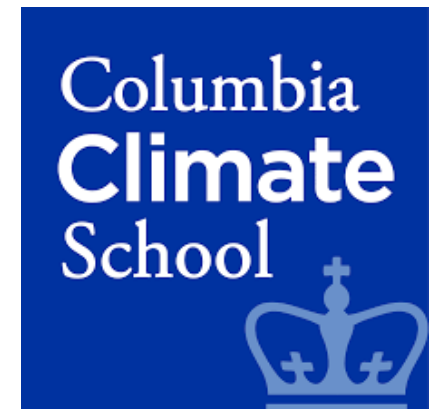


Trends in PM air pollution in the context of internal climate variability

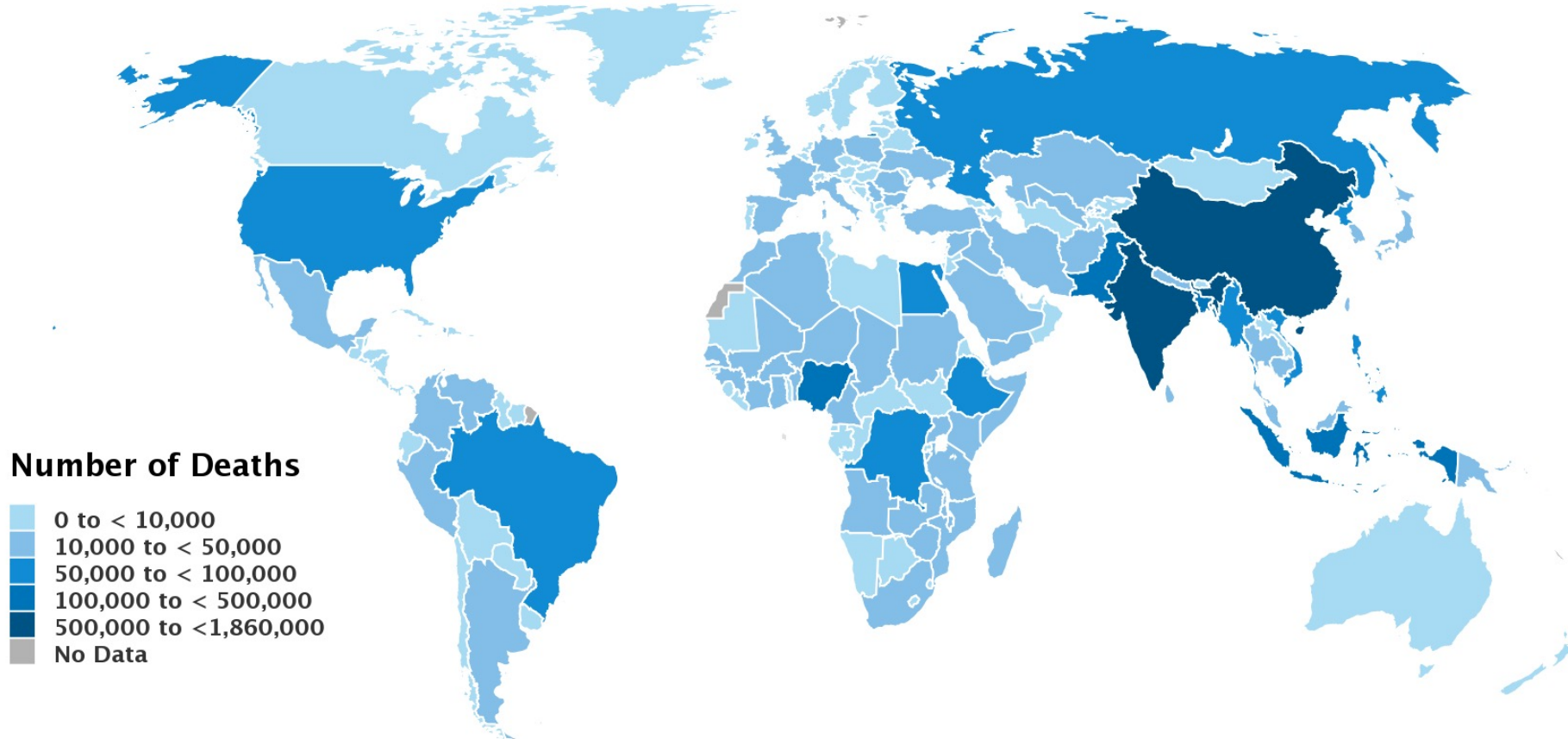
Olivia Clifton, Susanne Bauer, Kostas Tsigaridis, Larissa Nazarenko, Gregory Faluvegi

