

Connecting air quality and health with management: Progress from the UNC HAQAST Team

J. Jason West

Environmental Sciences & Engineering
University of North Carolina, Chapel Hill

west.web.unc.edu

@ProfJasonWest

Global Mapping of Ozone Surface Concentration for Global Burden of Disease

- A first global map has been delivered and used in GBD 2017, another will be delivered next month for GBD 2019.
- 2 papers (Chang, GMD).

Global Air Quality and Health Co-benefits of the Paris Agreement Pledges

- Completing analysis of emissions from GCAM and planning to start simulations this summer.

Tiger Team: Efficacy of Environmental Regulations in the Eastern US

Trends in US air pollution-related deaths since 1990

- 1 paper published (Zhang, ACP), 1 in preparation.

Tiger Team: California fires

Mapping pollutant concentrations, and analyzing health impacts.

Tiger Team: Global indicators

Global indicators of surface ozone.

Total 4 papers published, many more in preparation.

Mapping Global Surface Ozone Concentrations

Goal: Estimate global surface ozone concentrations by statistically fusing global ozone observations and an ensemble of global models.

Stakeholder partners: Global Burden of Disease Assessment – Michael Brauer (UBC), Rick Burnett (Health Canada), Bryan Hubbell (EPA).

Team: Jason West, Marc Serre, Marissa Delang, Jacob Becker, Stephanie Cleland, Elyssa Collins (UNC), Owen Cooper, Kai-Lan Chang (U Colorado & NOAA)



A new method (M³Fusion v1) for combining observations and multiple model output for an improved estimate of the global surface ozone distribution

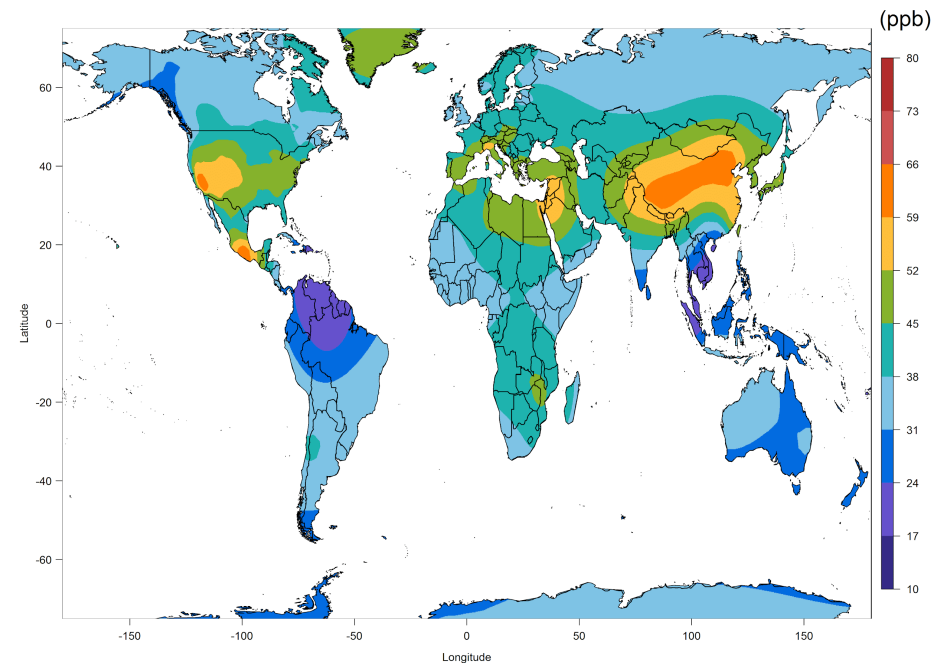
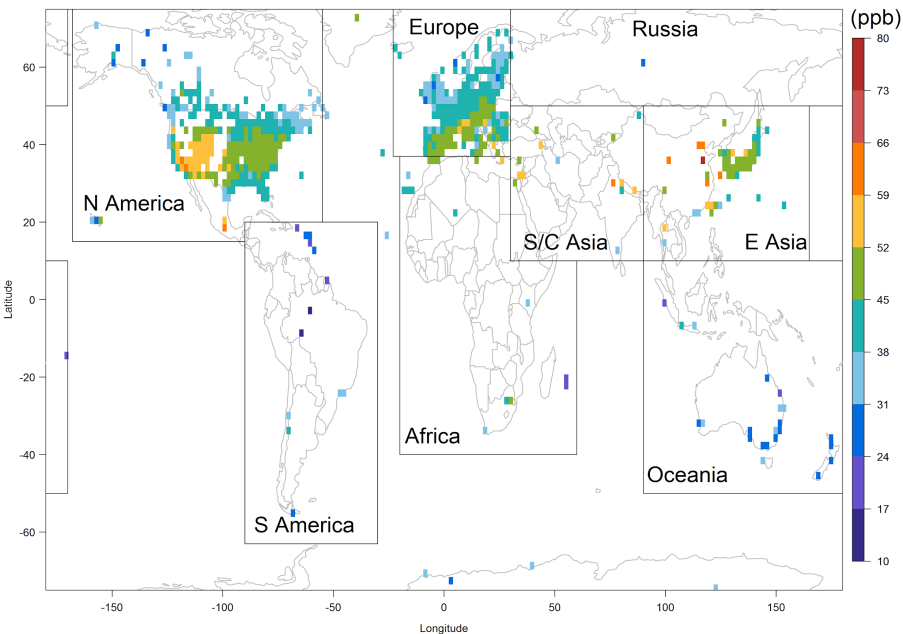
Kai-Lan Chang^{1,2,3}, Owen R. Cooper^{2,3}, J. Jason West⁴, Marc L. Serre⁴, Martin G. Schultz⁵, Meiyun Lin^{6,7}, Virginie Marécal⁸, Béatrice Josse⁸, Makoto Deushi⁹, Kengo Sudo^{10,11}, Junhua Liu^{12,13}, and Christoph A. Keller^{12,13,14}

Ozone metric: 2008-2014 average of 6-month average 8-hr.
daily maximum surface ozone concentration

