

# Dense networks and next generation satellites: Data for new approaches to understanding the atmosphere

An aerial photograph of San Francisco, California, taken at dusk. The city's dense urban landscape is illuminated by streetlights and building lights, creating a warm glow against the deep blue twilight sky. The Golden Gate Bridge is visible in the distance, spanning the water. The Transamerica Pyramid stands out prominently among the skyscrapers in the downtown area.

Ronald C. Cohen  
UC Berkeley

\$ BAAQMD, NSF, NASA, UC Berkeley,  
HEI, EDF, Koret Foundation

EPA Air Sensors 2018

# Connecting Science to Policy

Can observations provide better tools for epidemiology? And individual health?

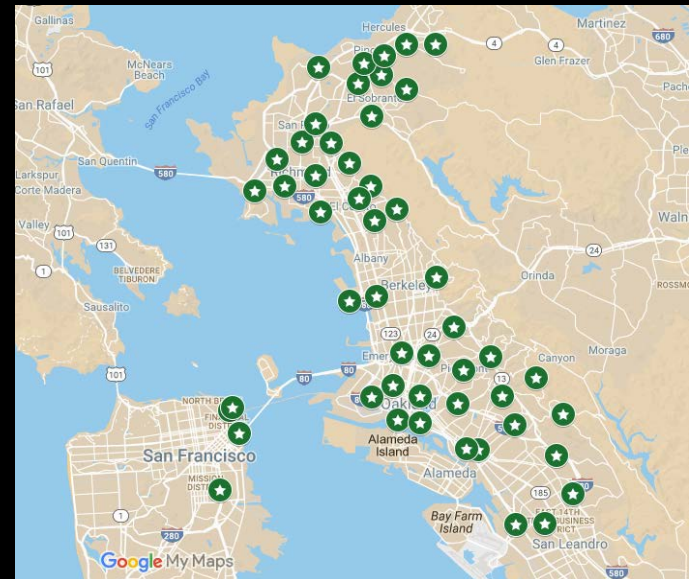
How can observations directly evaluate the efficacy of emissions policies? On relevant timescales?



New dense observing systems coupled to similarly high resolution inverse and assimilation models are changing how we approach these questions.

Instead of extrapolating from points we are building maps and making movies.

Remote sensing  
 $\text{NO}_2$ ,  $\text{CO}$ ,  $\text{CO}_2$ , ...



Ubiquitous sensing  
 $\text{CO}_2$ ,  $\text{NO}_2$ ,  $\text{O}_3$ , ...

New questions that are hard (impossible?) to answer with the current suite of instruments—but would be possible with dense networks.

Is cold start the dominant source of  $\text{NO}_x$ ; if so what changes in spatial pattern have occurred are occurring?

Are emissions of household organics competitive with emissions from vehicles as source of urban reactive carbon? (McDonald et al. Science 2018)

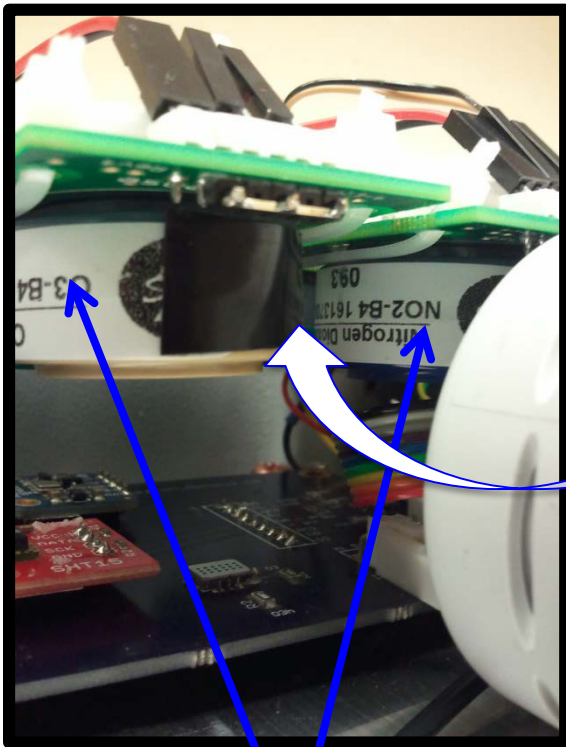
Observations:

Berkeley Atmospheric CO<sub>2</sub>

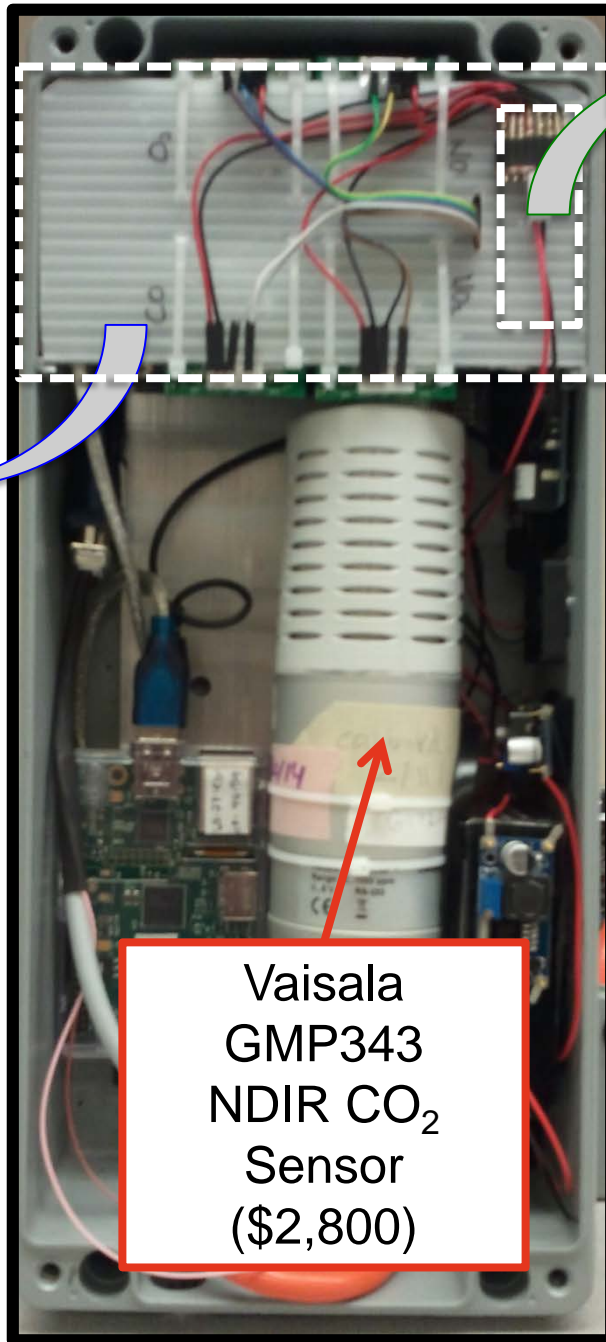
Observation Network

Models:

WRF-CHEM at 1km resolution  
home built emission inventory



Alphasense B4  
Electrochemical  
 $O_3$ , CO, NO &  
 $NO_2$  Sensors  
(\$216 ea.)