NASA GISS & Columbia University



Air pollution leading cause of premature mortality globally

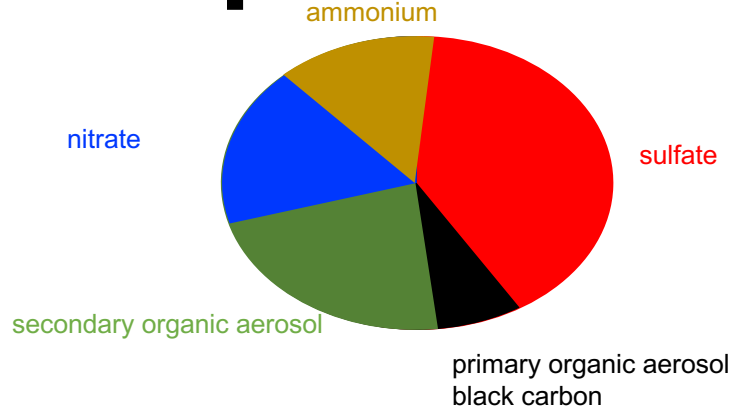
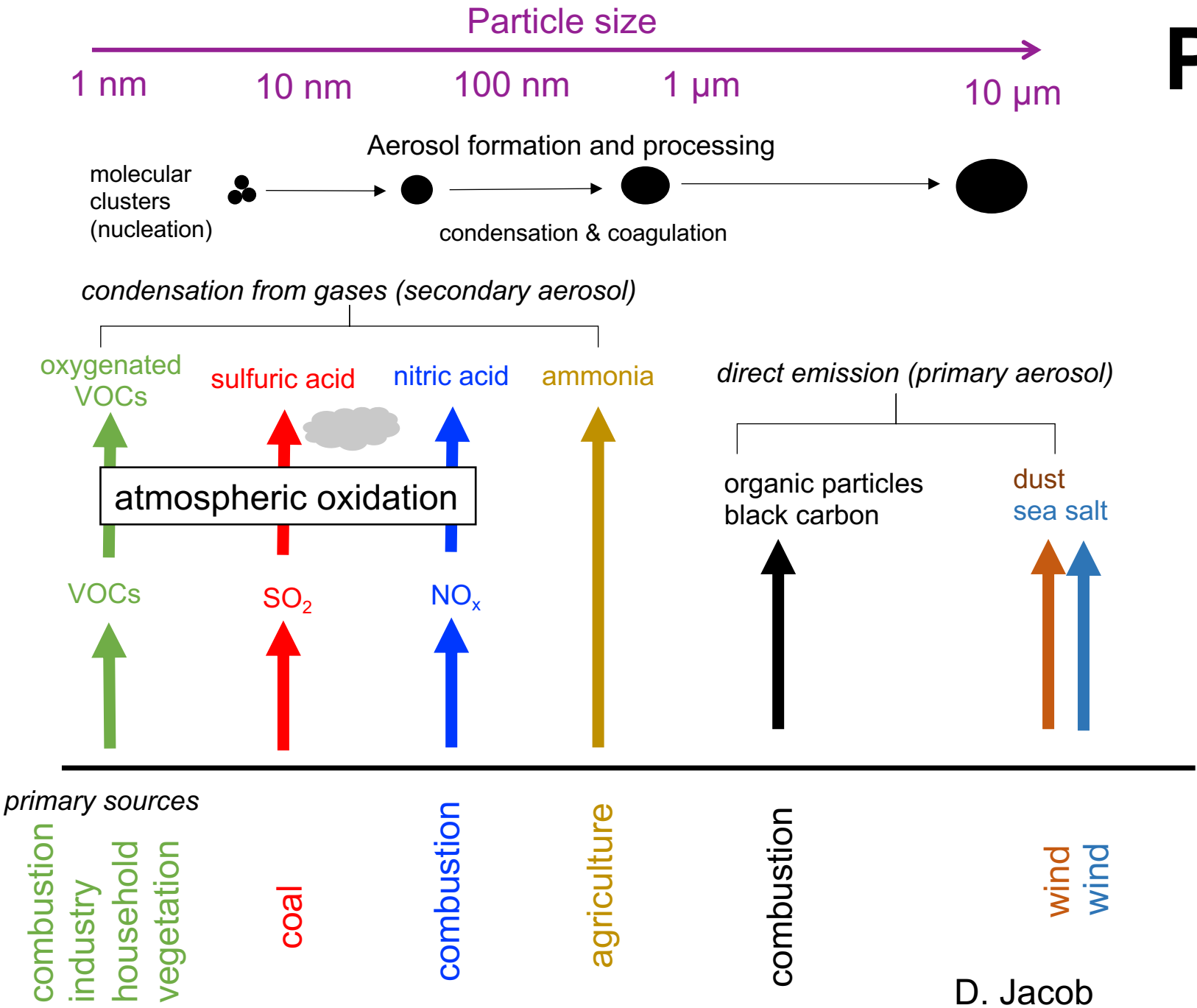
Premature deaths attributable to air pollution, 2019



