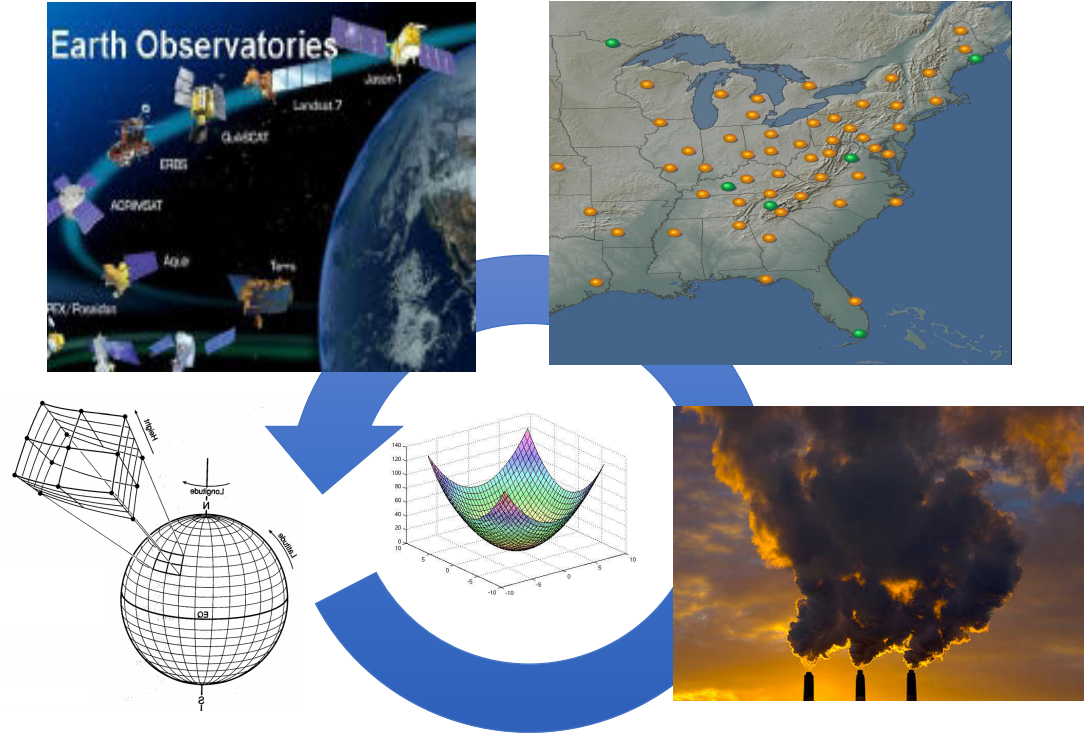


Top-down constraints on emissions of NO_2 , SO_2 and NH_3