Mapping Global Surface Ozone Concentrations for GBD 2017

Step 1 – Spatial interpolation of TOAR measurements

4801 sites, averaged within 2°x2° grid cells

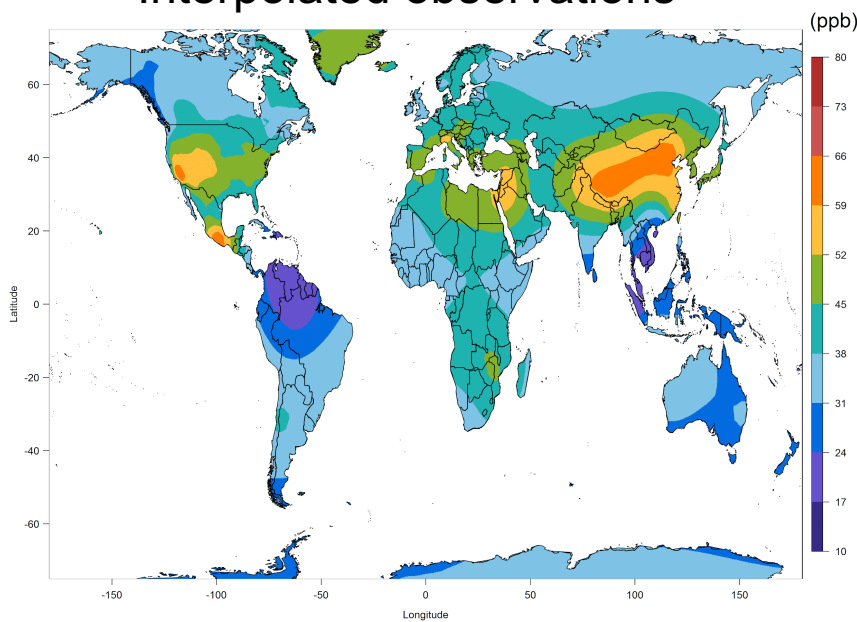


Mapping Global Surface Ozone Concentrations for GBD 2017

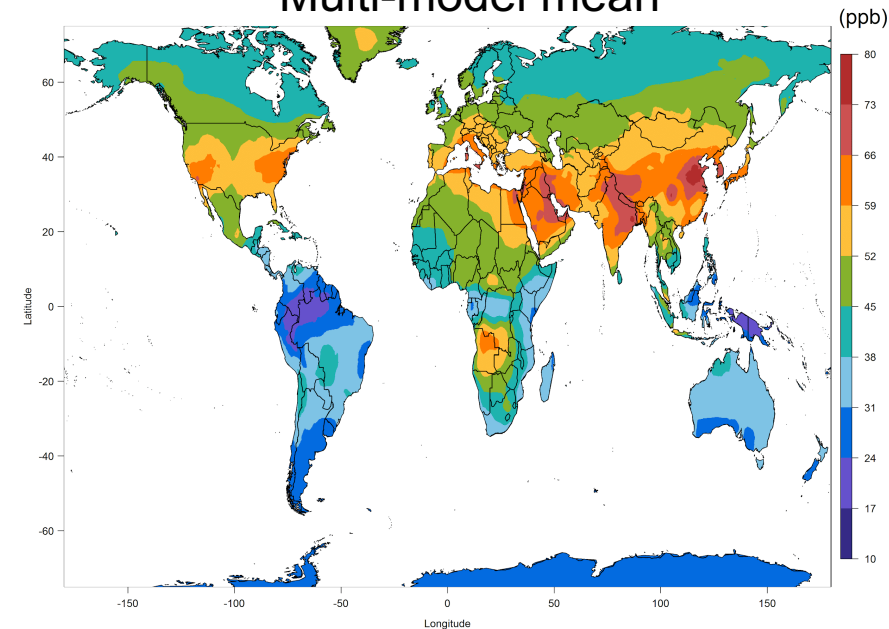
Step 2 – Evaluate each model with respect to observations

Model ID	Group	Resolution	Meteorological Forcing [†]	References
CHASER (MIROC-ESM)	Nagoya University; Japan Agency for Marine-Earth Science and Technology (JAMSTEC), Japan	$2.8^{\circ} \times 2.8^{\circ}$	C2	Sudo et al. (2002a, b); Watanabe et al. (2011)
GEOSCCM	NASA Goddard Space Flight Center, USA	$2.5^{\circ} \times 2^{\circ}$	C2	Oman et al. (2011)
GFDL-AM3	NOAA Geophysical Fluid Dynamics Laboratory, USA	$2^{\circ} \times 2^{\circ}$	C1SD	Lin et al. (2012, 2014, 2017)
G5NR-Chem	NASA Goddard Space Flight Center, USA	$0.125^{\circ} \times 0.125^{\circ}$	*	Hu et al. (2018)
MOCAGE	Centre National de Recherches Météorologiques; Météo France, France	$2^{\circ} \times 2^{\circ}$	C2	Josse et al. (2004); Teyssède et al. (2007)
MRI-ESM1r1	Meteorological Research Institute, Japan	$2.8^{\circ} \times 2.8^{\circ}$	C2	Adachi et al. (2013)

Interpolated observations



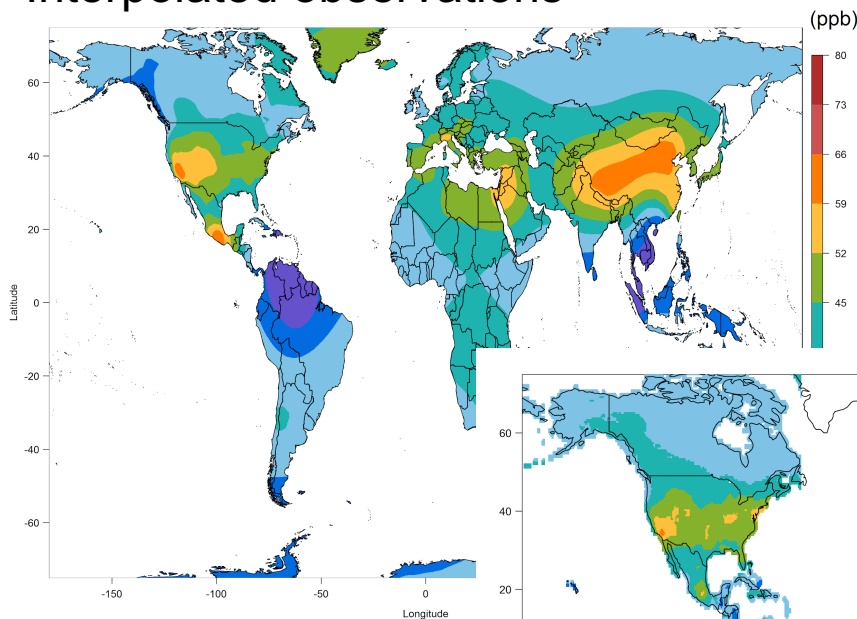
Multi-model mean



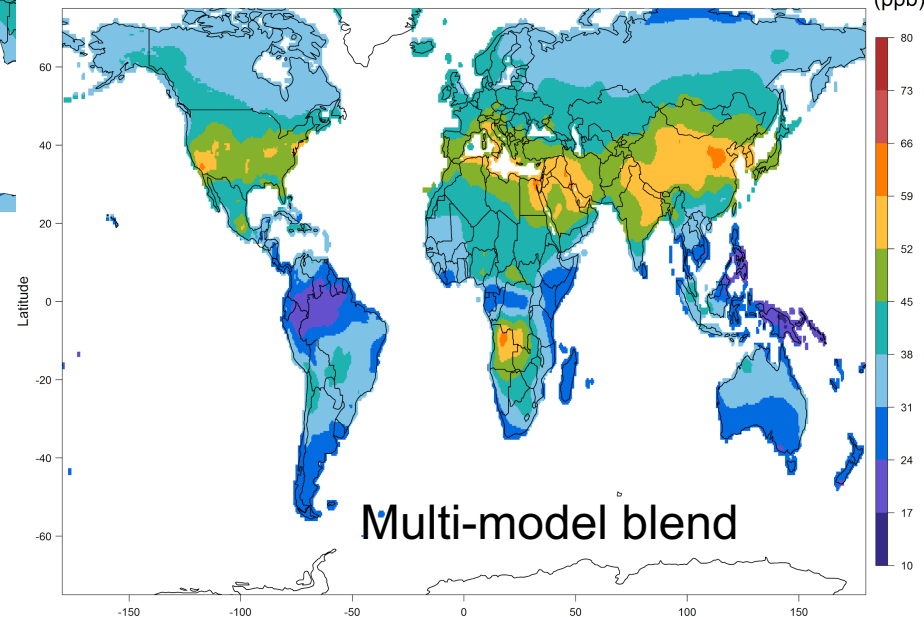
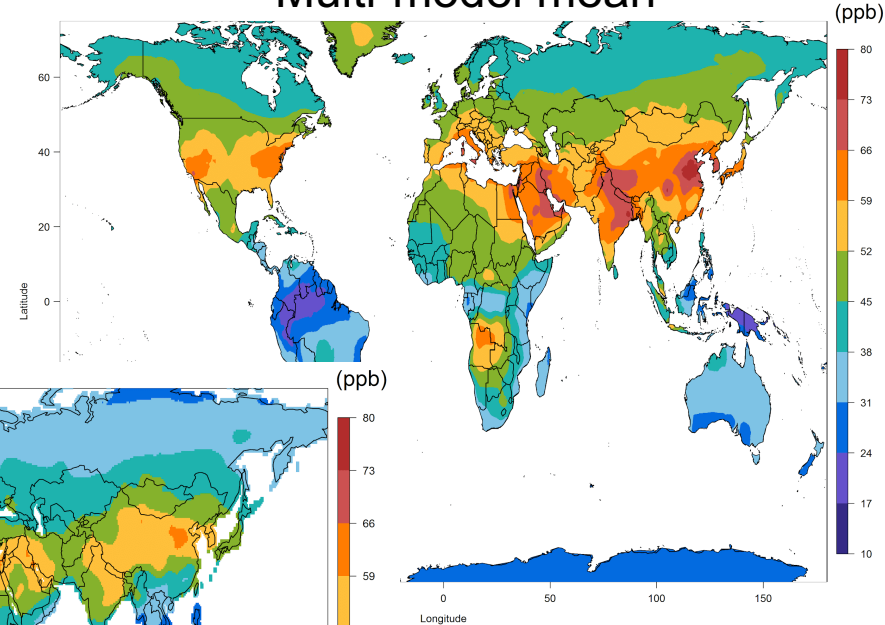
Mapping Global Surface Ozone Concentrations for GBD 2017

Step 3 – Find the linear combination of models in each world region that minimizes error with respect to the interpolated measurement surface