- 4.14 million deaths attributable to outdoor PM_{2.5} exposure in 2019

State of Global Air, 2020

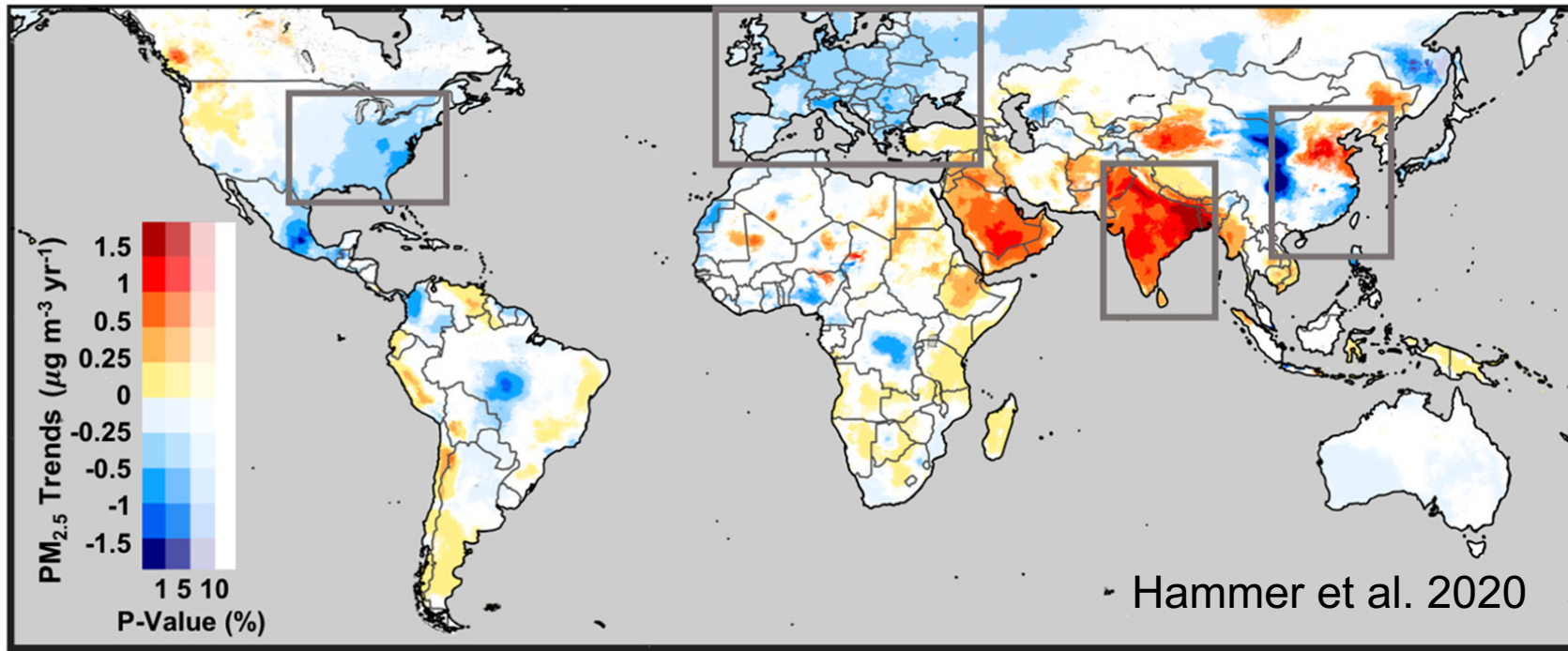
PM_{2.5} sources & composition



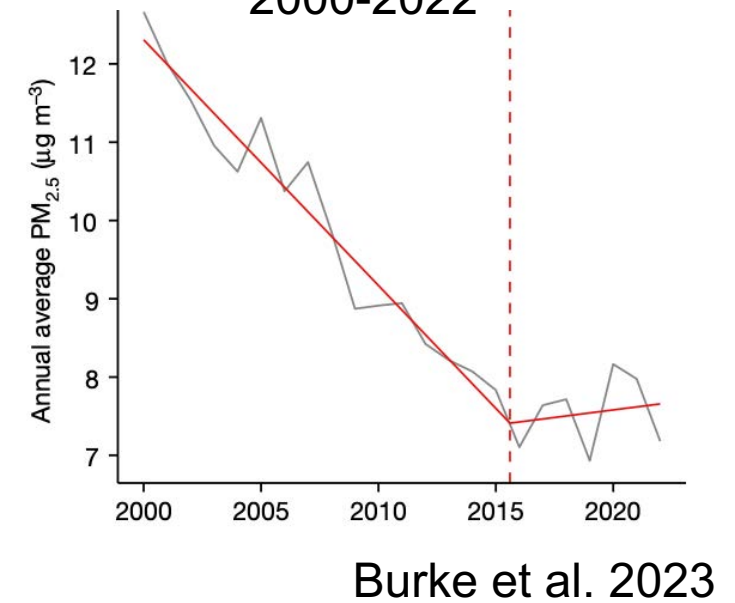
- PM_{2.5} composition varies by location, season, etc.
- Different PM types have different sources, some overlapping
- Particles transported, removed by wet and dry deposition

Large, regionally distinct changes in $PM_{2.5}$ over the past few decades, corresponding mostly to anthropogenic emissions