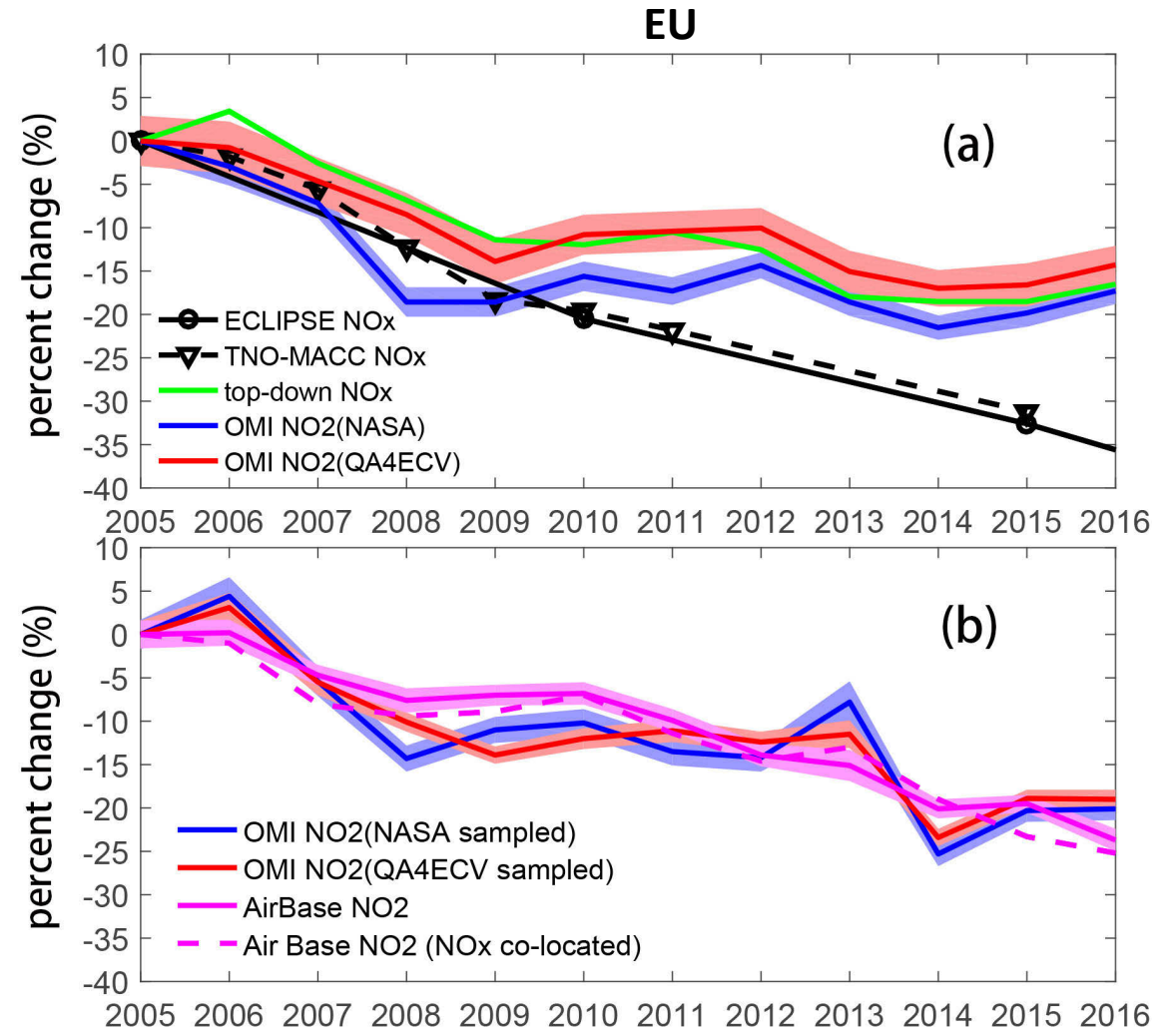
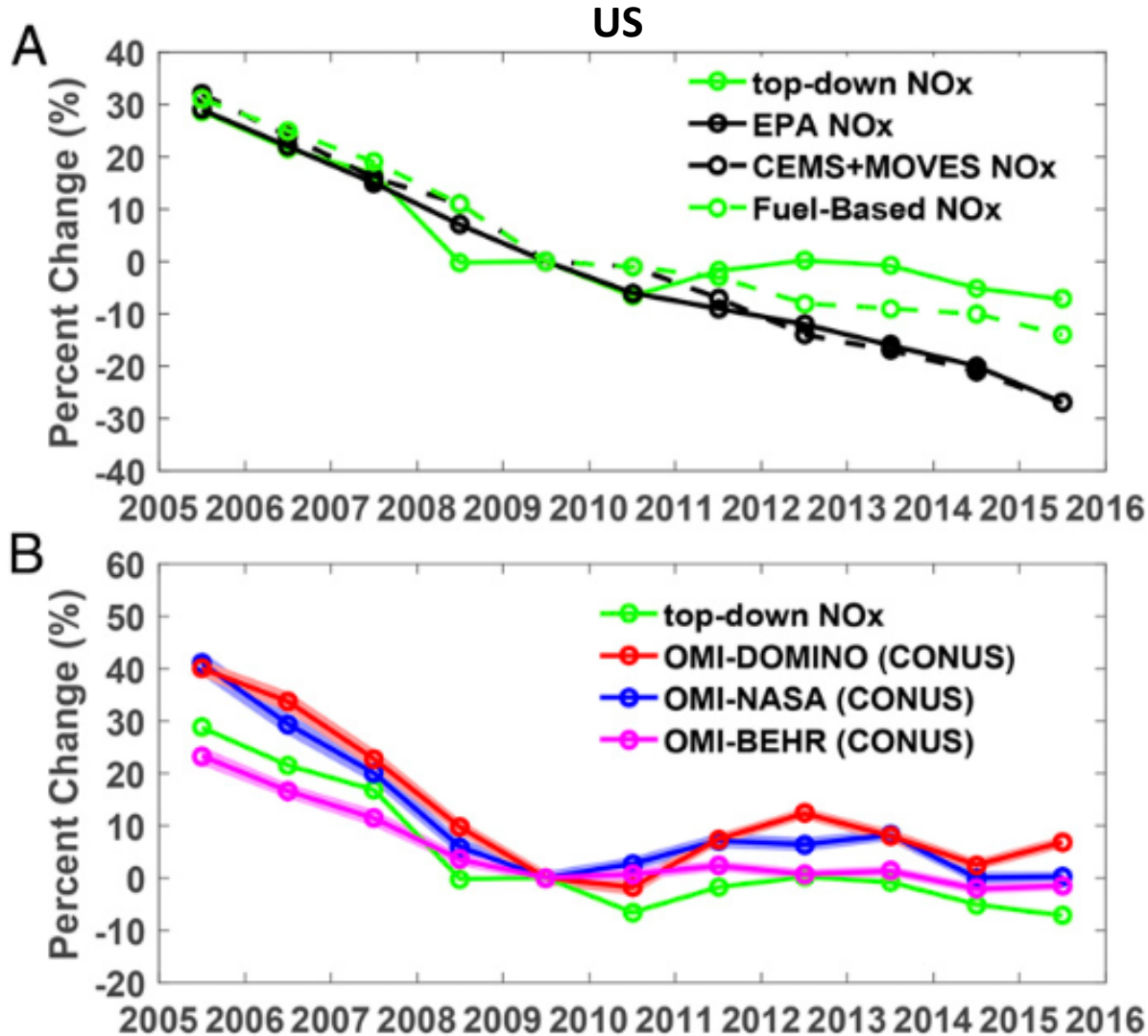