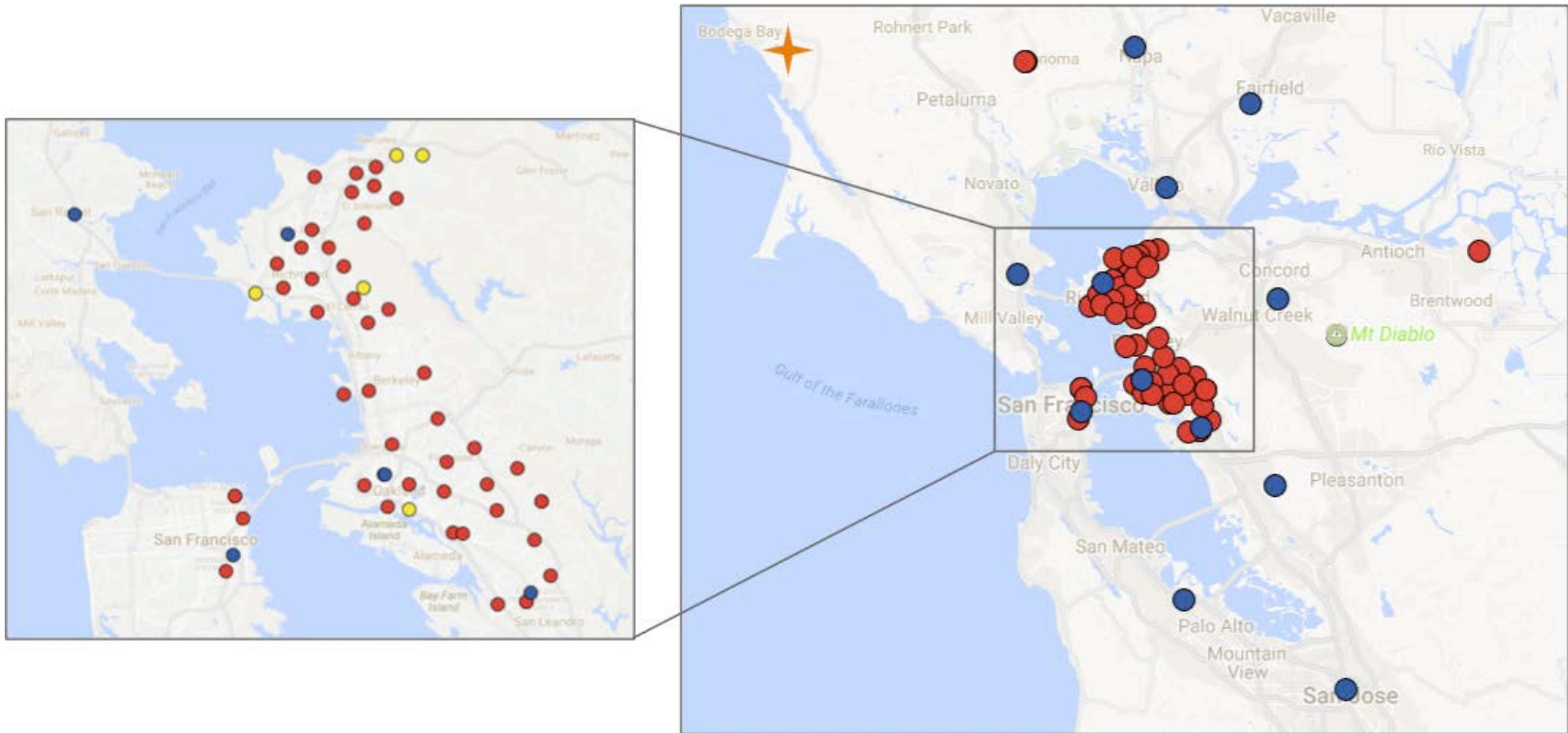


Vaisala  
GMP343  
NDIR  $CO_2$   
Sensor  
(\$2,800)



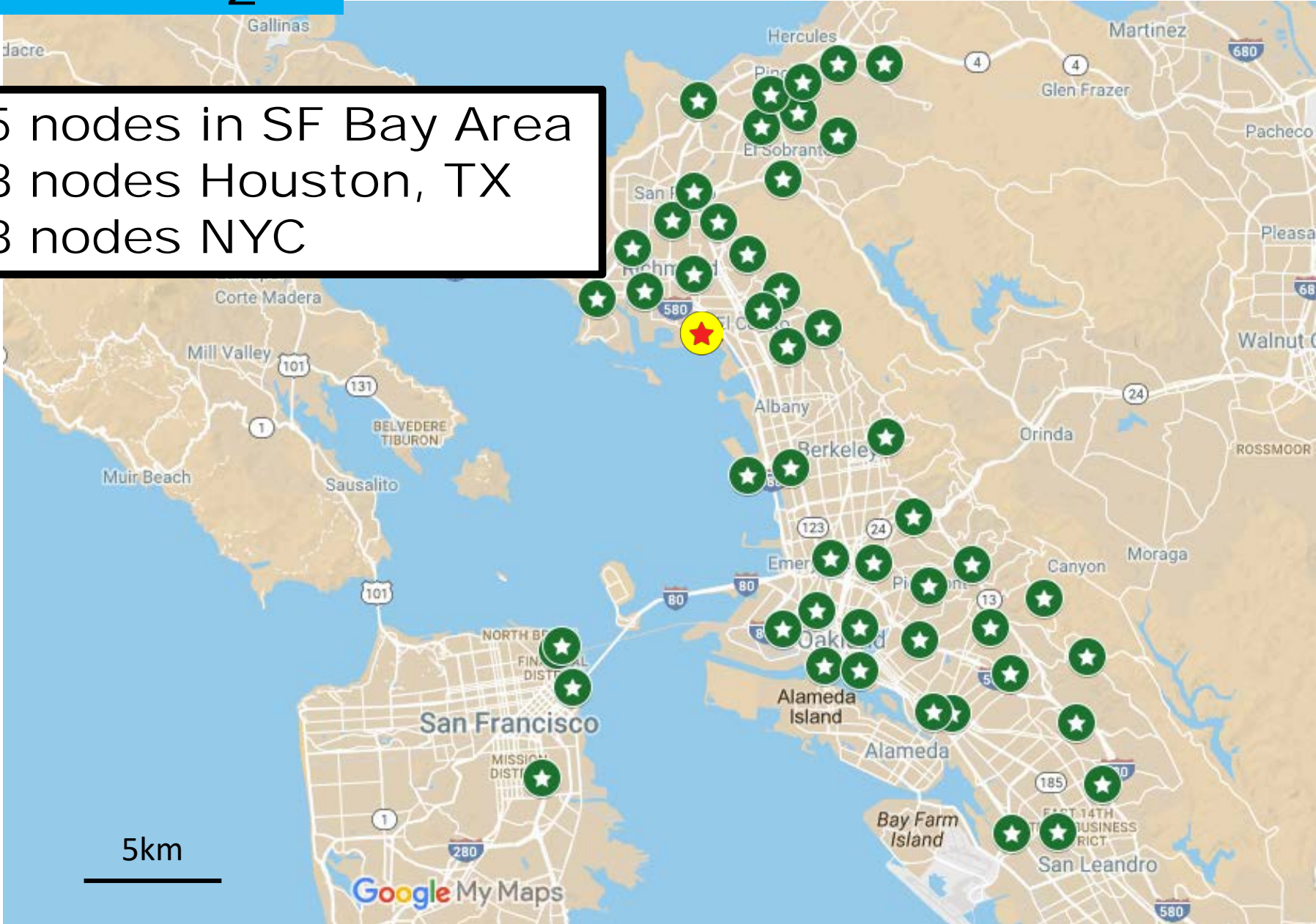
Shinyei PPD42NS  
nephelometric  
particulate matter  
sensor  
(\$16)

# Red and Yellow BEACO<sub>2</sub>N. Blue BAAQMD



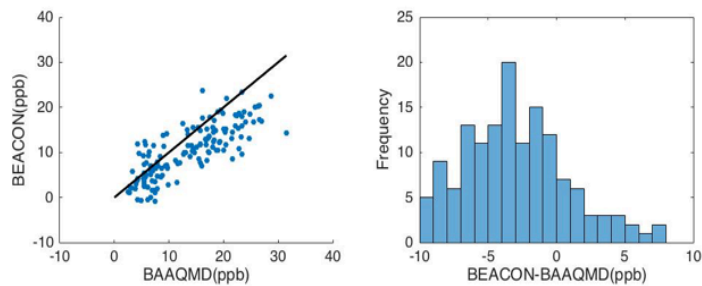
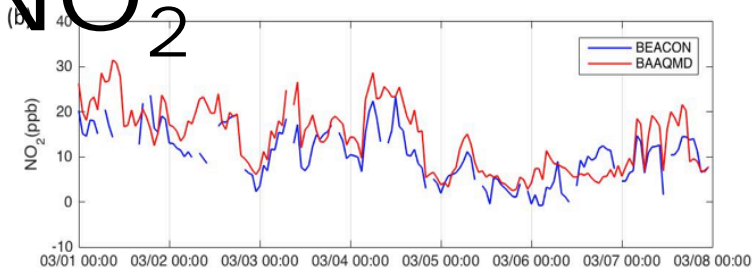
# BEACO<sub>2</sub>N