Interpolated observations



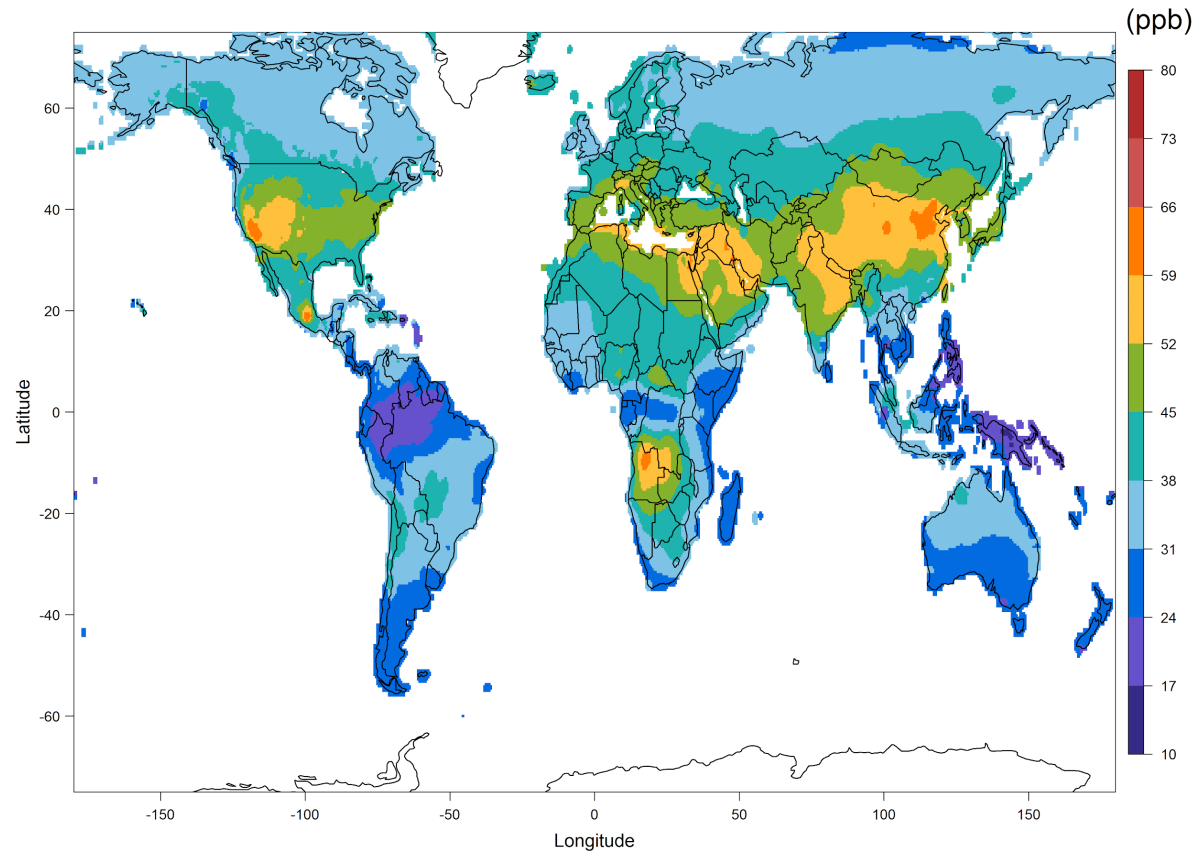
Multi-model mean



Mapping Global Surface Ozone Concentrations for GBD 2017

Step 4 – Bias correct within 2° of observation sites (using interpolated surface)

Final Fused Ozone Product

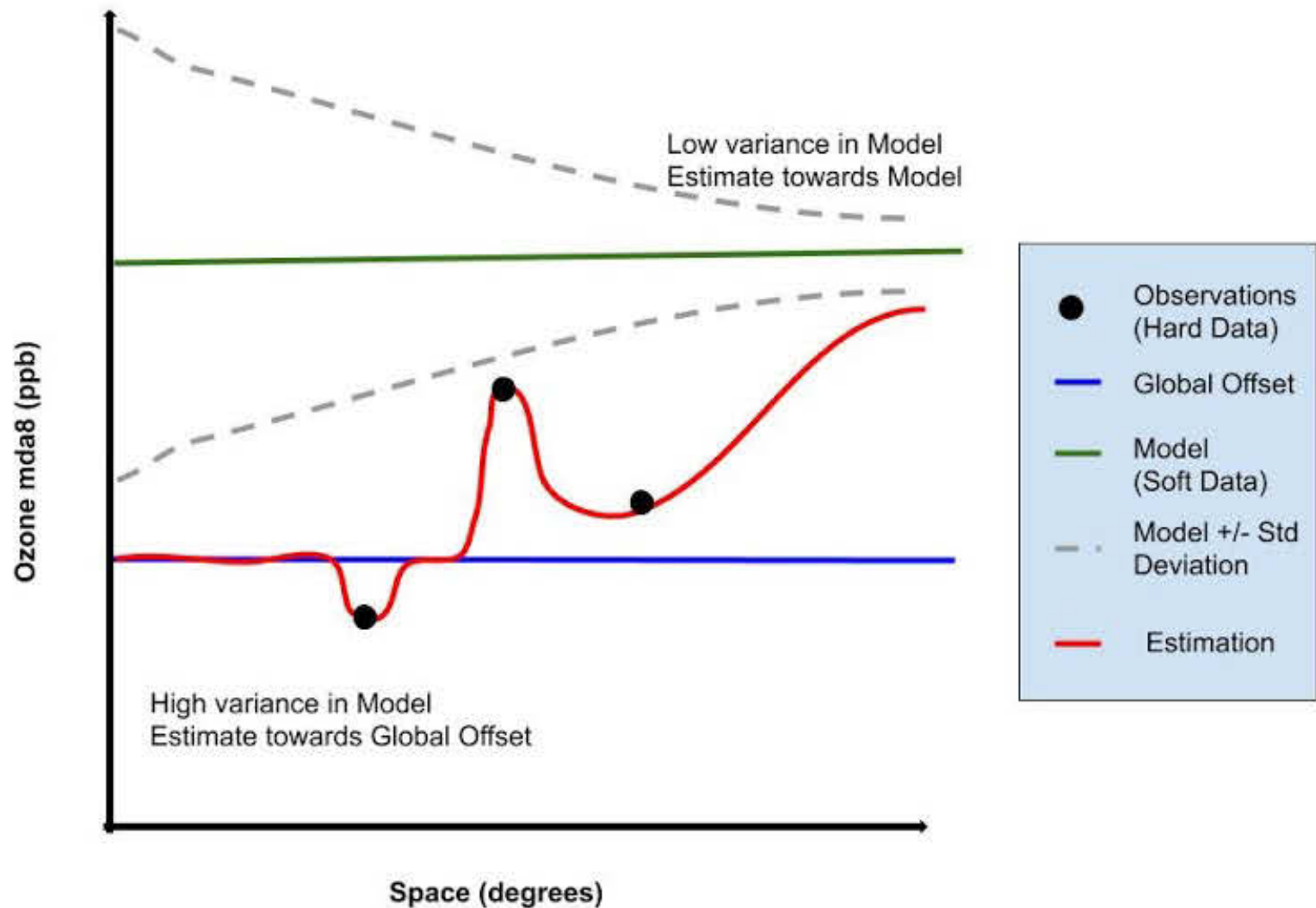


- GBD 2017 – 470,000 (180,000 – 770,000) deaths globally

Global Ozone Mapping for GBD 2019

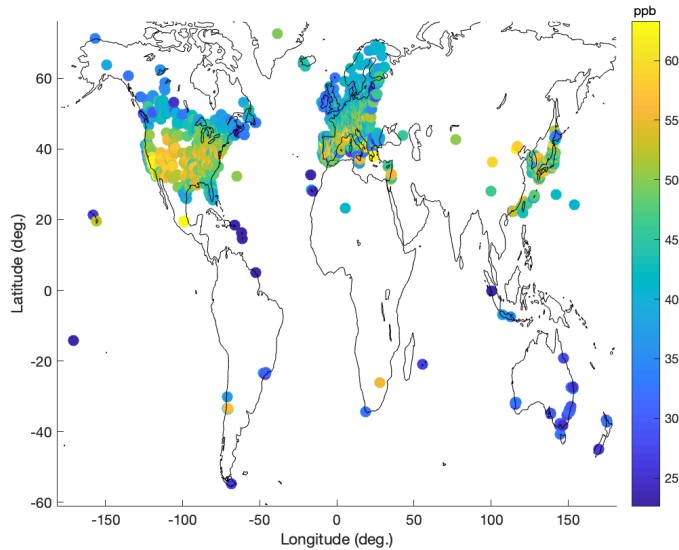
- Produce ozone maps for all years from 1990 to 2017.
 - Perform a new data fusion with observations using Bayesian Maximum Entropy method.
 - Add new observations from China and elsewhere, updated models.
 - Add fine spatial structure using NASA G5NR (0.125°).
-
- New maps have nearly been completed for 1990 to 2010.
 - See poster by Jacob Becker and Marissa DeLang