Daven K. Henze, Zhen Qu, Hansen Cao (University of Colorado, Boulder);
Zhe Jiang (USTC);

Helen Worden (NCAR); John Worden (JPL); Kazuyuki Miyazaki (JPL); Brian McDonald (NOAA); Ave Arellano (University of Arizona); Dylan B. A. Jones (University of Toronto); Emily Fischer (CSU); Liye Zhu (Sunyast-Sen U.); Folkert Boersma (KNMI); Zbigniew Klimont (IIASA); Mark Shephard (EC), Karen Cady-Pereira, Matt Alvarado, Chantelle Lonsdale (AER); Can Li (NASA Goddard); Nicolas Theys (BIRA)

Funding: NASA HAQAST, ACMAP, TASNPP

Declining trends in NO_x emissions have been decreasing, in contrast to inventories

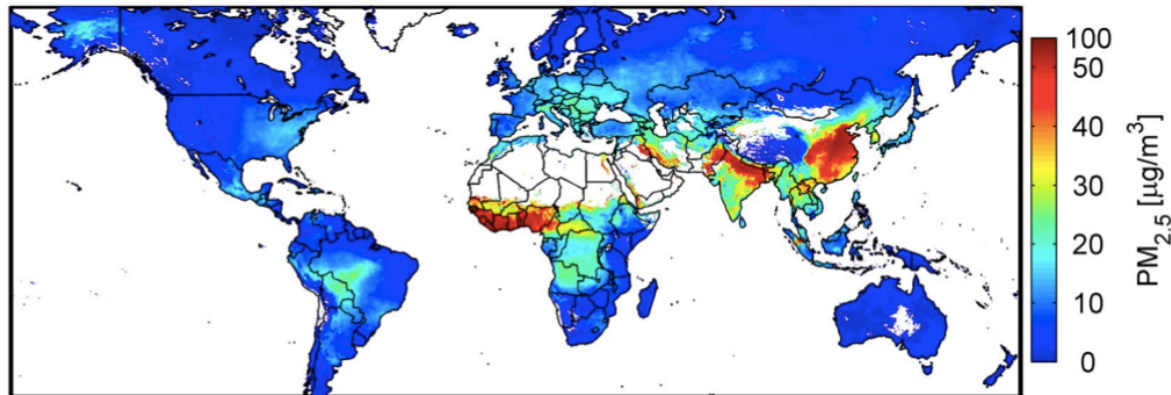


Inverse modeling and data assimilation

- Formally combining air quality model simulations with observational datasets
 - Plume models
 - Mass-balance
 - 3D-Var, 4D-Var, EnKF,....
- Provide a means of:
 - Updating concentration estimates (e.g., forecasting)
 - Improving emissions estimates (for forecasting, reanalysis, AQ planning)
- “4th generation”/ “instrumented models” can incorporate remote-sensing into efforts that are often entirely model-based:
 - Source attribution studies
 - Boundary conditions for regional modeling
 - Health impacts studies for species whose surface concentrations are hard to observe from satellites

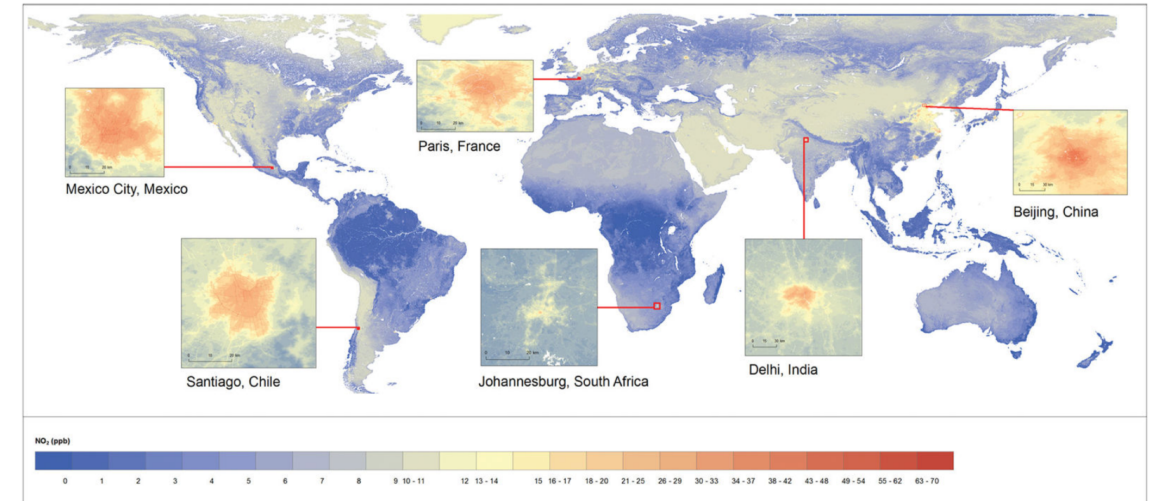
Role of chemical remote-sensing observations in health studies