65 nodes in SF Bay Area  
18 nodes Houston, TX  
8 nodes NYC

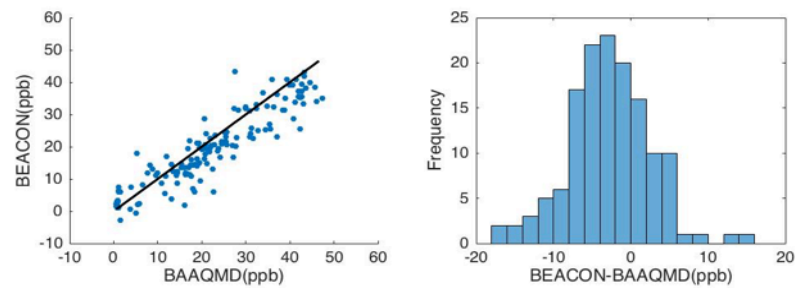
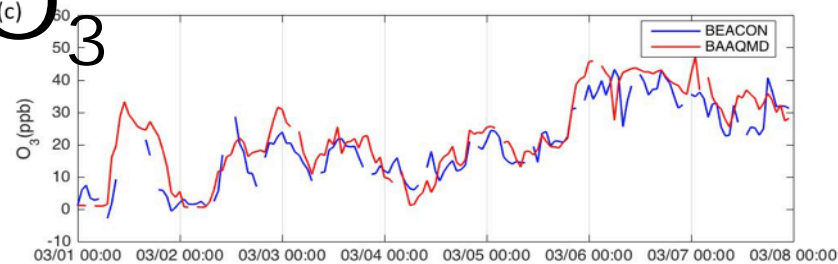




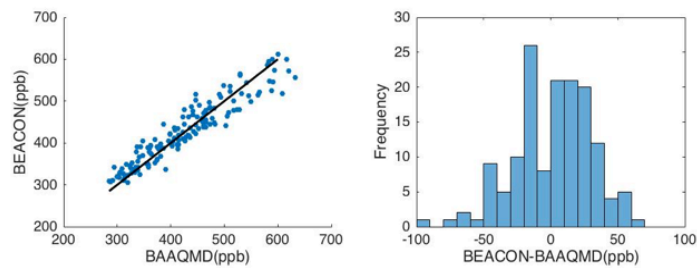
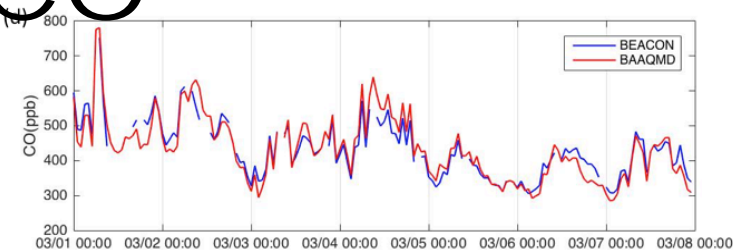
# NO<sub>2</sub>



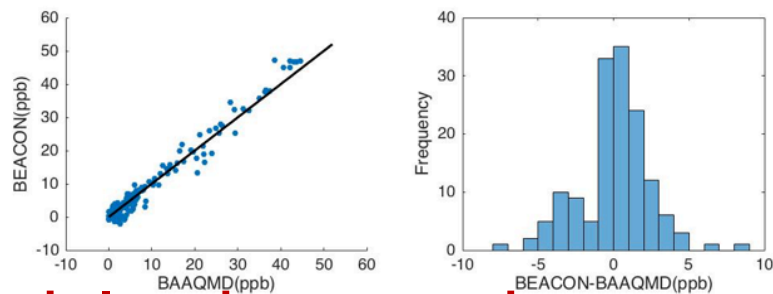
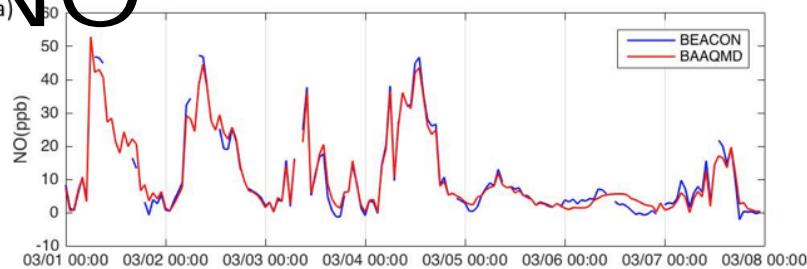
# O<sub>3</sub>



