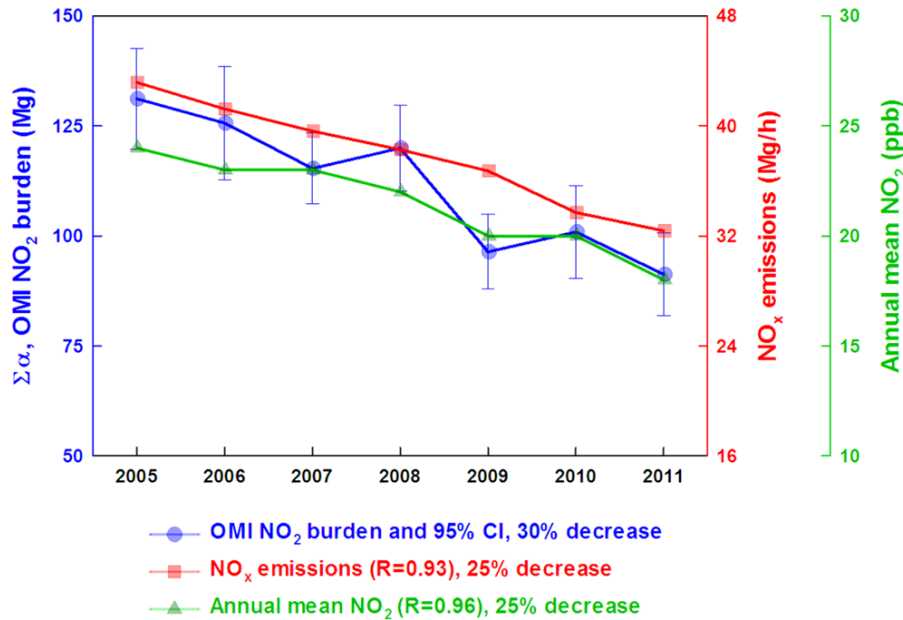
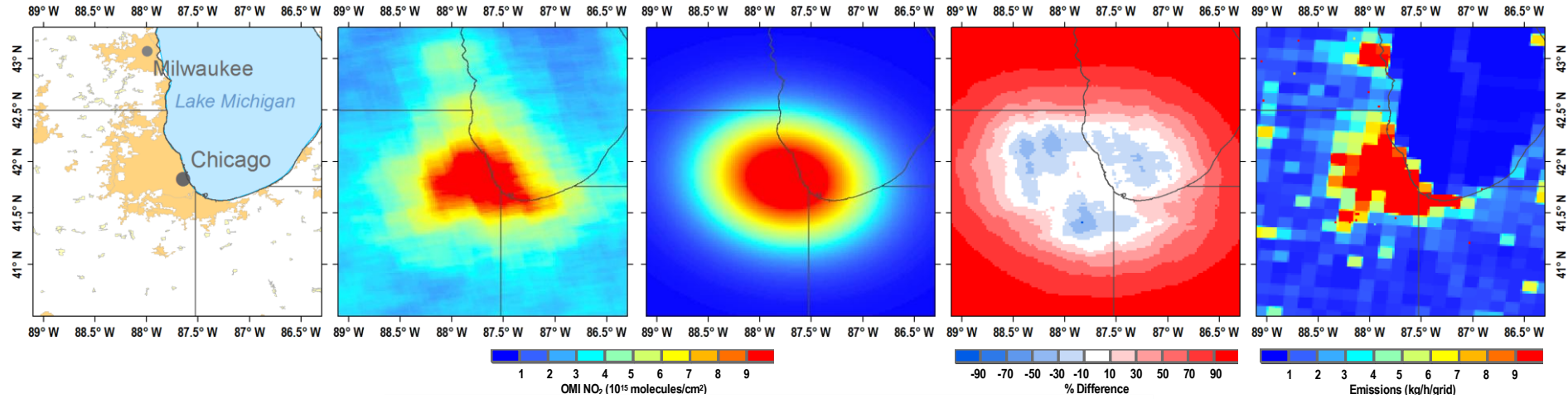
Urban Area