PM_{2.5}



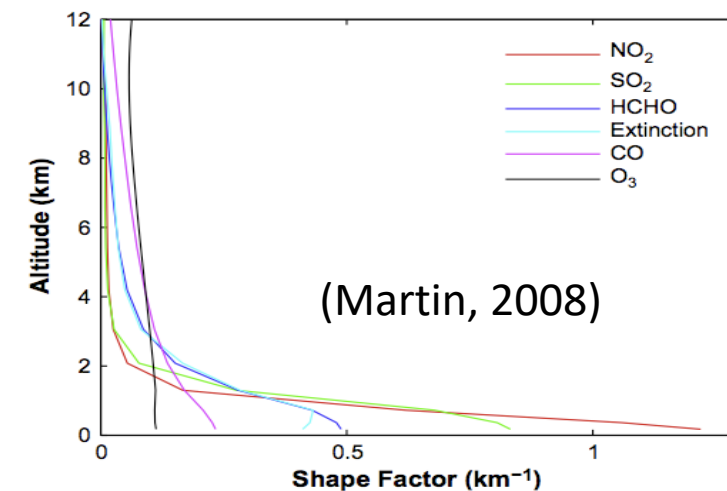
van Donkelaar et al., 2016: MODIS, MISR, SeaWiFS, CALIOP

NO₂



Larkin et al., 2017: SCIAMACHY, GOME-2

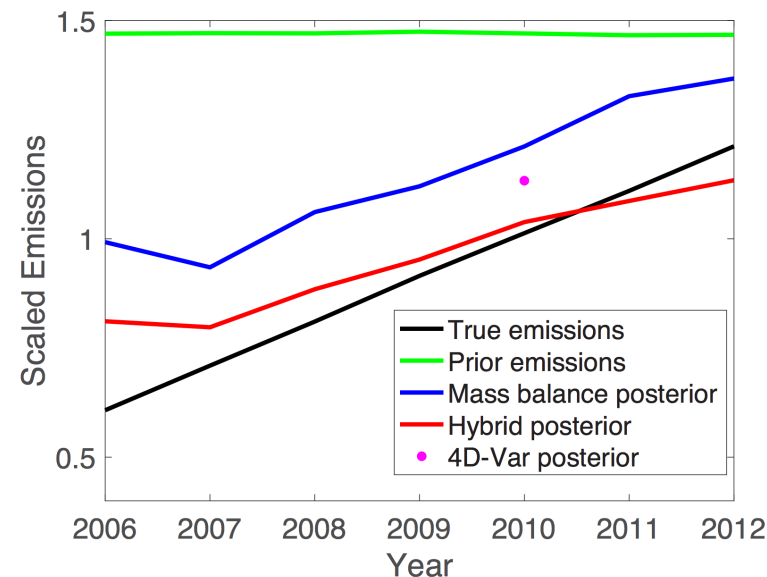
- Numerous studies of health impacts of PM_{2.5} and NO₂ rely on satellite-informed exposure estimates
- In contrast, O₃ health impact studies (e.g., GBD) have relied on models and surface monitoring networks



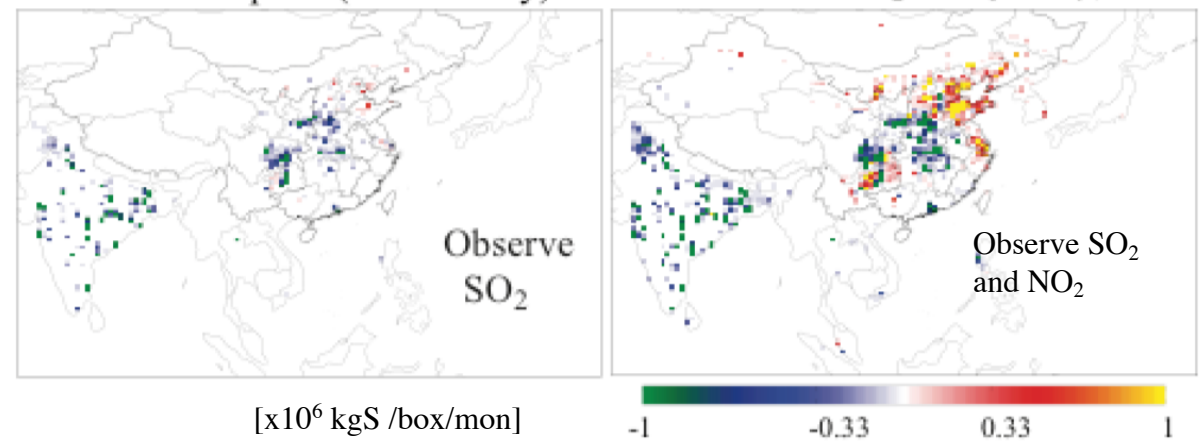
Ongoing inverse modeling efforts in our group (Zhen Qu)

- Develop hybrid mass-balance / 4DVar
 - Computationally efficient
 - Accurate
- Application to high-resolution emissions and trends of NO_x (Qu et al., 2017) and SO₂ (Qu et al., in press) using OMI satellite data
- Joint assimilation of NO₂, SO₂ (Qu et al., in press) and CO (in progress)
- Inversion method based on sectors (e.g., transportation, egu, residential, etc.) rather than species (in progress)