# CO

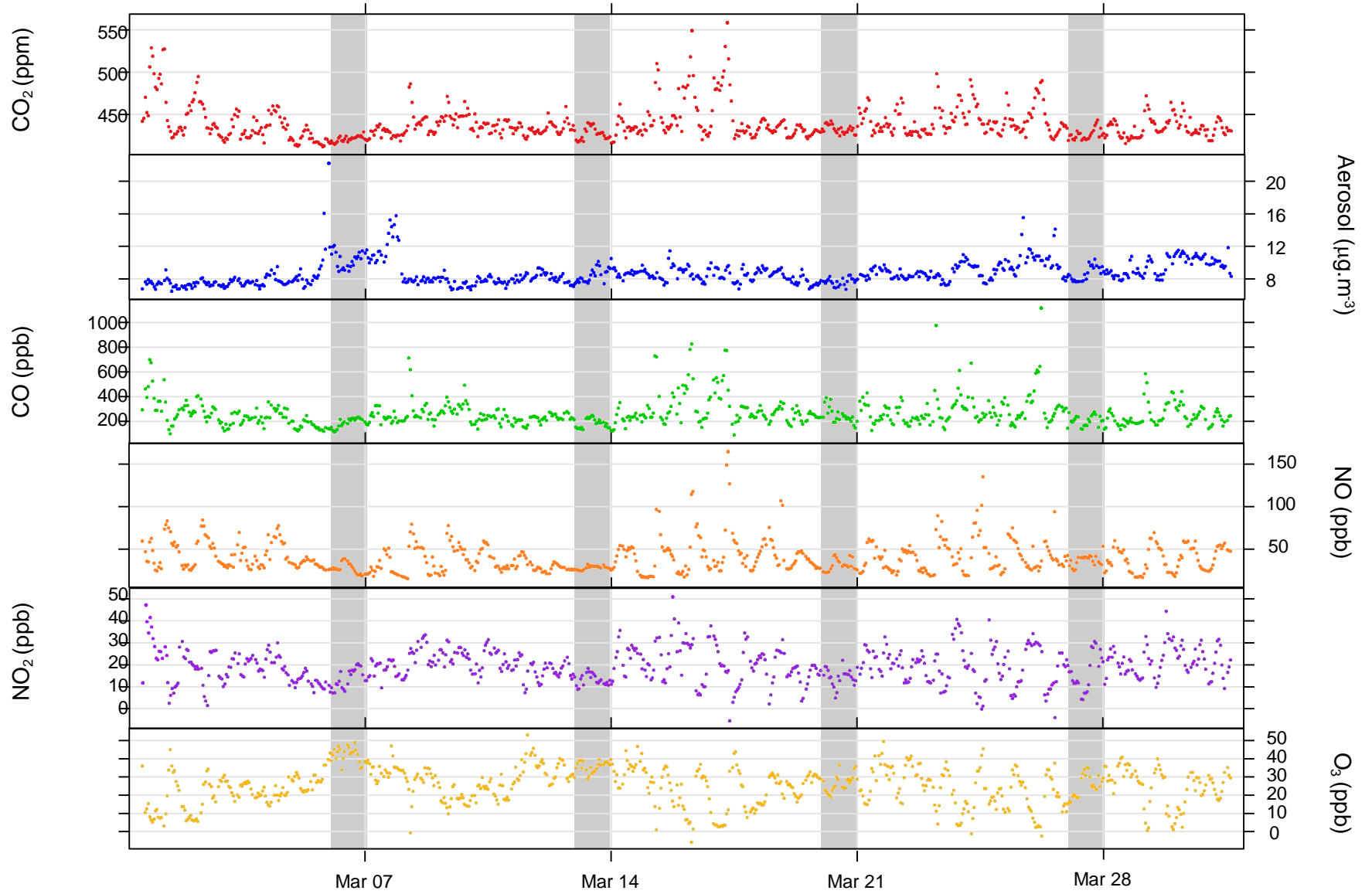


# NO

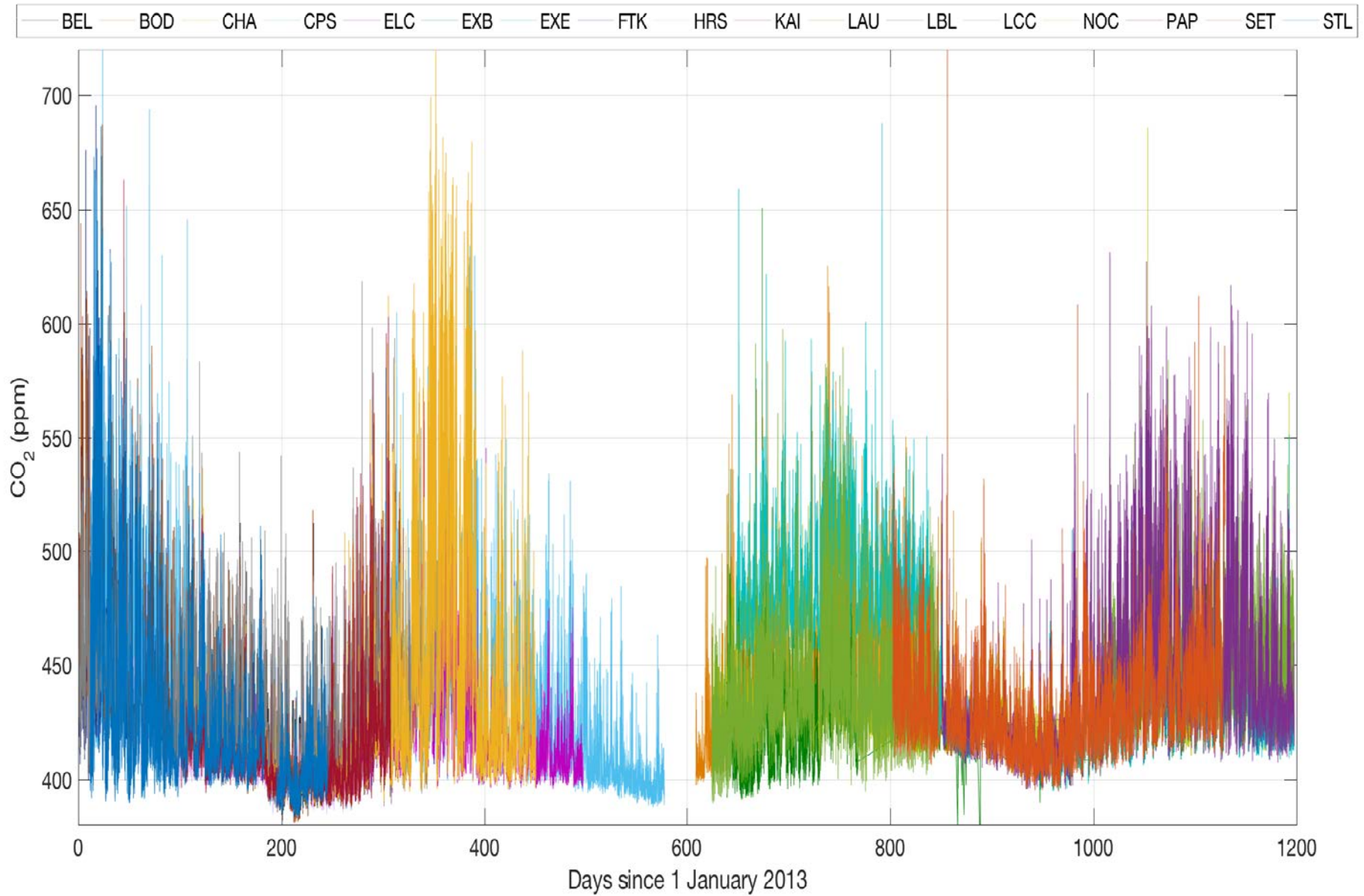


Blue BEACON<sub>2</sub>N Red Independent

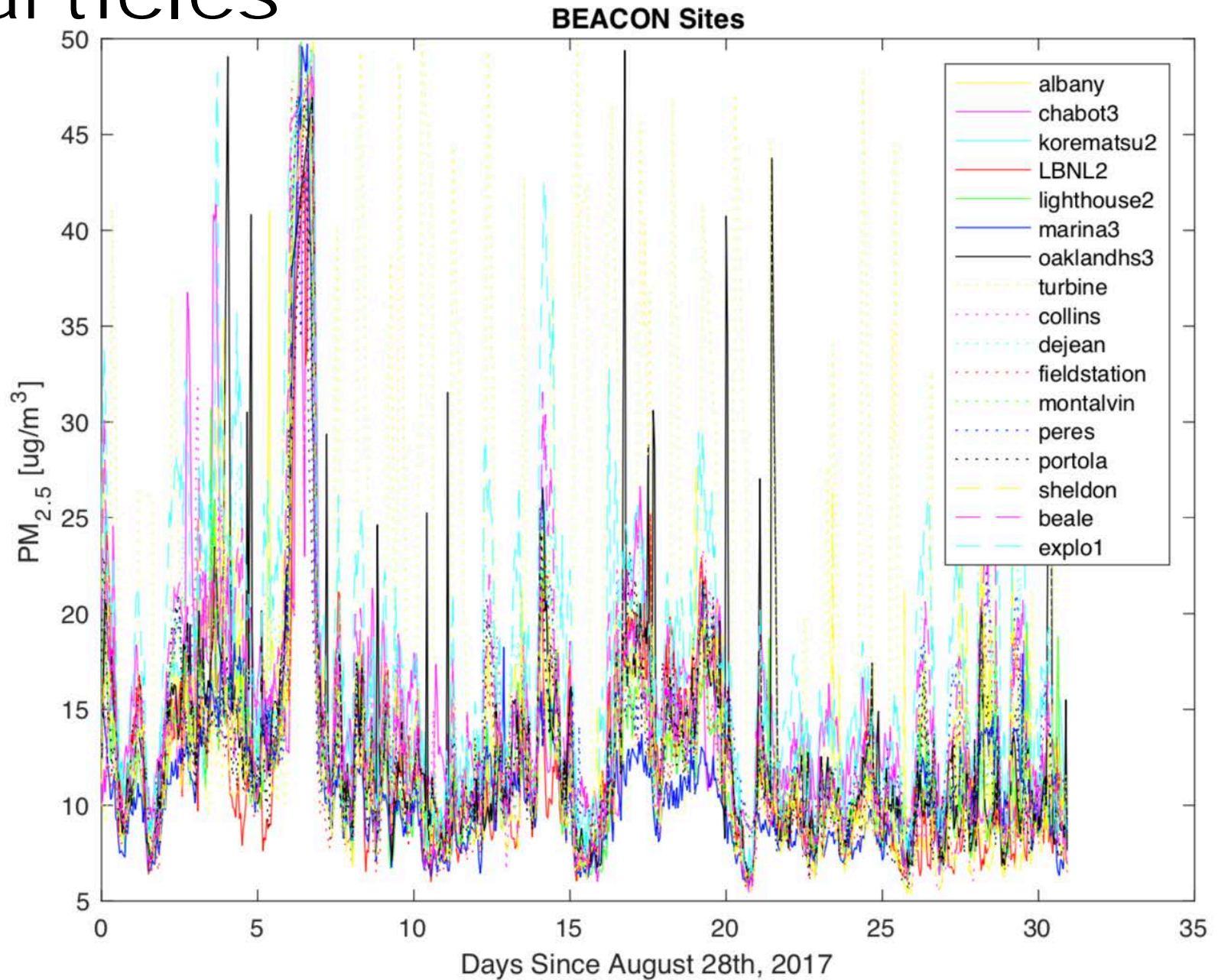
# Laney - March 2016



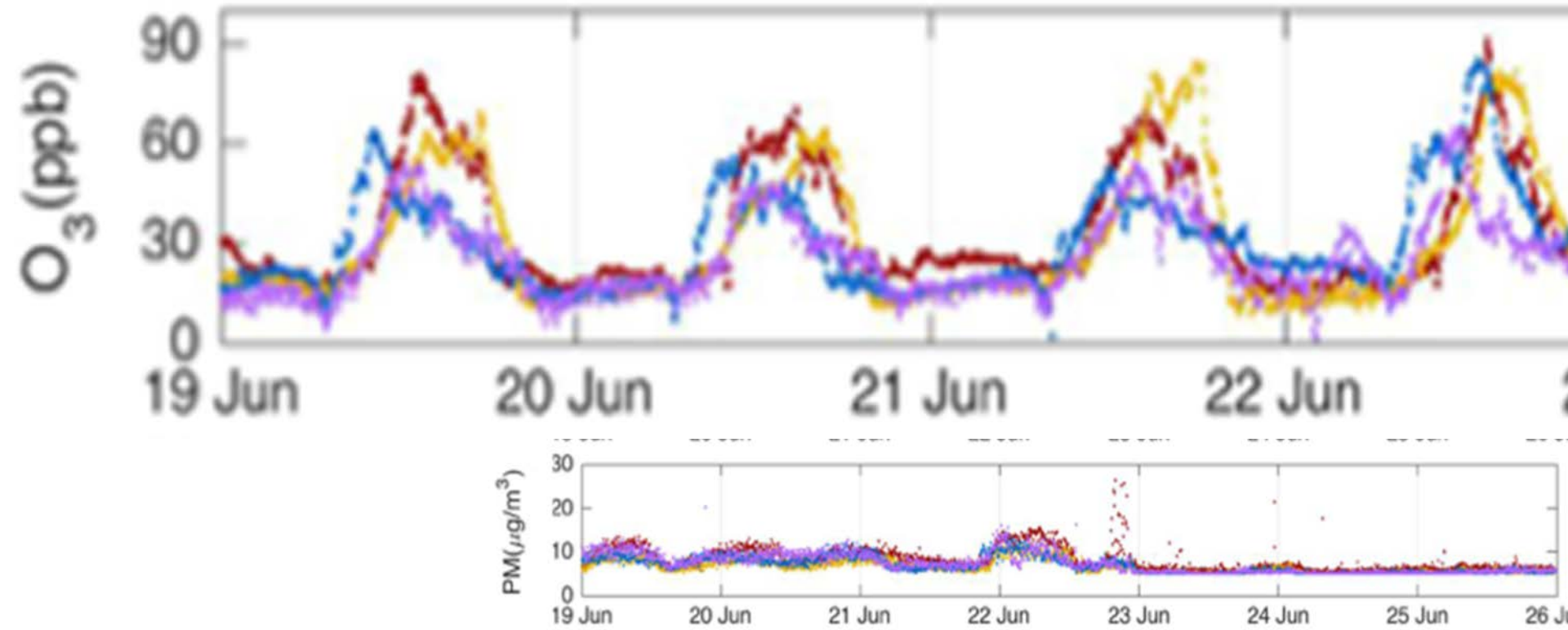
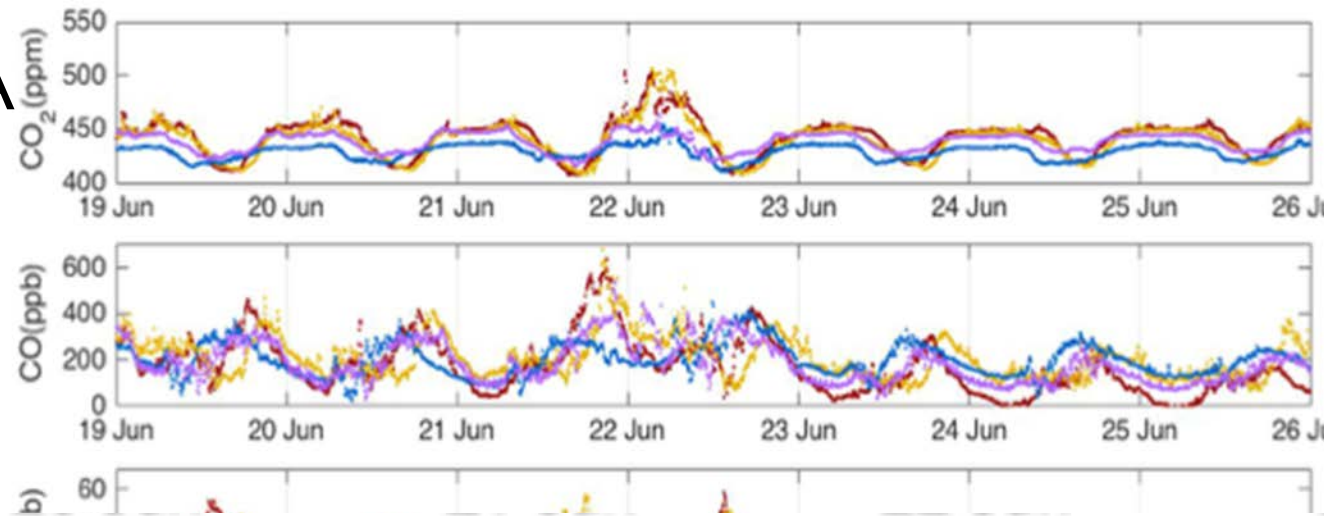
# CO<sub>2</sub>



# Particles



# 4 locations in Richmond, CA



# BEACO<sub>2</sub>N: A high spatial resolution observing system for GHGs (CO<sub>2</sub>) and air quality (CO, O<sub>3</sub>, NO, NO<sub>2</sub>, particles)

## CO<sub>2</sub>

A.A. Shusterman, V. Teige, A.J. Turner, C. Newman, J. Kim, and R.C. Cohen: *The BErkeley Atmospheric CO<sub>2</sub> Observation Network: initial evaluation*, Atmos. Chem. Phys., 2016.

A.J. Turner, A.A. Shusterman, B.C. McDonald, V. Teige, R.A. Harley and R.C. Cohen, *Network design for quantifying urban CO<sub>2</sub> emissions: Assessing tradeoffs between precision and network density* Atmos. Chem. Phys., 2016.