Trends in satellite-based $PM_{2.5}$ over 1998-2018



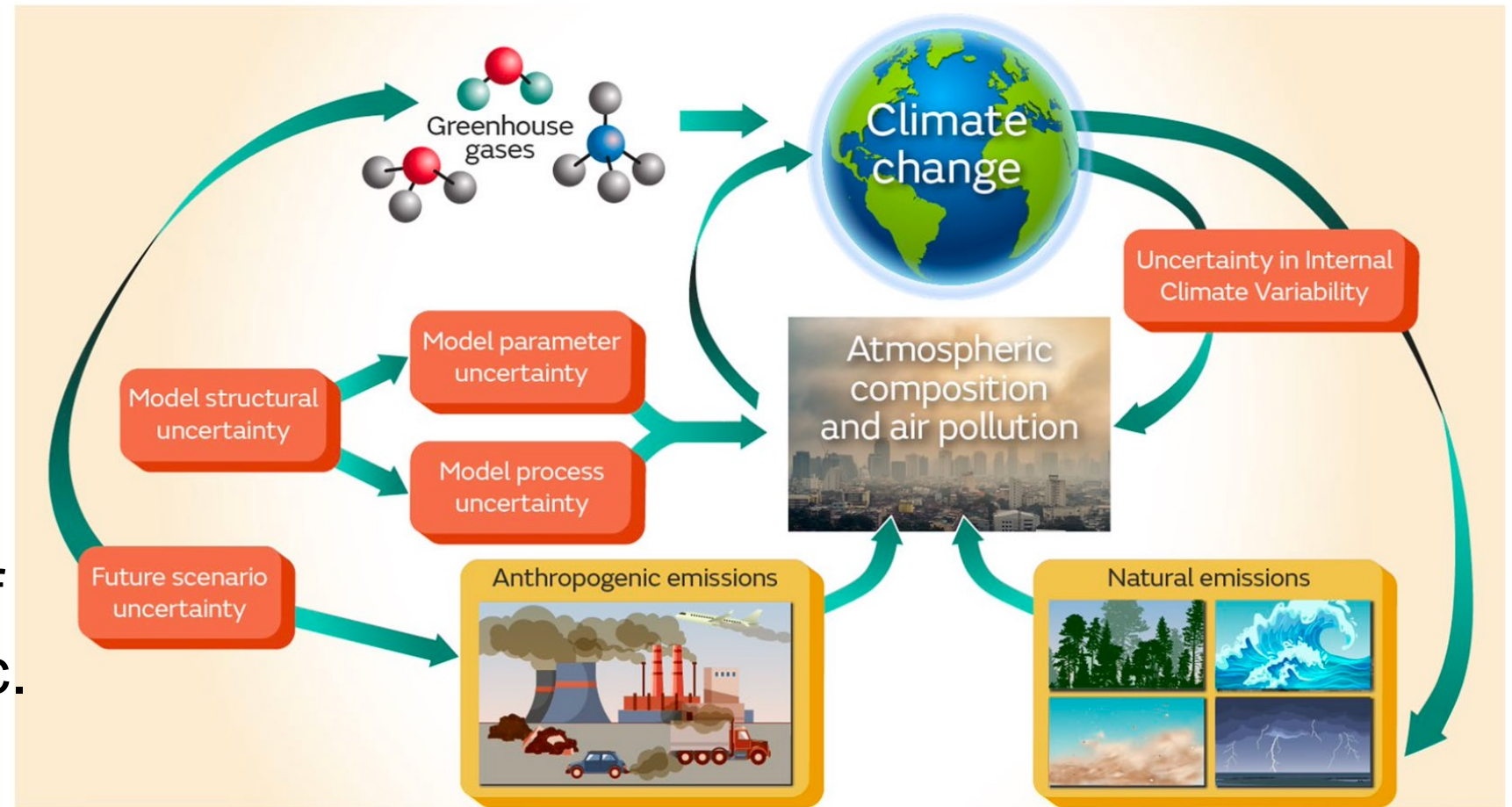
Ground-based monitoring stations across CONUS over 2000-2022



We rely on observational records of PM to estimate the efficiency of anthropogenic emission controls and advance process understanding of PM.

Internal climate variability = key, overlooked uncertainty for PM_{2.5}?

- Follows strong links with meteorology (e.g., transport, aerosol formation & processing, wet and dry deposition, natural emissions of PM & precursors inc. 🌳 & 🔥)



Overarching goal: Understand extent to which observed PM_{2.5} trends are imprinted by climate variability