Bayesian Maximum Entropy Methods

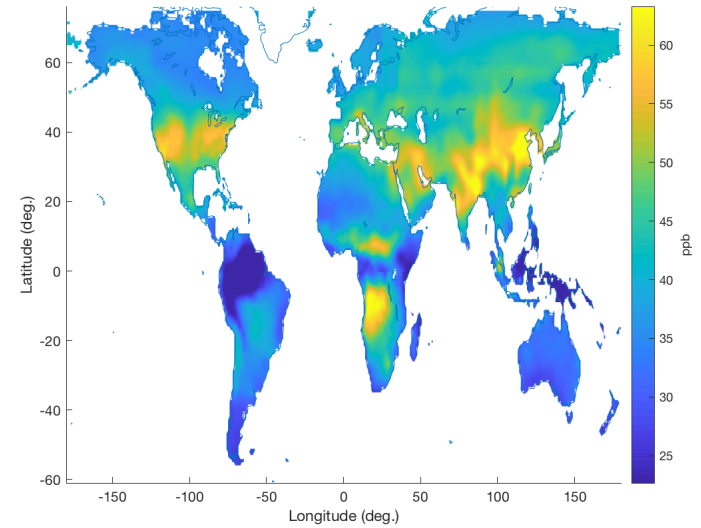


Global Ozone Mapping for GBD 2019

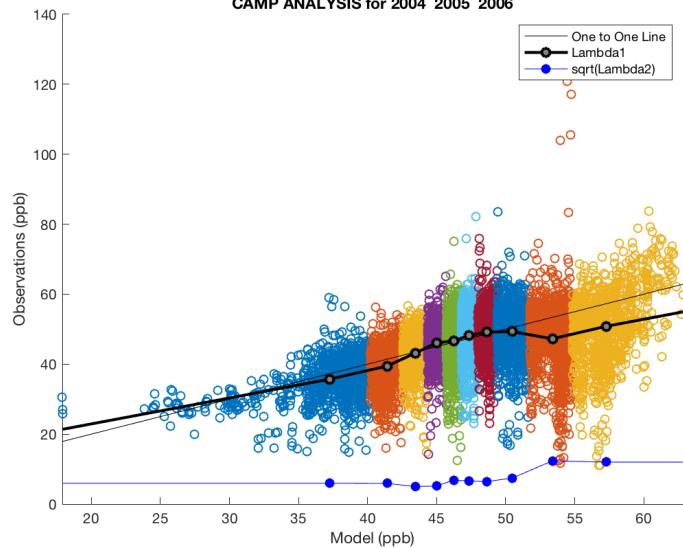
Observations for 2005



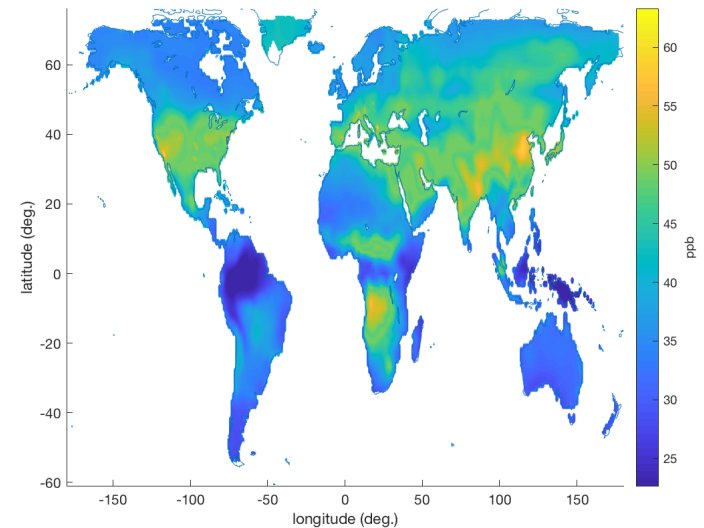
M³Fusion for 2005



CAMP ANALYSIS for 2004 2005 2006



BME Estimate for 2005



MAPPING THE AIR QUALITY & HEALTH IMPACTS OF THE 2017 CALIFORNIA WILDFIRES



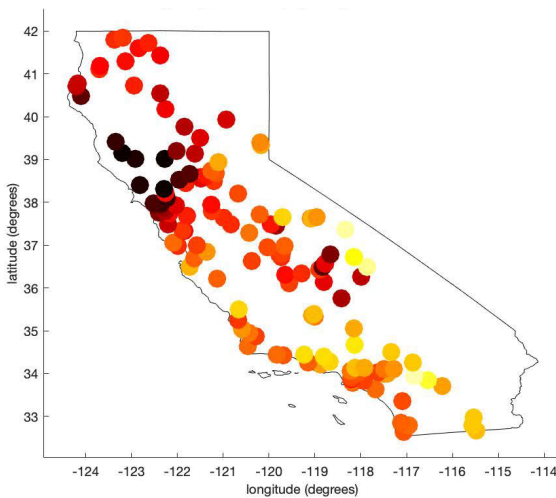
UNC
GILLINGS SCHOOL OF
GLOBAL PUBLIC HEALTH
DEPARTMENT OF ENVIRONMENTAL
SCIENCES AND ENGINEERING

Stephanie Cleland, Marc Serre, Jacob Becker, Marissa DeLang, & Jason West

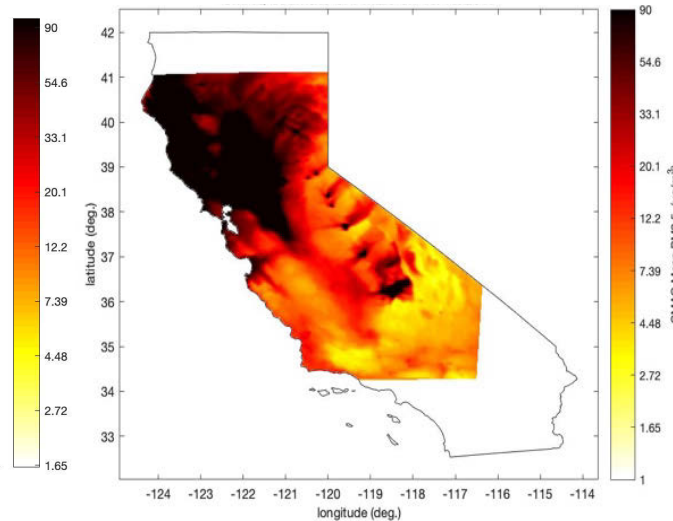
Goal 1: Map $PM_{2.5}$ during the Oct. 2017 wildfires, fusing together observed & modeled $PM_{2.5}$ concentrations

- Surface observation data from both FRM/FEM & temporary monitoring stations
- Model data from BAAQMD CMAQ run