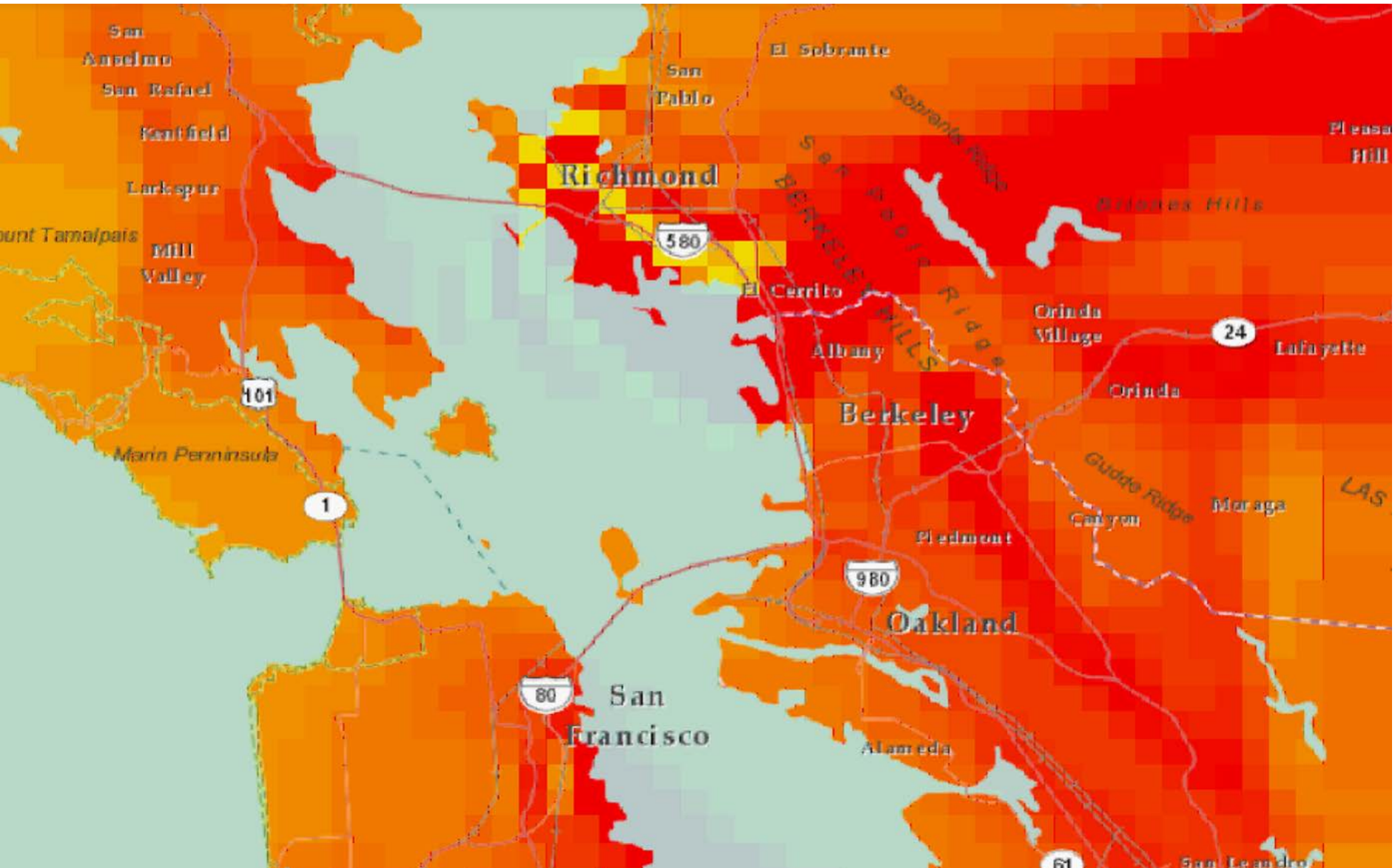
A.A. Shusterman, J. Kim, K.J. Lieschke, C. Newman, P.J. Wooldridge, R.C. Cohen, *Observing local CO<sub>2</sub> sources using low-cost, near-surface urban monitors*, Atmos. Chem. Phys. Disc. 2018.

## AQ gases

J. Kim, A.A. Shusterman, K.J. Lieschke, C. Newman, and R.C. Cohen, *The BErkeley Atmospheric CO<sub>2</sub> Observation Network: field calibration and evaluation of low-cost air quality sensors*, Atmos. Meas. Tech., 2018.

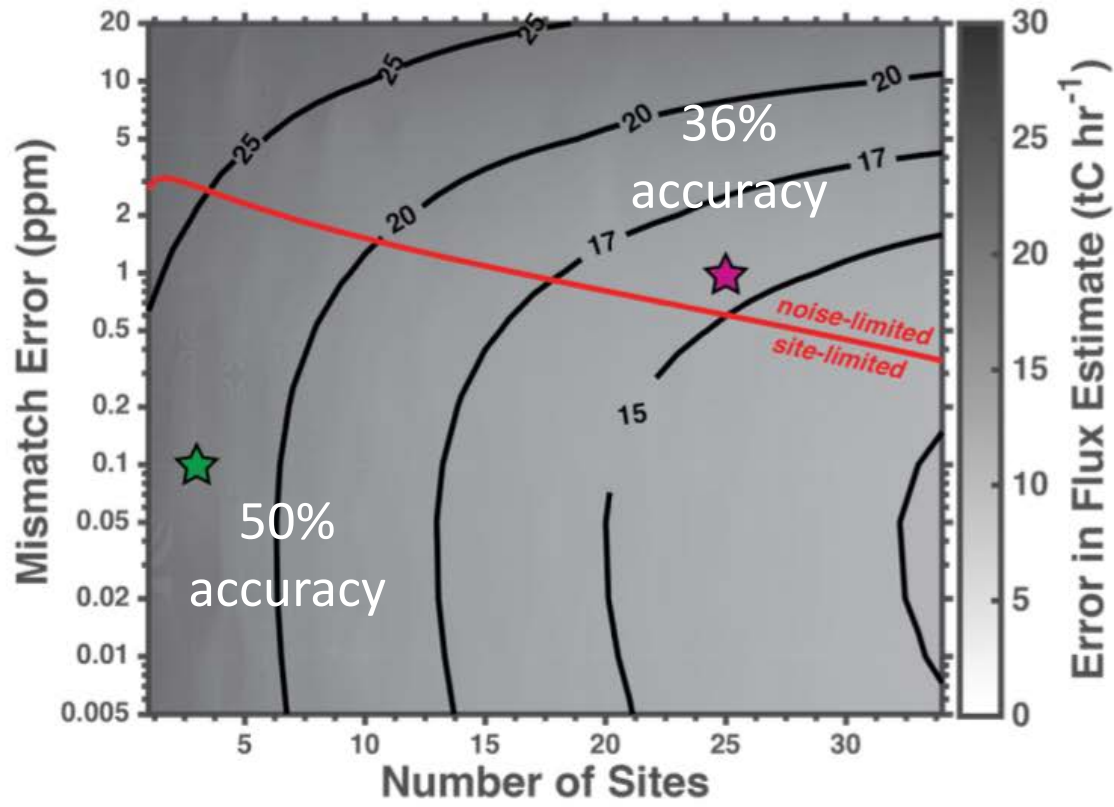
## Aerosol

# 1 frame from a 1km CO<sub>2</sub> movie



Observations aim to test this or other model

# Networks are better



Large numbers of low cost instruments (★) will outperform a few state-of-the-art high cost ones (★) for quantifying emissions within a city.