Testing new hybrid method with NO₂ pseudo inversion

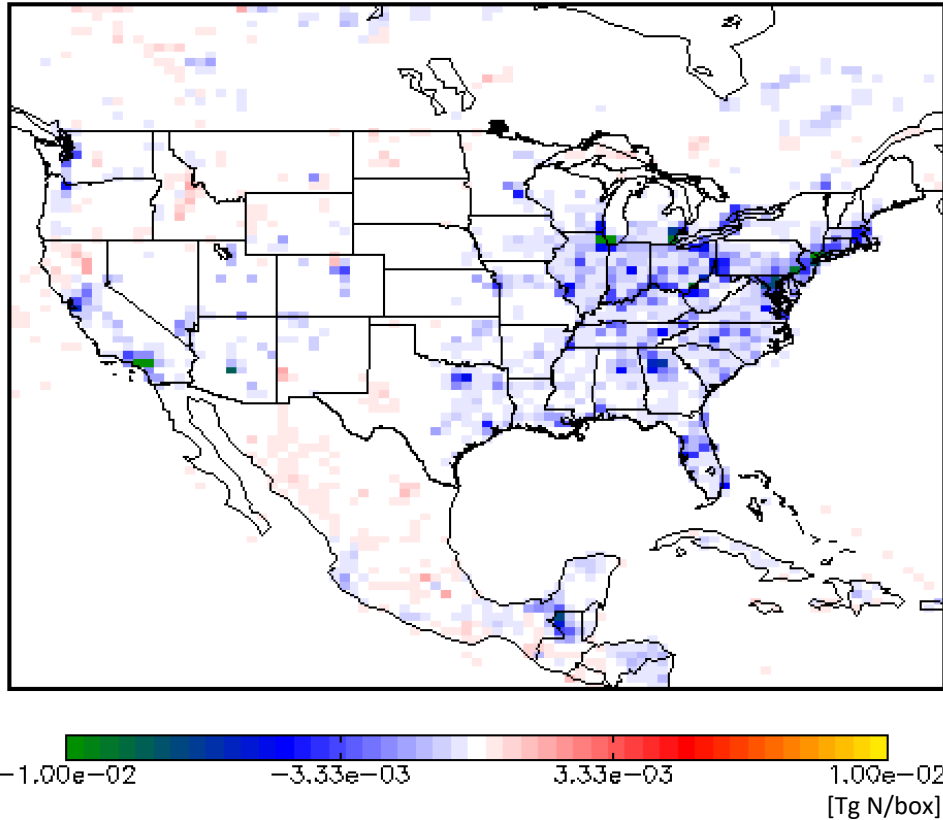


SO₂ emissions constrained by OMI SO₂ and NO₂
Posterior - prior (individually) Posterior - prior (jointly)