CMIP6 historical simulations with interactive aerosols

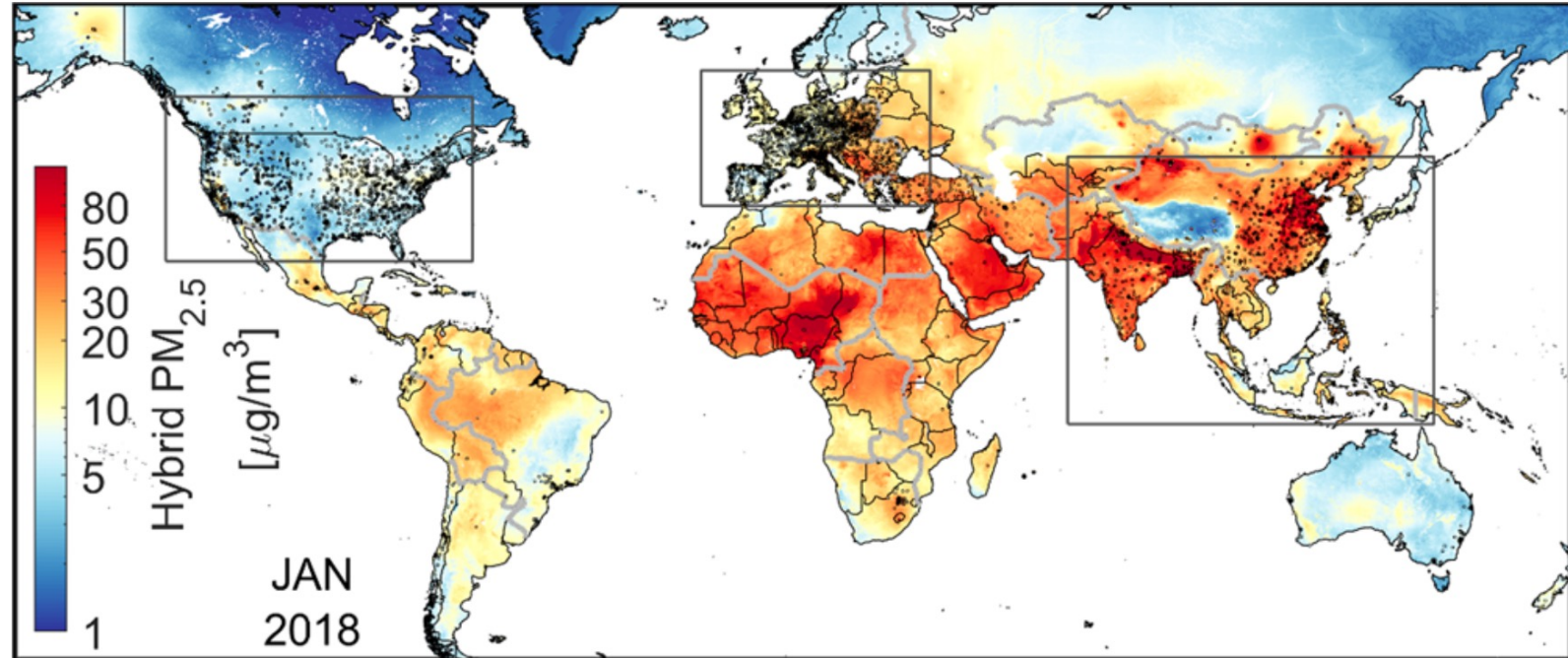
- GISS E2.1
 - Two aerosol configurations – OMA & MATRIX
 - 10 ensemble members each
- CESM2-WACCM6
 - 3 ensemble members in archive (expanded to 12 – Fiore et al., 2022)

First steps

- 1) Establish whether the set of simulations captures “observed” PM trends as well as the “observed” range of interannual variations.
- 2) Use this analysis to uncover model strengths and limitations.