OMI BEHR 2005

2-D Gaussian Fit

% Difference

NO_x Emissions (EI)



Example: Houston

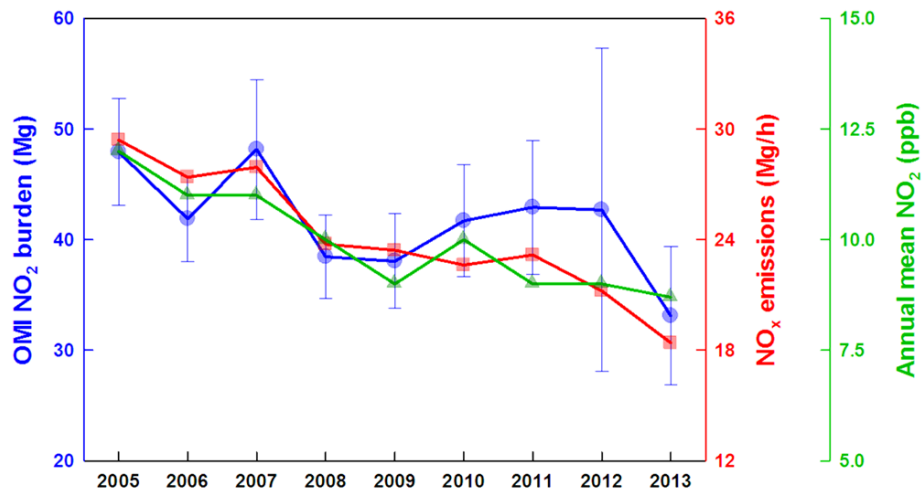
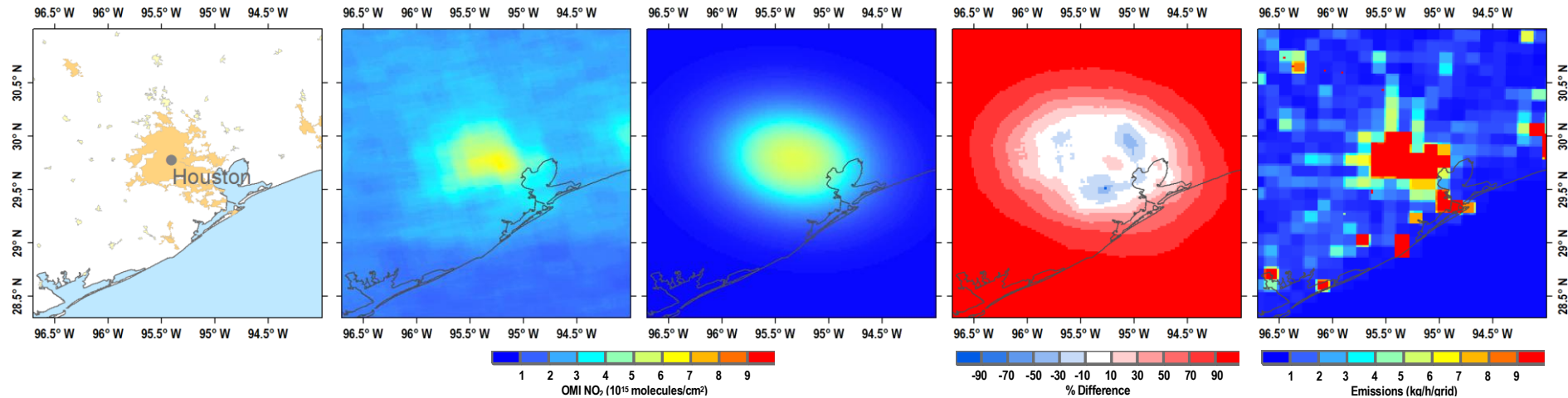
Urban Area