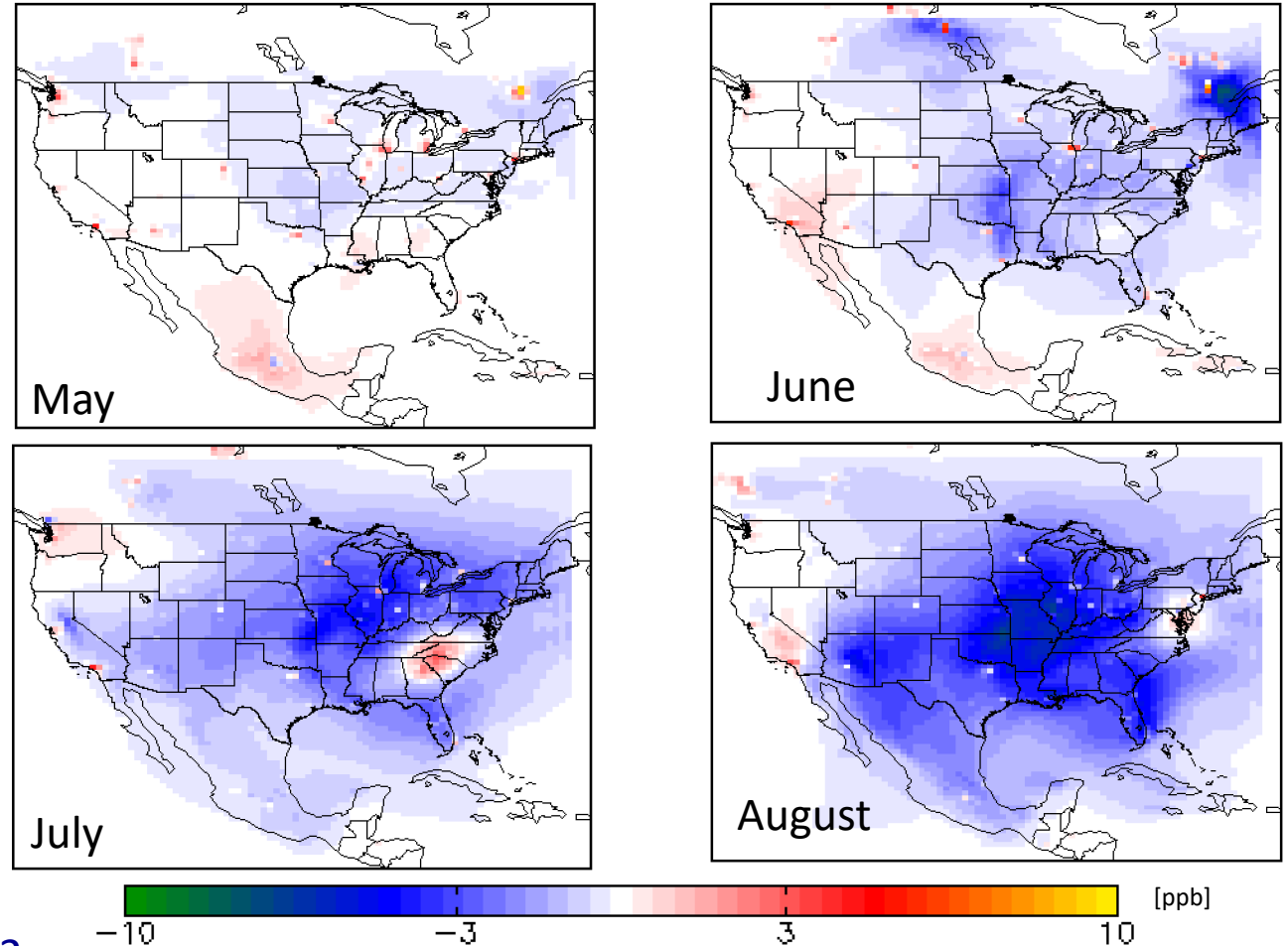


Inverse modeling of NO_x emissions in North America

2012 – 2005 annual NO_x emissions



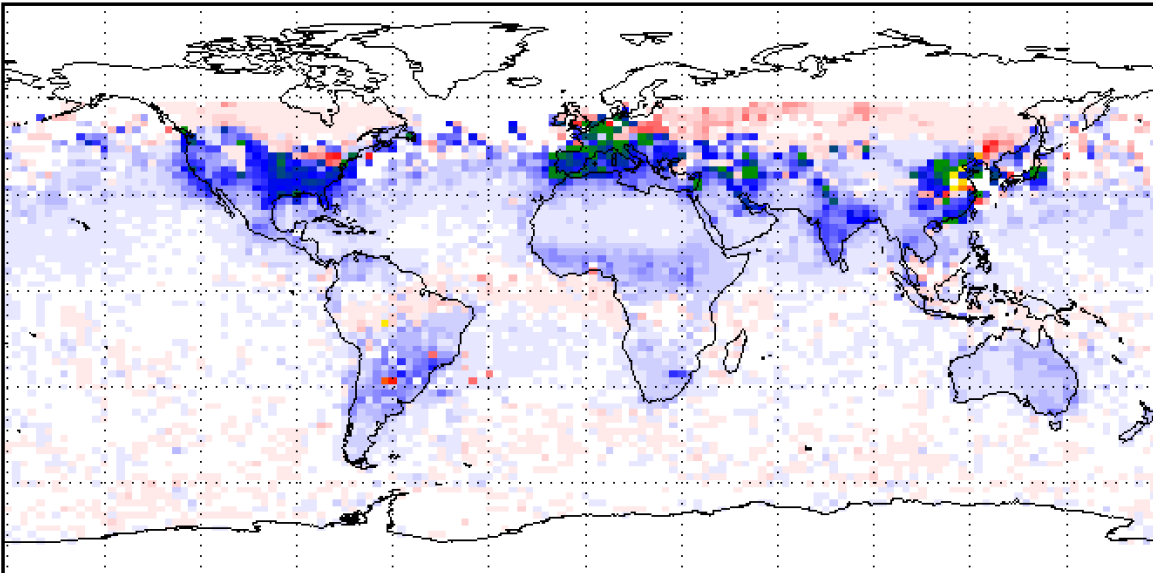
Surface 2010 O₃ concentration (posterior NO_x – prior NO_x)



- NO_x decrease by 30 - 40 % from 2005 to 2012 in EUS
- NO_x increase in Mexico by ~ 40 – 100% from 2005 - 2012
- NO_x emission overestimated in US bottom-up inventory in 2010
- Simulated O₃ are generally overestimated in US using HTAP 2010 emissions