PM_{2.5} observation-based products

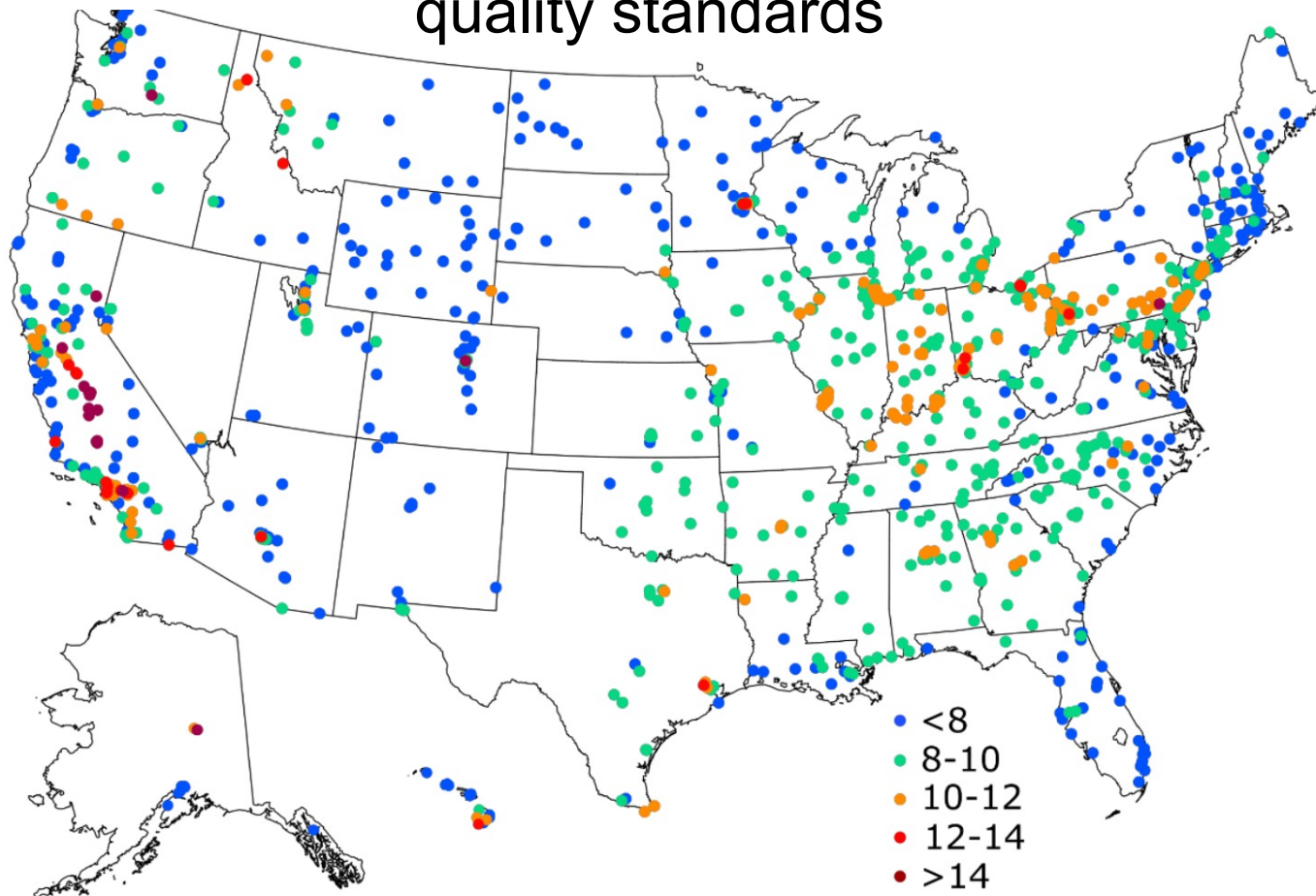
- Few places have long-term ground-based observations of PM, but most places have none (or only starting in more recent years)
- Over two decades of satellite retrievals of quantities like aerosol optical depth (AOD) that can be linked to near-surface PM