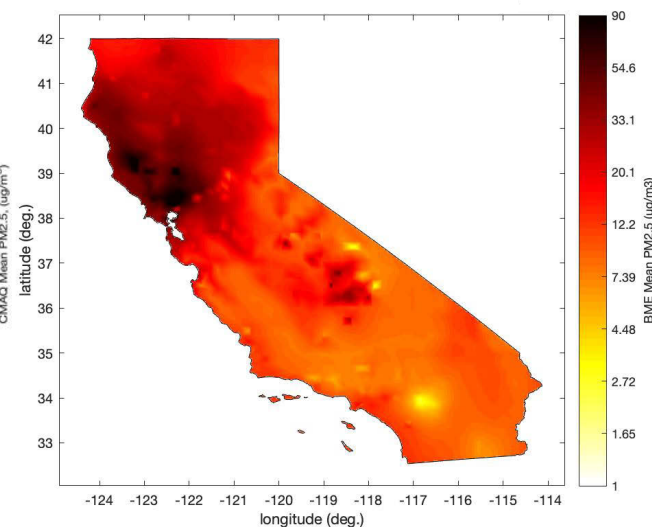
Observed $PM_{2.5}$



CMAQ Model $PM_{2.5}$



BME Data Fusion



maps for Oct. 10, 2017, 24-hr average $PM_{2.5}$

MAPPING THE AIR QUALITY & HEALTH IMPACTS OF THE 2017 CALIFORNIA WILDFIRES

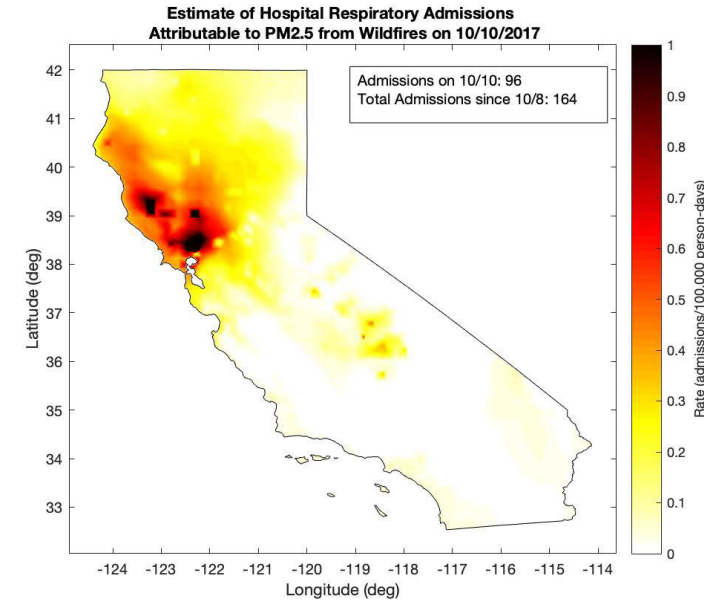


UNC
GILLINGS SCHOOL OF
GLOBAL PUBLIC HEALTH
DEPARTMENT OF ENVIRONMENTAL
SCIENCES AND ENGINEERING

Stephanie Cleland, Marc Serre, Jacob Becker, Marissa DeLang, & Jason West

Goal 2: Use the $PM_{2.5}$ map to estimate the acute health impact of the Oct. 2017 wildfires, specifically the attributable respiratory hospital admissions

Future work will extend this approach to more health endpoints and pollutants.



- See poster by Stephanie Cleland

Health Benefits of Decreases in PM_{2.5} and Ozone in the United States, 1990-2015

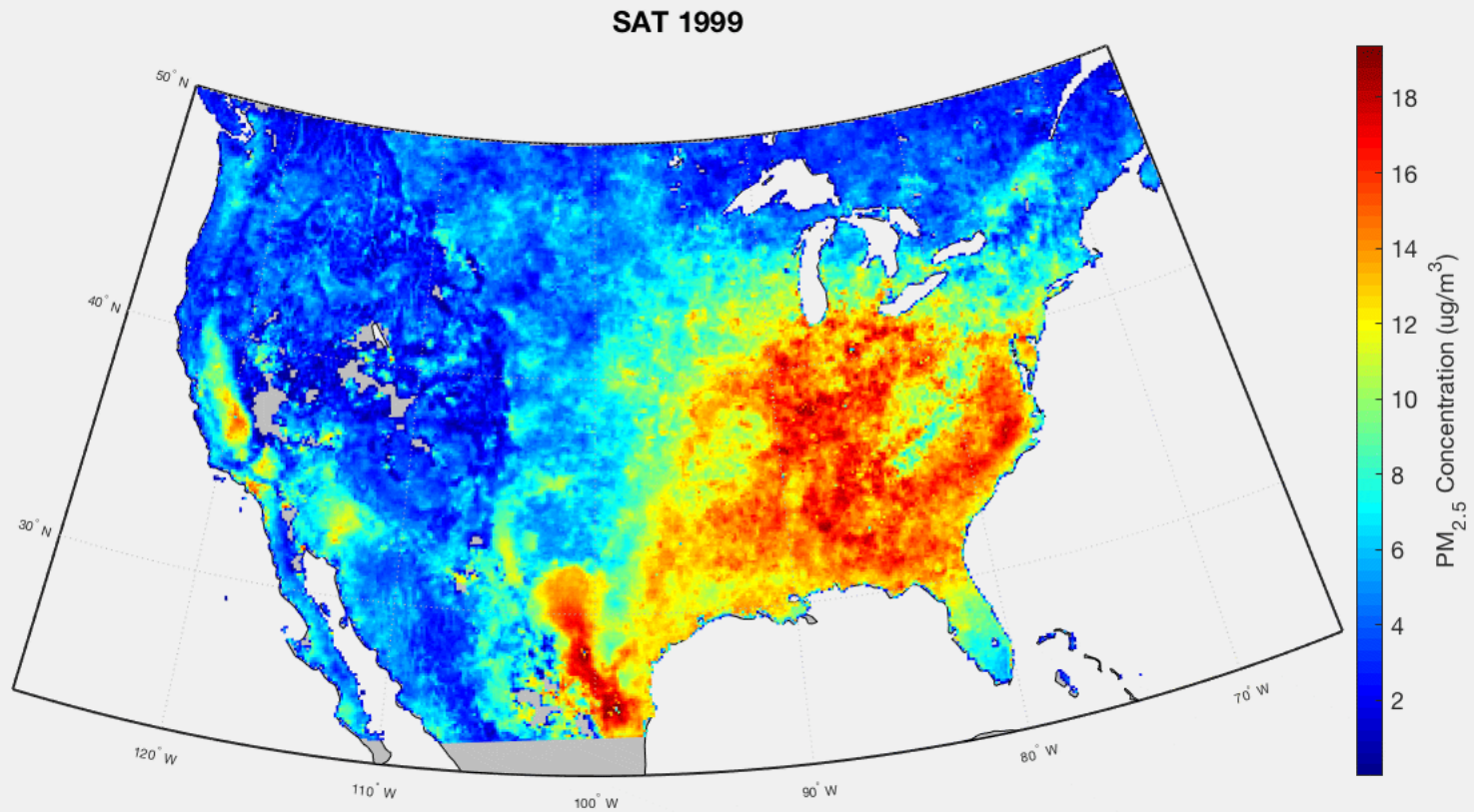
Omar Nawaz, Yuqiang Zhang, Daniel Q. Tong, Aaron van Donkelaar, Randall Martin, J. Jason West

* Air pollutant datasets:

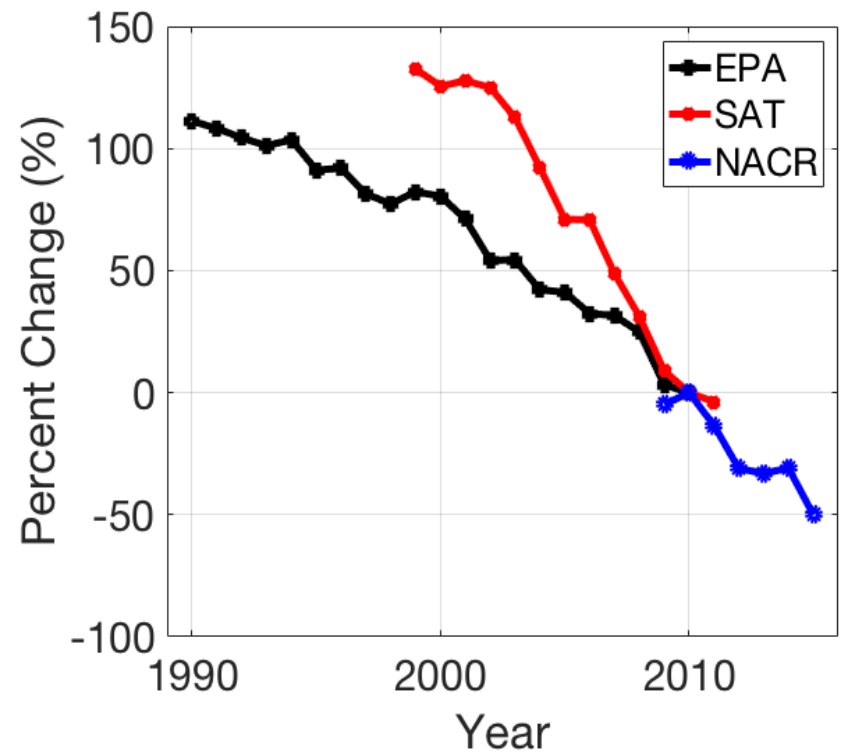
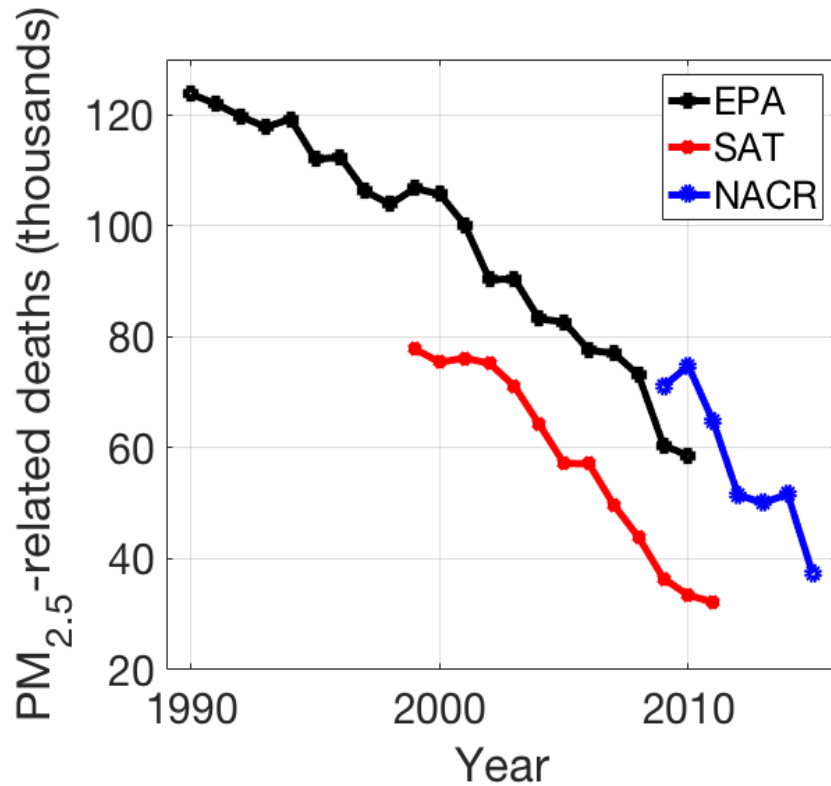
- A 21-year CMAQ simulation (1990-2011) [EPA](#)
- The North American Chemical Reanalysis (2009-2015) [NACR](#)
- N. America PM_{2.5} satellite-derived data combined with a model and surface observations (1999-2012) [SAT](#)

* We use annual county-level population and baseline cause-specific mortality rates from the CDC to assess air pollution mortality in each year.

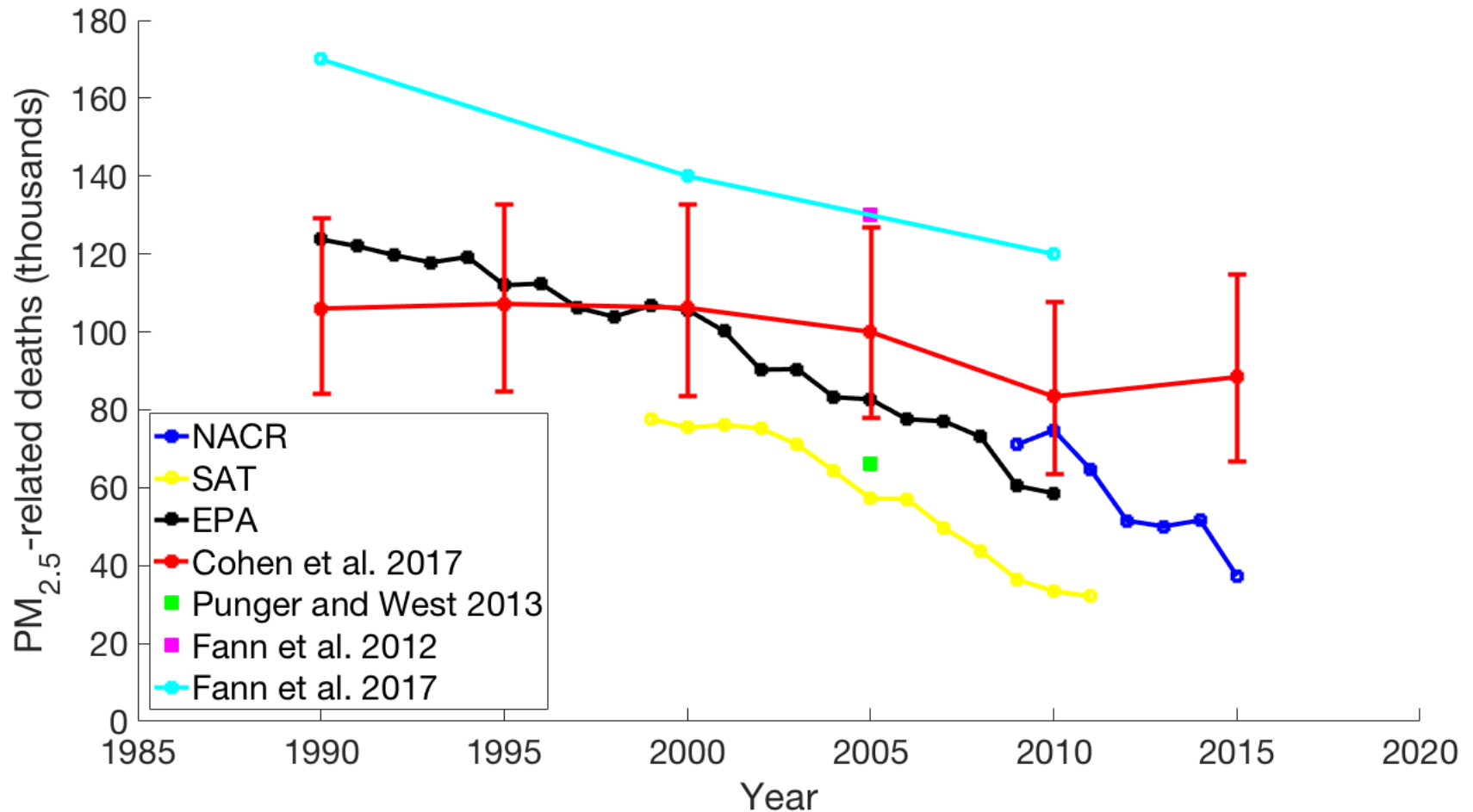
Trends in PM_{2.5} (SAT)