OMI NASA SP 2005

2-D Gaussian Fit

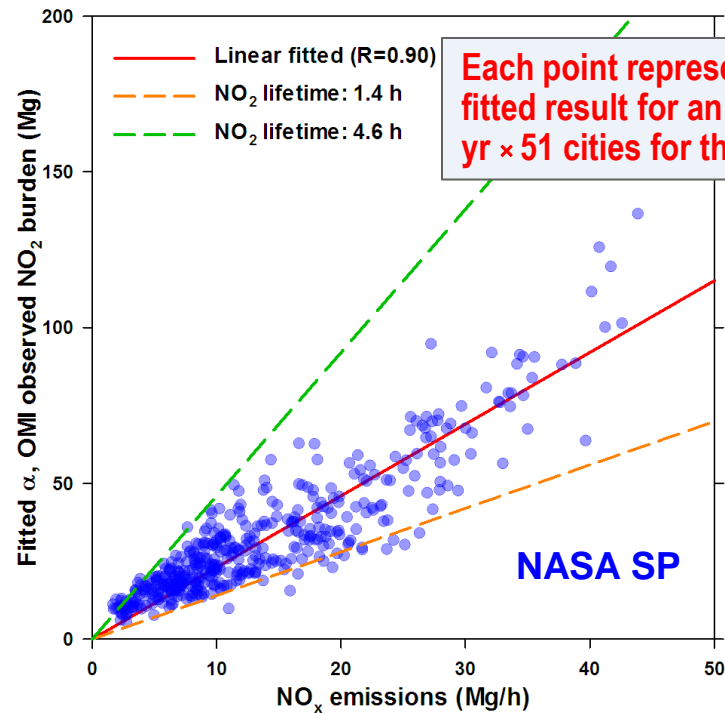
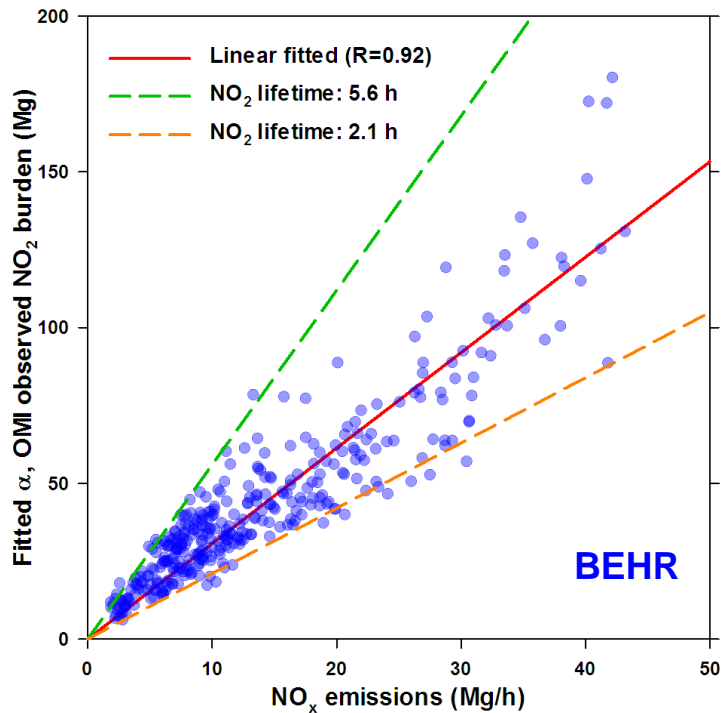
% Difference

NO_x Emissions (EI)



- OMI NO₂ burden and 95% CI, 31% decrease
- NO_x emissions (R=0.80), 37% decrease
- ▲— Annual mean NO₂ (R=0.72), 28% decrease