*Turner et al. ACP 2016.*



# Issues we need to think about:

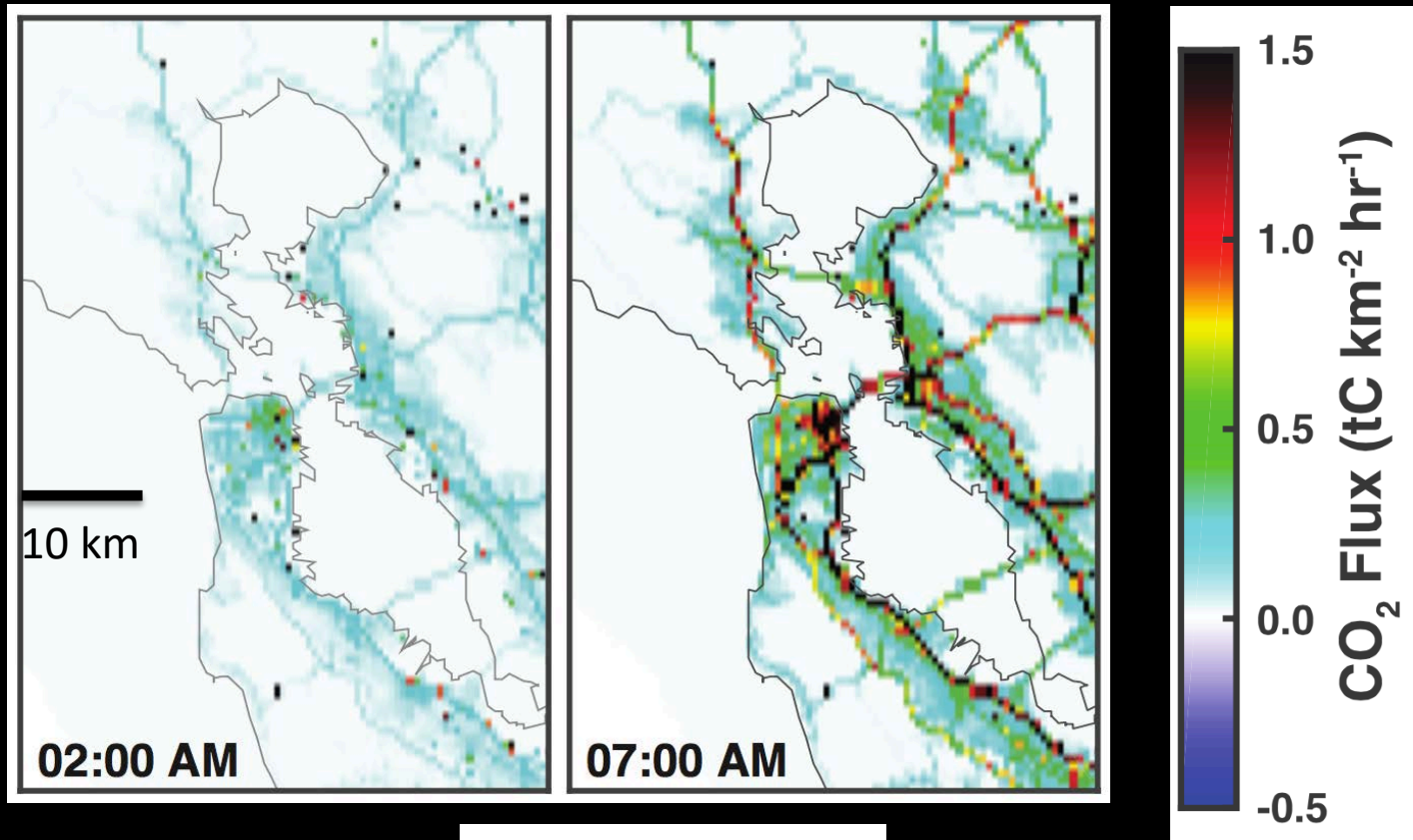
1-Calibrate and performance of a network, not single devices

2-Calibration and performance of a networked **system** of multispecies and multi-instrument (**including surface and satellite**) measurements not just one chemical at a time

3-Inverse and assimilation models coupled to observations

What is the accuracy/precision of a map?  
Does it go as  $\sqrt{n}$  of the number of  
measurements?

Emission Inventory

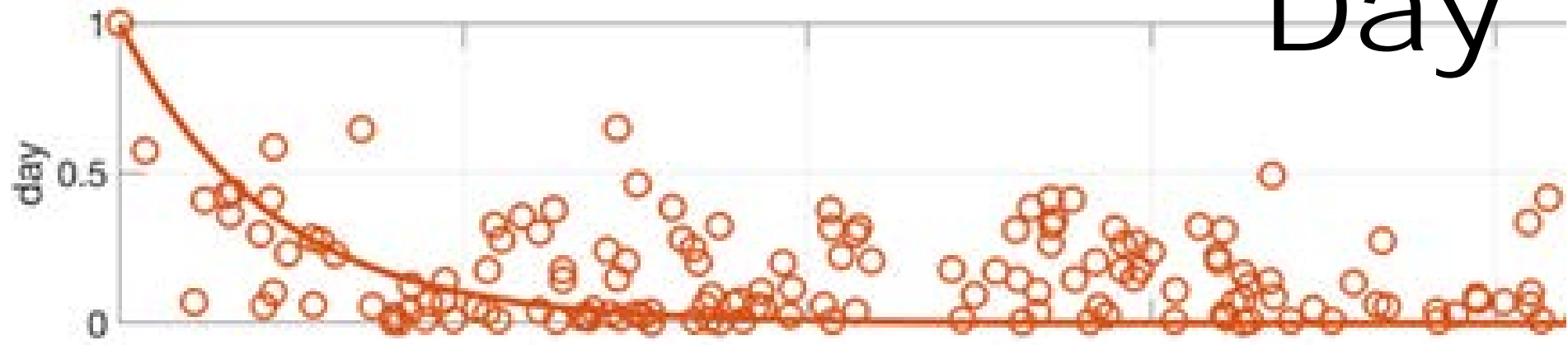


(Turner et al., 2016)