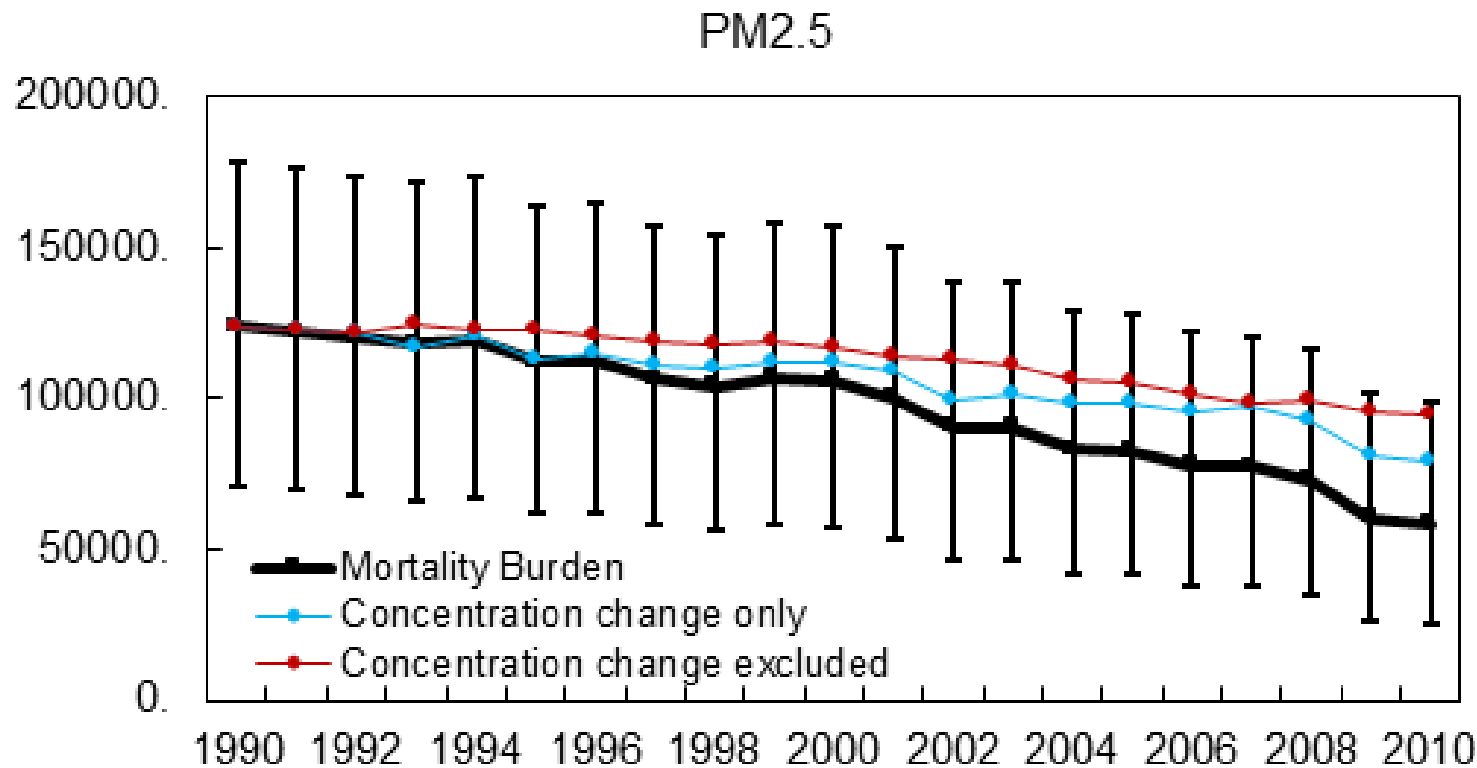
US PM_{2.5}-related deaths



Comparison with Other Studies (PM_{2.5})

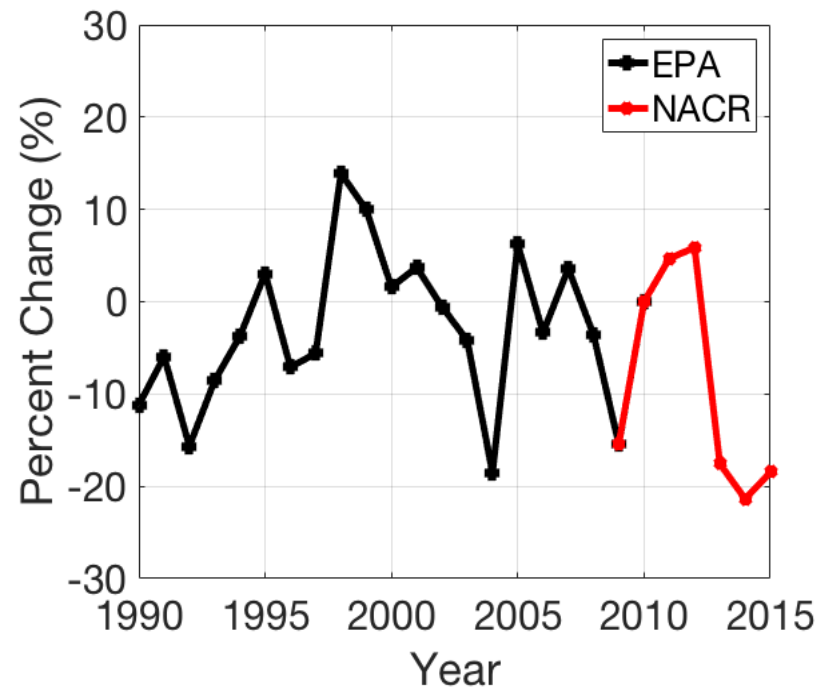
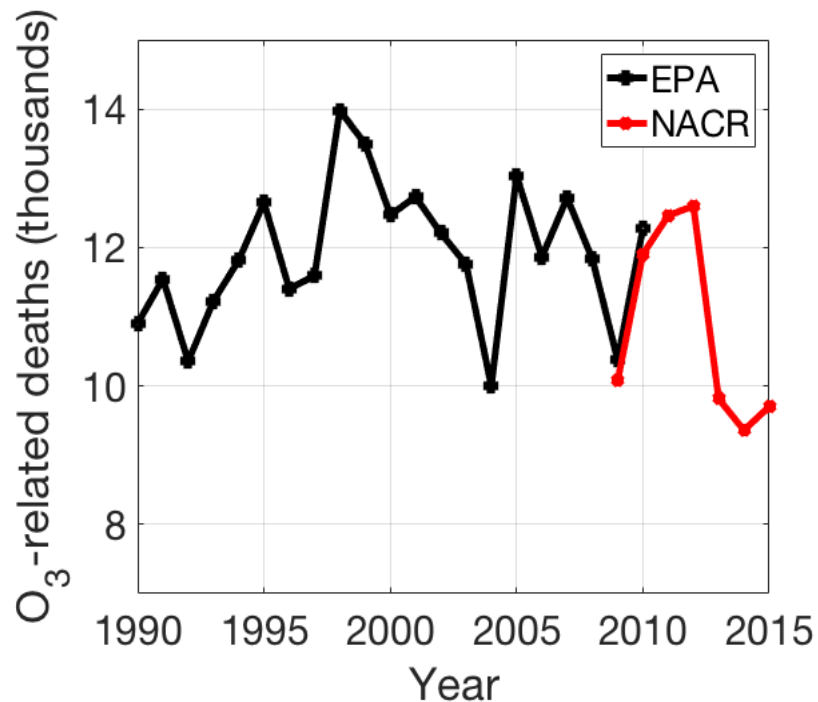


PM_{2.5} Mortality Burden (EPA)