NO_x emissions vs OMI NO₂ burden



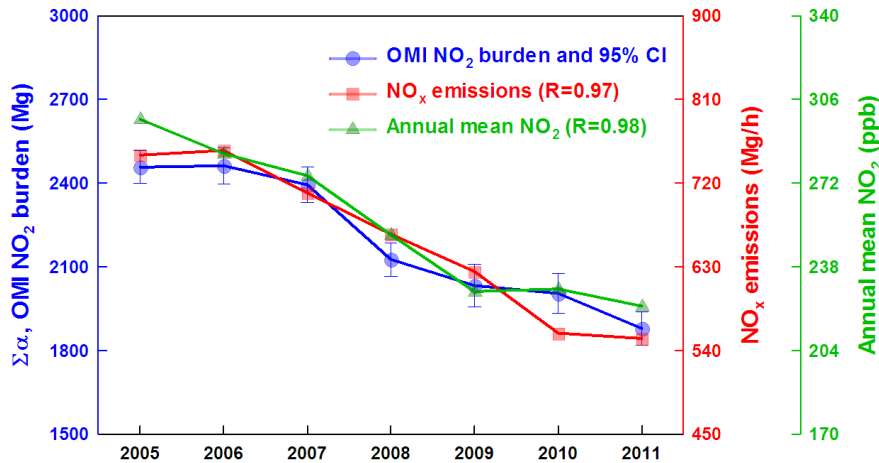
Each point represents a yearly fitted result for an urban area: 9 yr × 51 cities for the NASA SP.

- Good agreement between NO_x emissions and OMI NO₂ observations
- The 95% CI of the summertime NO₂ dispersion lifetime in U.S. urban areas
 - **Berkeley retrievals** 2.1~5.6 h **NASA retrievals** 1.4~4.6 h
- Uncertainties of urban NO_x emissions estimated from OMI NO₂ observations
 - **Berkeley retrievals** ±45% **NASA retrievals** ±57%