Impact of assimilation depends on OMI NO₂ retrieval product and O₃ metric

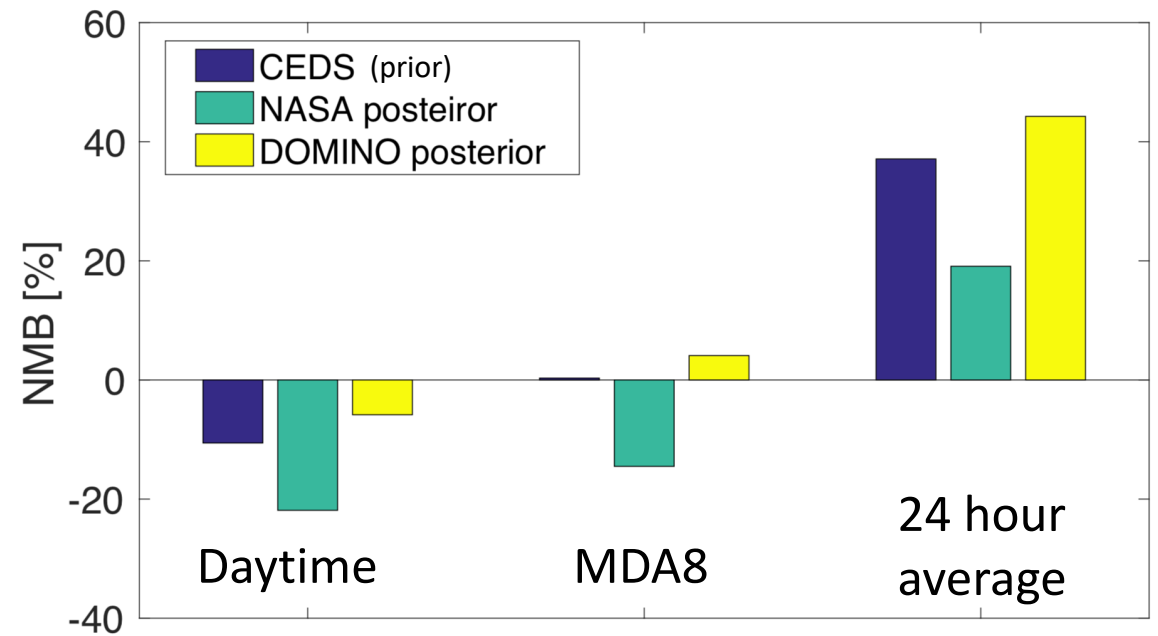
OMI VCD_{SP} - VCD_{DOMINO}, Jan 2015



NO₂ column densities from SP are ~50% smaller than that from DOMINO in densely populated and industrial regions.

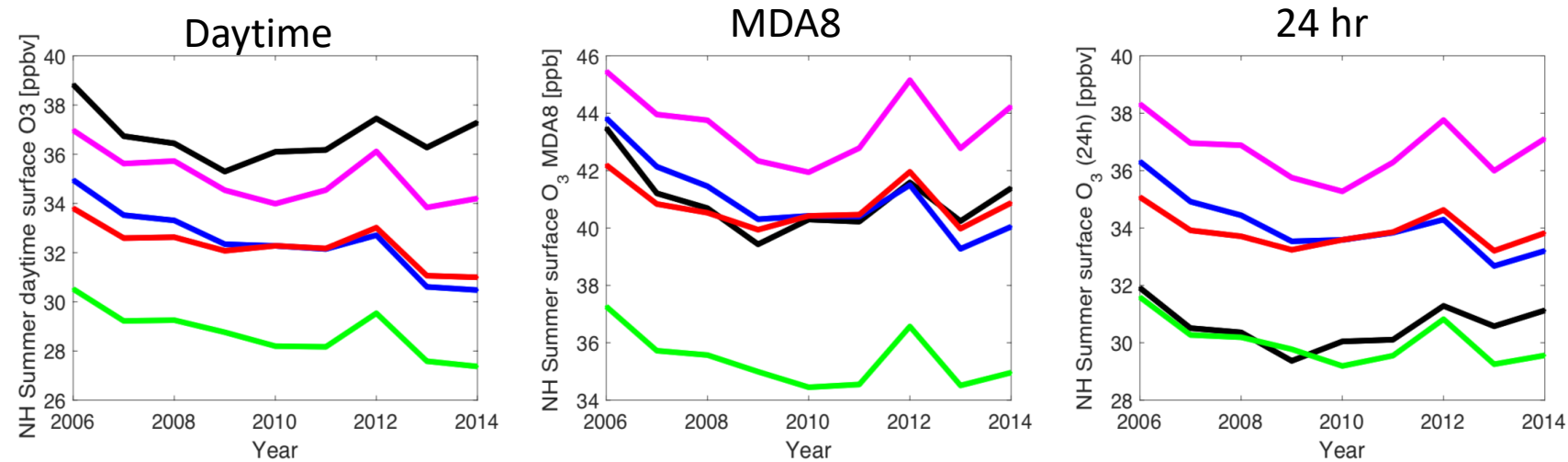
(Qu et al., 2017; Canty et al., 2015; Zheng et al., 2014)

Evaluation of global posterior emissions in GCv12 compared to O₃ (TOAR, 2019)