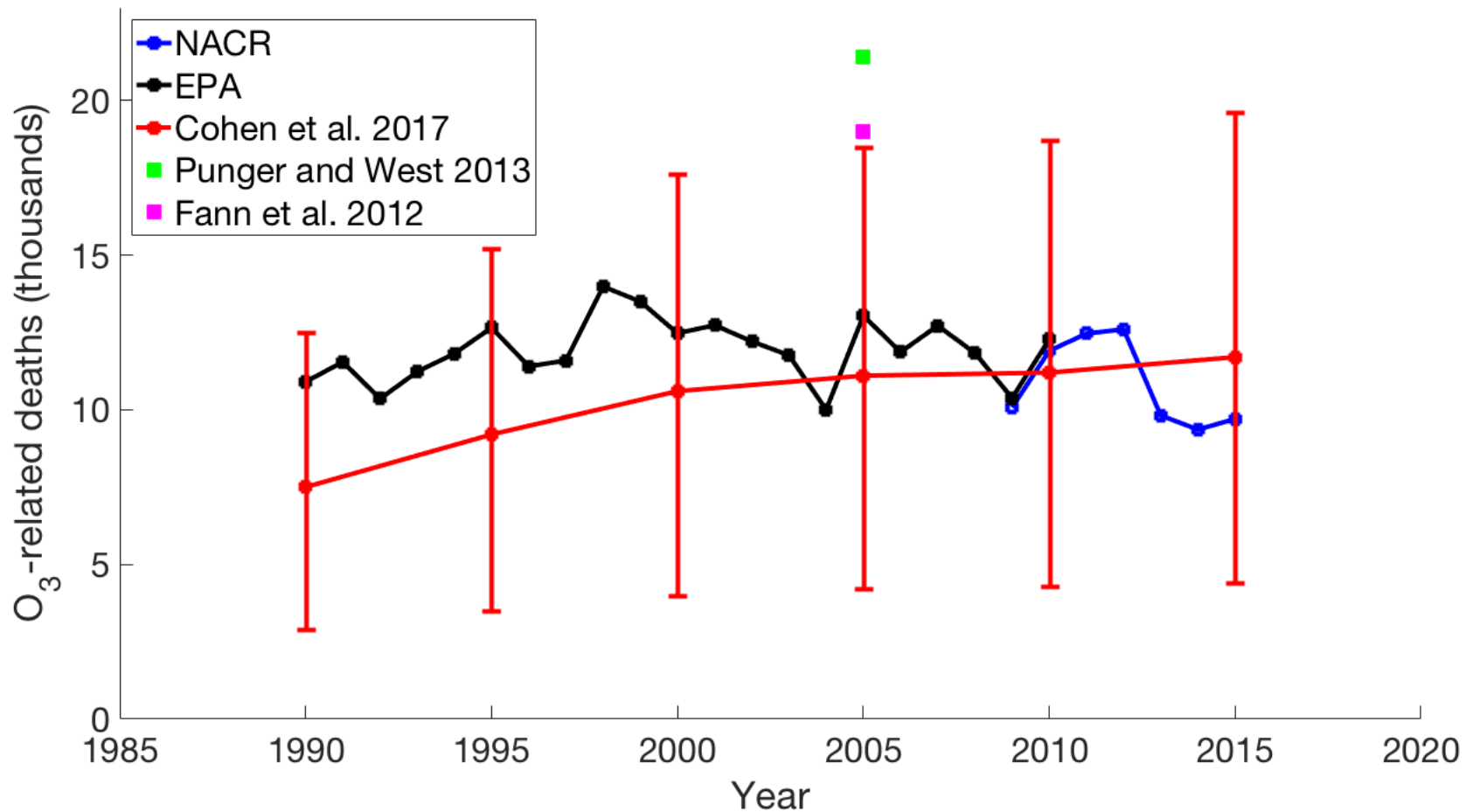


- PM_{2.5} mortality decreased by 53% from 123,700 (70,800-178,100) deaths in 1990 to 58,600 (24,900-98,500) in 2010.
- Without the decrease in PM_{2.5} since 1990, the burden would have only decreased by 24%.
- PM_{2.5} reductions since 1990 have decreased deaths in 2010 by about **35,800**.

US O₃-related deaths

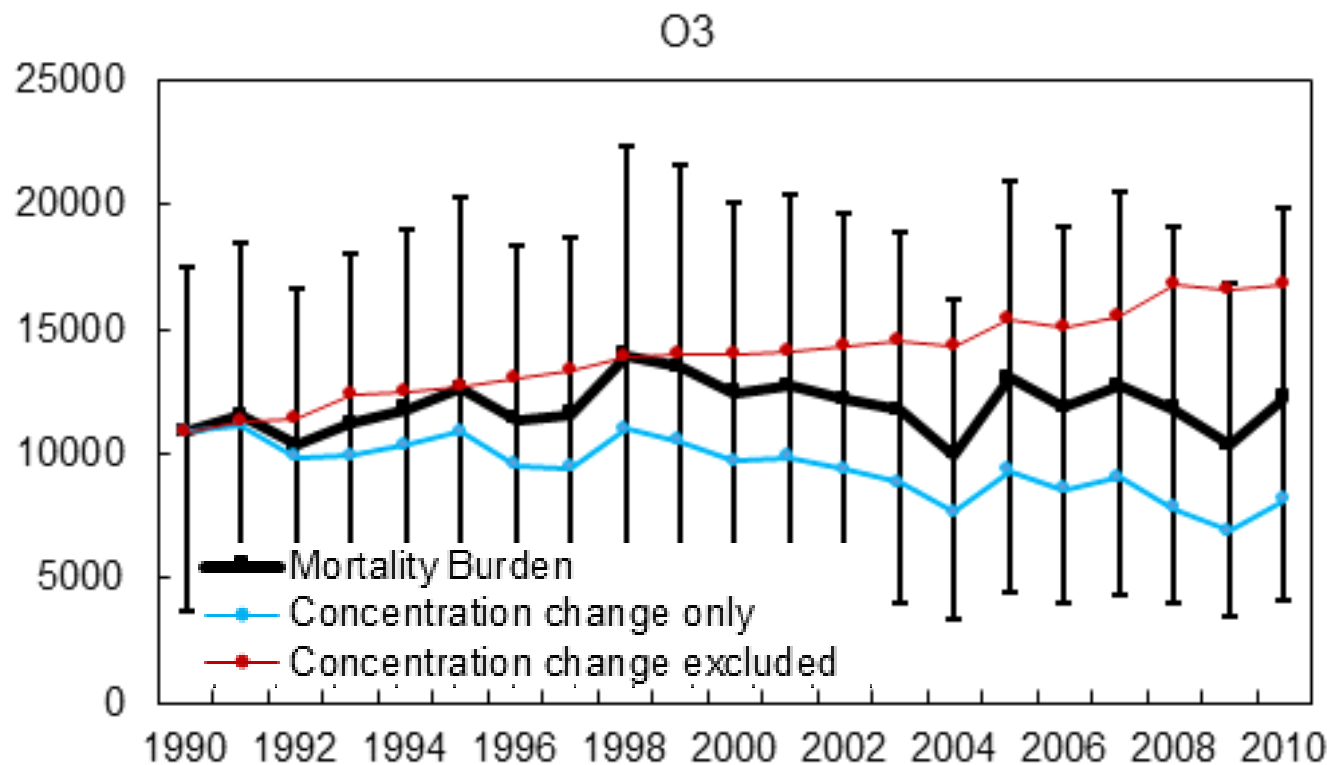


Comparison with Other Studies (O₃)



Zhang, ACP 2018; Nawaz, in prep.

O₃ Mortality Burden (EPA)



- Ozone mortality increased by 13% from 10,900 (3,700-17,500) deaths in 1990 to 12,300 (4,100-19,800) in 2010.
- Without the decrease in ozone since 1990, the burden would have increased by 55%.
- Ozone reductions since 1990 have decreased deaths in 2010 by about 4,600.

In the US, air pollution kills:

109,000 (2017 from GBD), 1 in 25 US deaths

47,000 (2015 our work), 1 in 58 US deaths

Diabetes: **80,000**

Influenza & pneumonia: **52,000**

All suicides: **45,000**

All transportation accidents: **43,000**

Breast cancer: **42,000**

All gun shootings: **39,000**

Prostate cancer: **30,000**

Parkinson's: **30,000**

Leukemia: **23,000**

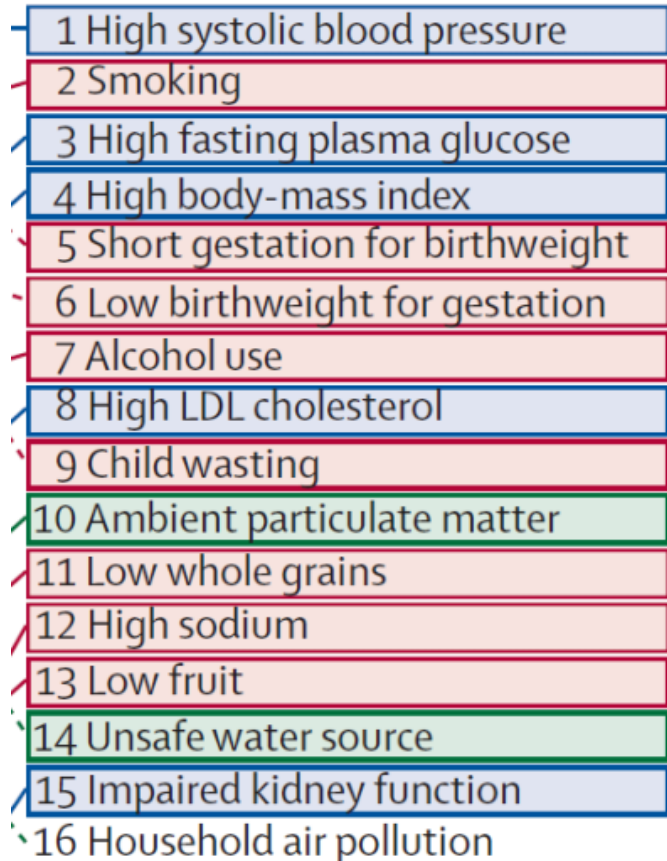
HIV AIDS: **6,000**

2016 data from CDC

Global burden of disease of air pollution (2017)

Global Deaths per Year

Ambient PM _{2.5} pollution:	2.9 (2.5 – 3.4) million	} 1 in 19 deaths globally!
Ambient ozone pollution:	0.47 (0.18 – 0.77) million	
Household air pollution from solid fuels:	1.6 (1.4 – 1.9) million	1 in 45 deaths globally!



Ambient PM_{2.5} pollution is the 8th leading risk factor for death globally.

Burnett et al. (PNAS, 2018) estimate 8.9 (7.5-10.3) million deaths from PM_{2.5} in 2015.