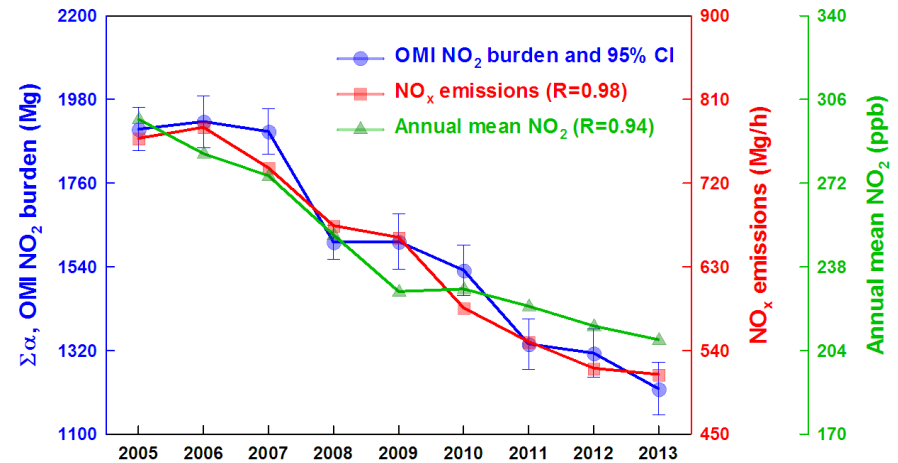


Trend of the OMI NO₂ burden summed over all selected urban areas

BEHR



NASA SP



From 2005 to 2011

From 2005 to 2013

Total amount of NO₂ observed by the OMI over selected urban areas

24% decrease

36% decrease

Total NO_x emissions from selected urban areas

26% decrease

33% decrease

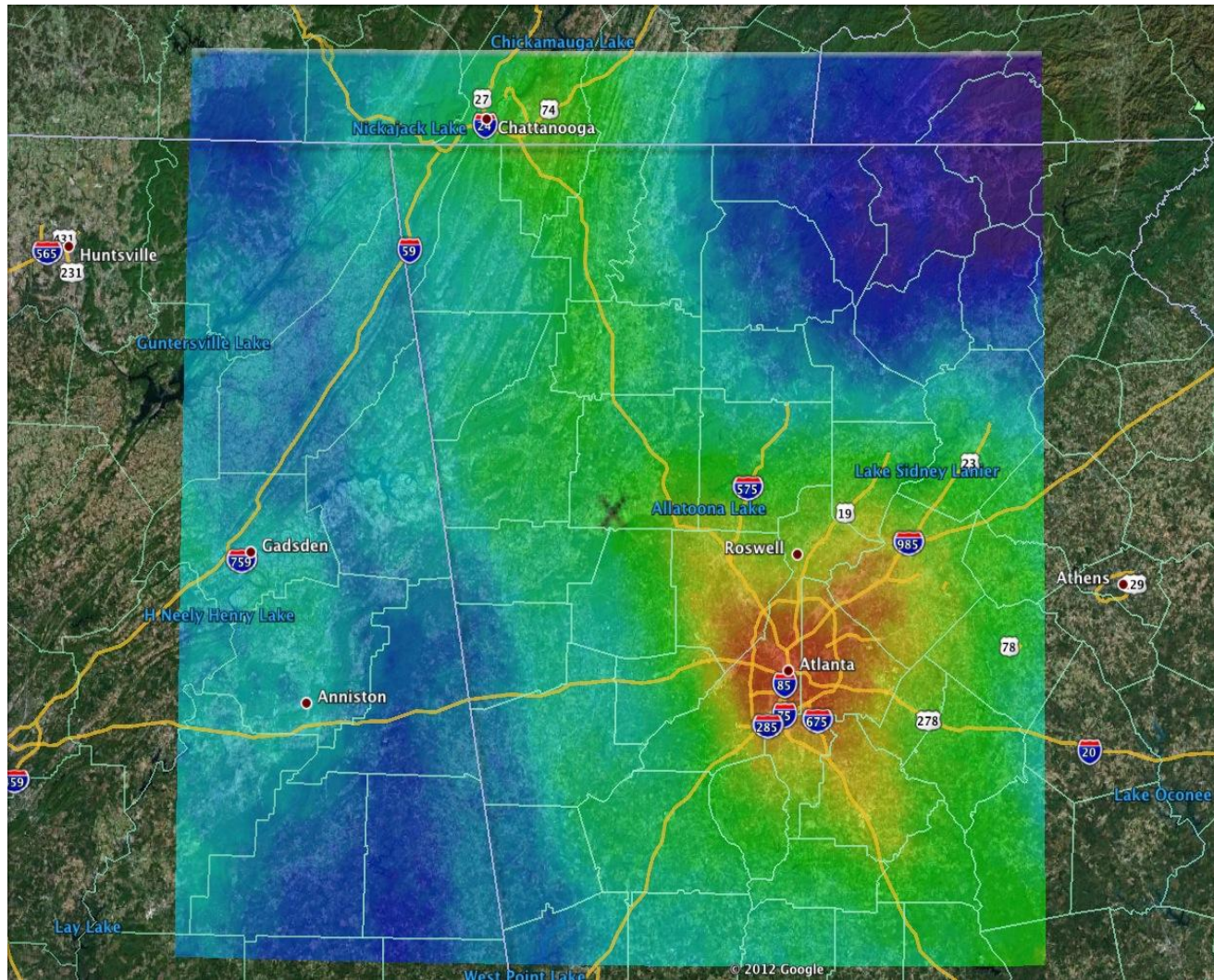
Averages of annual mean NO₂ concentrations in selected urban areas

25% decrease

30% decrease



OMI NO₂ around Atlanta using oversampling—a transportation signal?