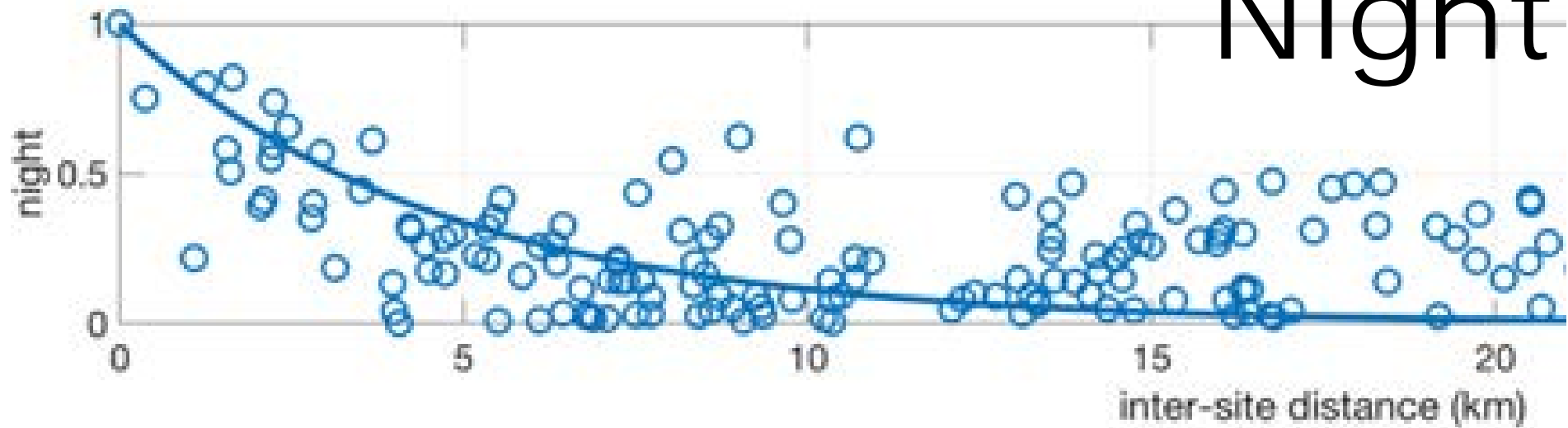


# CO<sub>2</sub> correlation length scale

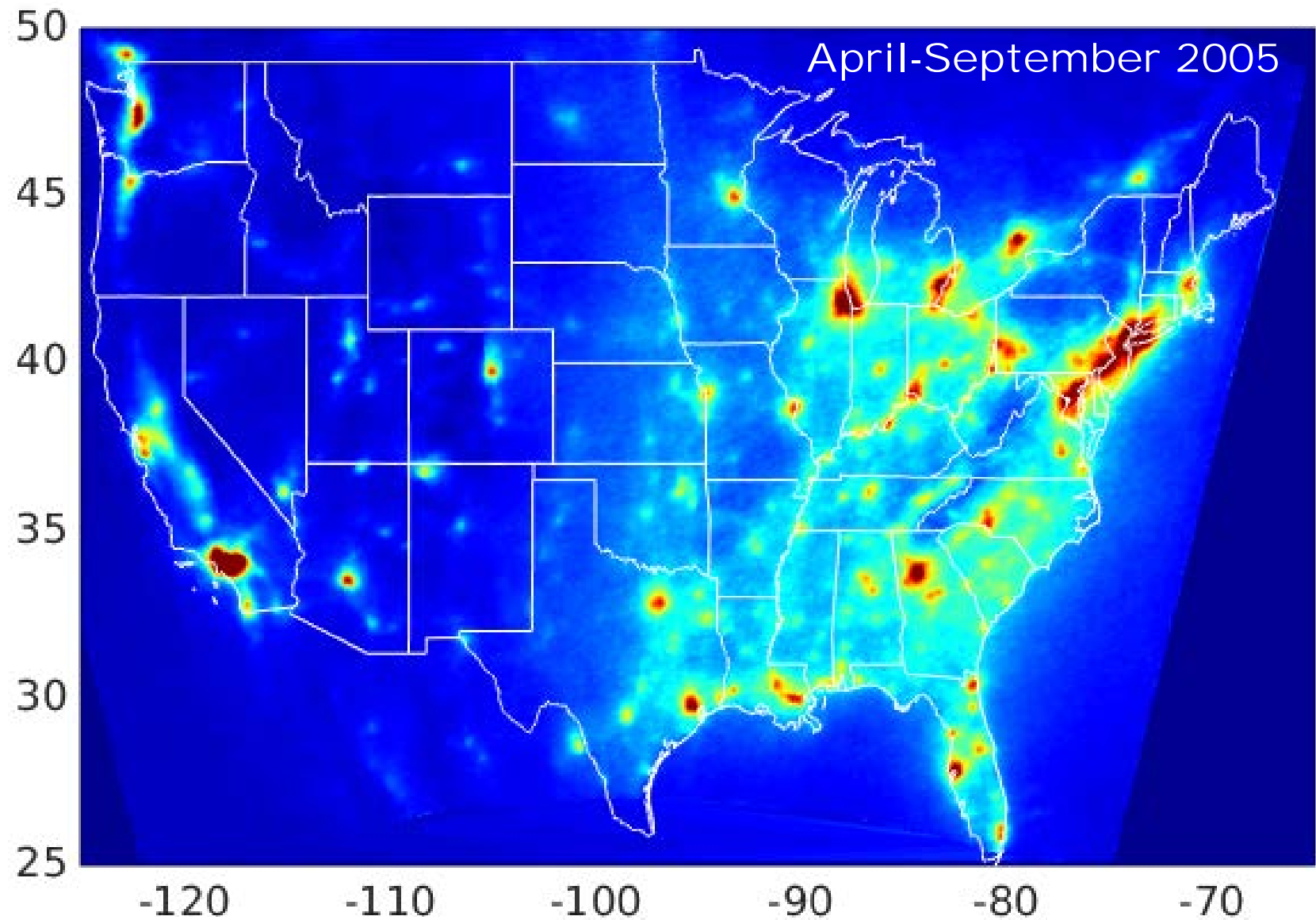
Day



Night

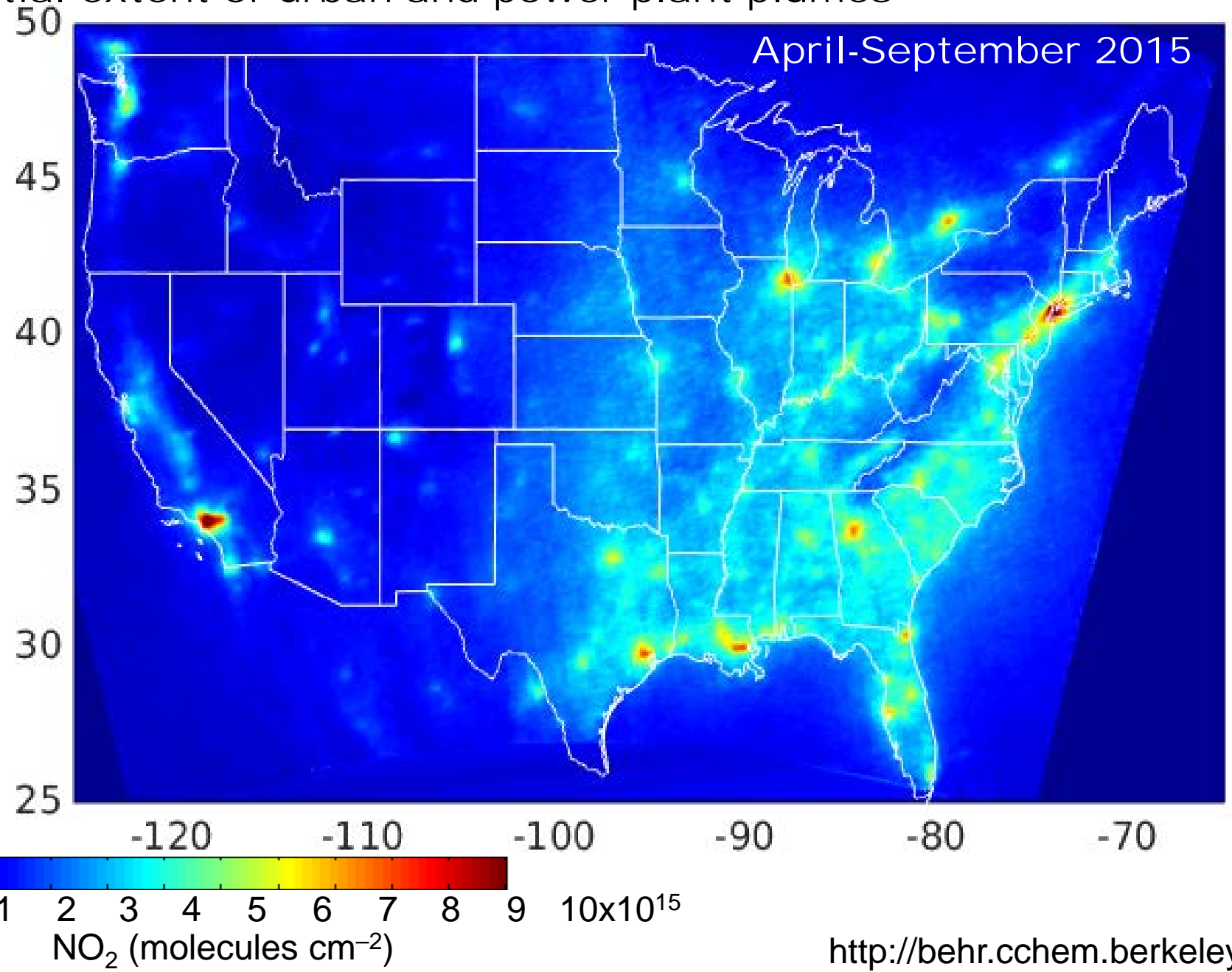


# A multicomponent observing system



0 1 2 3 4 5 6 7 8 9  $10 \times 10^{15}$   
NO<sub>2</sub> (molecules cm<sup>-2</sup>) OMI Berkeley High-resolution Retrieval (BEHR)

Large decreases over the last decade in U.S. result in smaller spatial extent of urban and power plant plumes



# Performance Standards

Precision and accuracy of individual instruments:

Independent of T and RH of sample and of the sensor location

Precision and accuracy should have Gaussian noise and improve with averaging over time and multiple sensors

Precision and accuracy of systems:

Information need not be at the individual sensor level—think weather; typically we care about a forecast that is a synthesis of billions of measurements

Cross sensitivities removed at system level

# Conclusions and Outlook

High space and time resolution observations using **networks** with multiple chemicals and aerosol offer a new window into mechanisms affecting emissions and chemistry in cities.

## Challenges:

- learning to interpret dense networks as more than the sum of individual instruments.
- integrating multiple different measurement approaches
- learning to think about daily variability in ways that teach us about processes and personal exposure.





Thank you!