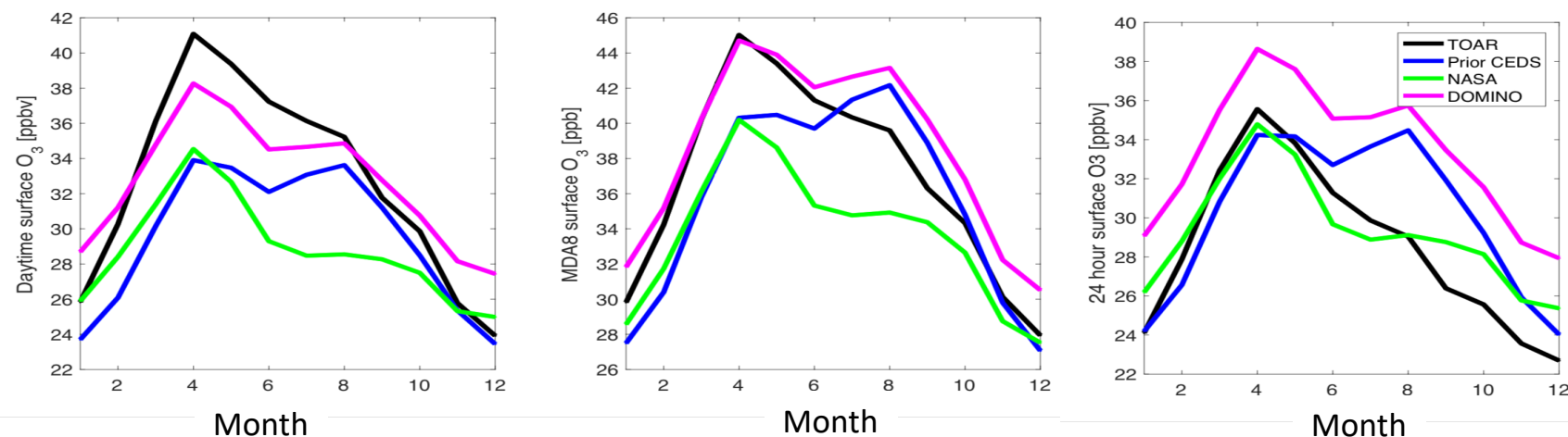
Differences across O₃ metrics highlights importance of upcoming geostationary (i.e., hourly) NO₂ satellite obs

Annual and seasonal trends are more consistently improved by OMI NO₂ inverse modeling



— TOAR — PRIOR — NASA — DOMINO

	Prior (CEDS)	NASA	DOMINO	
R	0.47	0.55	0.72	Daytime
R	0.81	0.85	0.89	MDA8
R	0.55	0.69	0.88	24 hr
NMB [%]	-11.57	-21.78	-4.56	Daytime
NMB [%]	0.22	-13.56	6.47	MDA8
NMB [%]	11.44	-1.85		20 24 hr



	Prior (CEDS)	NASA	DOMINO	
R	0.93	0.92	0.99	Daytime
R	0.90	0.98	0.97	MDA8
R	0.81	0.96	0.95	24 hr
NMB [%]	-9.68	-12.06	0.11	Daytime
NMB [%]	-3.25	-8.84	4.71	MDA8
NMB [%]	5.76	2.49	16.96	24 hr

Feeding Tiger Team activities

- Indicators (Anenberg)
 - trends in global O_3
 - to be included in ensemble of modeled O_3 simulations, weighted by accuracy vs surface obs, to generate O_3 exposure estimates (West)
 - will provide first satellite-informed O_3 exposures to GBD estimates of global O_3 health impacts
- Background O_3 (Neu)
 - evaluate metrics of remote, spring, Western US sites
 - provide boundary conditions for regional modeling
 - evaluate trends, source attribution in O_3 design values in Yuma (Qu et al., in prep)
- Regional Haze (Fiore)
 - evaluate role of emission trends



Year 3 Progress Update, PI Henze, Co-I's Anenberg, Kinney

Supporting health impact assessment tools using remote sensing and earth system models

- Source-receptor (SR) coefficients for PM_{2.5} and O₃ to CCAC impact toolkit (LEAP-IBC)
 - PM_{2.5} SR coefficients for all countries in Asia, Africa, Caribbean, Central & South America.
 - O₃ coefficients for each CCAC member nation
 - LEAP-IBC used in 25 countries, e.g. Nepal (Nakarmi et al., submitted) and Bangladesh (Kuylensstierna et al., in prep)
 - Impacts of cookstoves (Lacey et al., 2017) and fugitive diesel NO_x (Anenberg et al., 2017)
- High-resolution SR calculations for urban regions
 - In progress for Africa, West Africa, South America
- Expand health impacts
 - Global air pollution and asthma (Anenberg et al., 2018), comparisons across cities with eCO₂ (Anenberg et al., submitted)

*Year 3: 4 papers published, 2 accepted
4 in review, 4 in prep*

Tiger Team Participation

- Hi-Res PM_{2.5} (**Kinney**, Freedman)
 - Health impacts of different PM_{2.5} estimates (**Anenberg**)
 - Low-cost monitors deployed in NYC, Boston, SF (**Kinney**)
- Trends (Duncan, West) / Background O₃ (Neu) / Regional Haze (Fiore)
 - SO₂ and NO_x emission from OMI (Que et al., 2017; 2019; 2019; Zhang et al., PNAS, 2018)
 - O₃ source apportionment in Yuma, Az (Qu et al., in prep)
 - O₃ estimates from NO_x assimilation for boundary conditions
- Fires (O'Neill, **Kinney** Co-I)
 - Health impacts of 2017 Northern CA Fires
- Indicators (**Anenberg** PI)
 - Global O₃ concentrations from OMI-NO₂ assimilation results to contribute to statistically fused product (West) for GBD