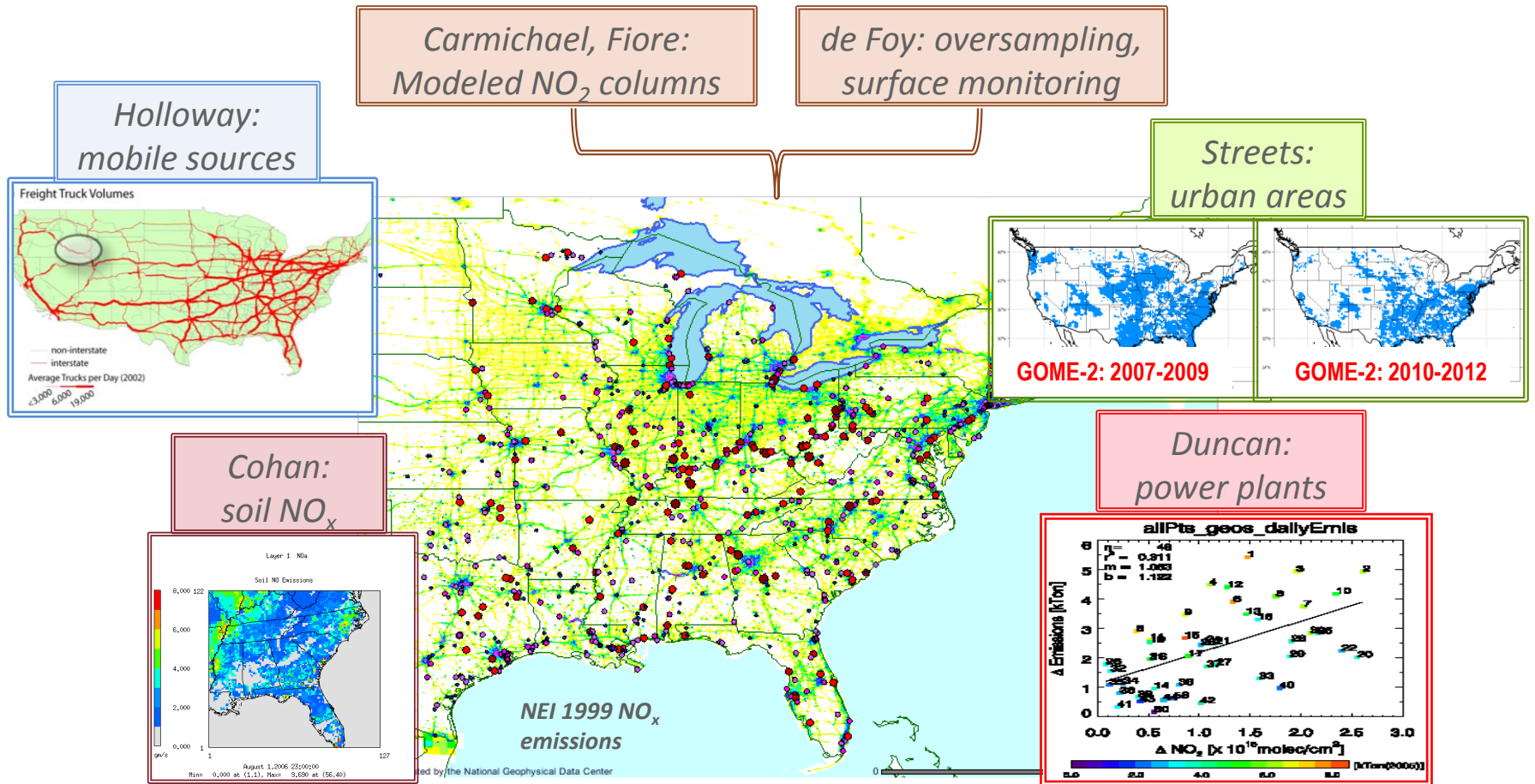


Bryan Duncan, unpublished, 2012

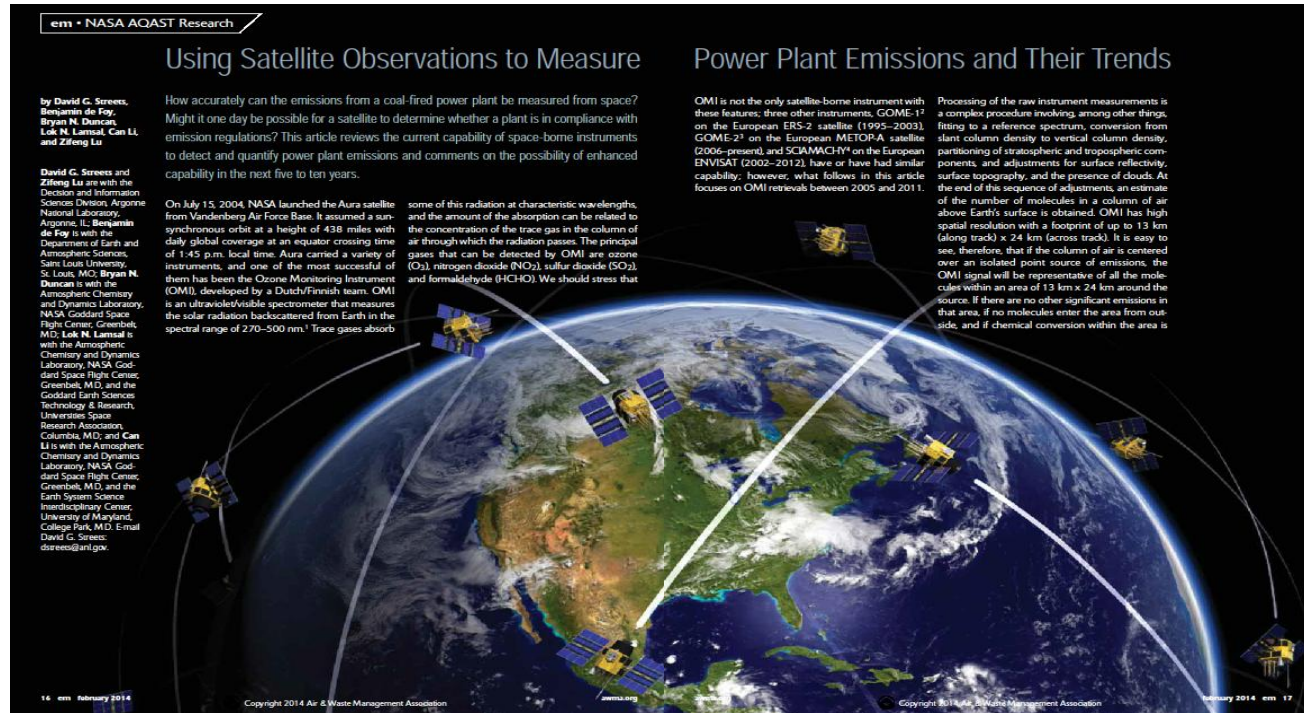


Relationships and trends among satellite NO₂ columns, NO_x emissions, and air quality in North America

David Streets (PI), Greg Carmichael, Dan Cohan, Ben de Foy, Bryan Duncan, Arlene Fiore, and Tracey Holloway



AQAST publications on emissions



- Duncan, B.N., et al., *Satellite Data for U.S. Air Quality Applications: Examples of Applications, Summary of Data End-User Resources, Answers to FAQs, and Common Mistakes to Avoid*, Atmospheric Environment, submitted (2014).
- Streets, D.G., et al., *Using Satellite Observations to Measure Power Plant Emissions and Their Trends*, EM Magazine, 16-21 (Feb 2014).
- Lu, Z., et al., *Ozone Monitoring Instrument Observations of Interannual Increases in SO₂ Emissions from Indian Coal-Fired Power Plants during 2005-2012*, Environmental Science & Technology, 47, 13,993-14,000 (2013).
- Duncan, B.N., et al., *The Observed Response of Ozone Monitoring Instrument (OMI) NO₂ Columns to NO_x Emission Controls on Power Plants in the United States: 2005-2011*, Atmospheric Environment, 81, 102-111 (2013).
- Streets, D.G., et al., *Emissions Estimation from Satellite Retrievals: A Review of Current Capability*, Atmospheric Environment, 77, 1011-1042 (2013).
- Lu, Z., and D.G. Streets, *Increase in NO_x Emissions from Indian Thermal Power Plants during 1996-2010: Unit-Based Inventories and Multi-Satellite Observations*, Environmental Science & Technology, 46, 7463-7470 (2012).