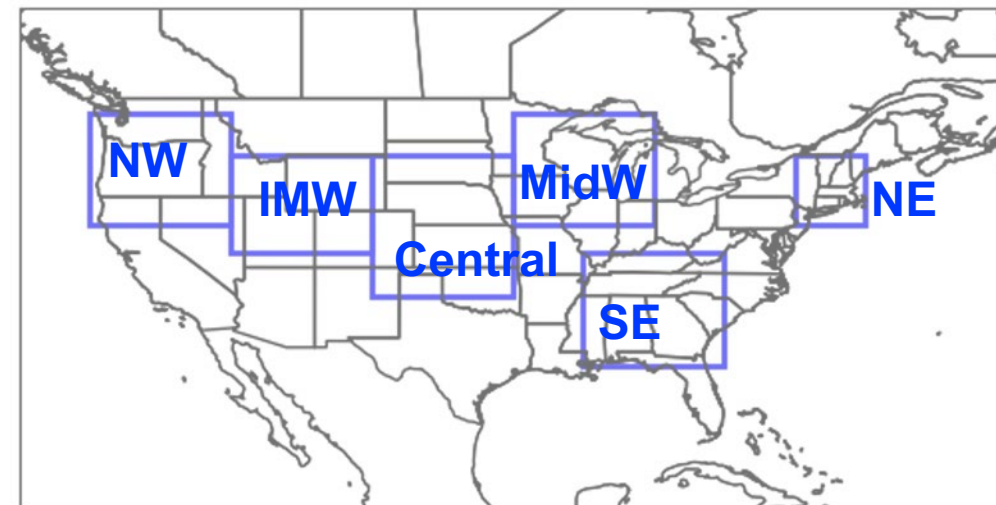
- Global monthly PM_{2.5} product (inc. uncertainty), from 1998 onwards (van Donkelaar et al. 2021)
- Combines satellite retrievals of AOD, chemical transport modeling & ground-based measurements

PM_{2.5} observation-based products

Annual PM_{2.5} concentrations ($\mu\text{g m}^{-3}$) 2013-2015 at EPA monitoring sites used for air quality standards

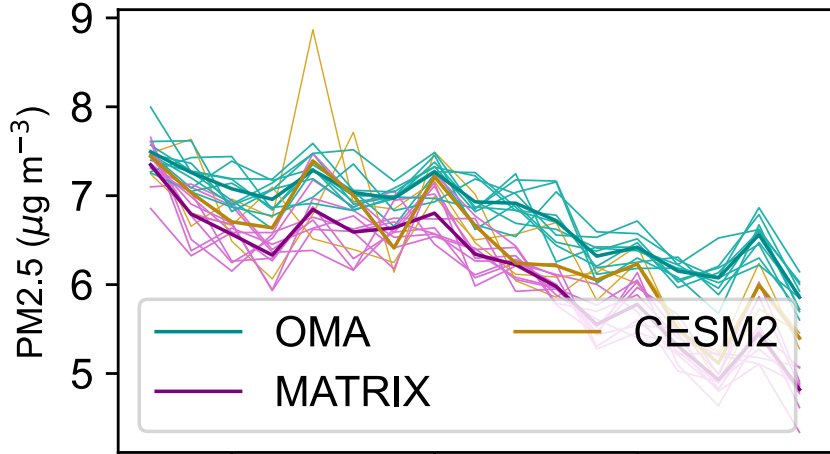


- 2° by 2.5° gridded product of monthly mean PM_{2.5} from sites
- 2003-2013
- Calculated by Klovenski et al. (2022) following methods of Schnell et al. (2015)

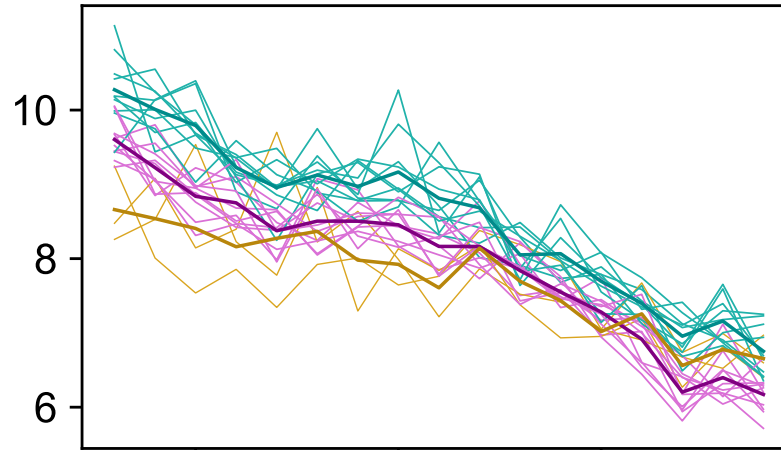


Annual PM_{2.5} over U.S. regions, from 1998 to 2014

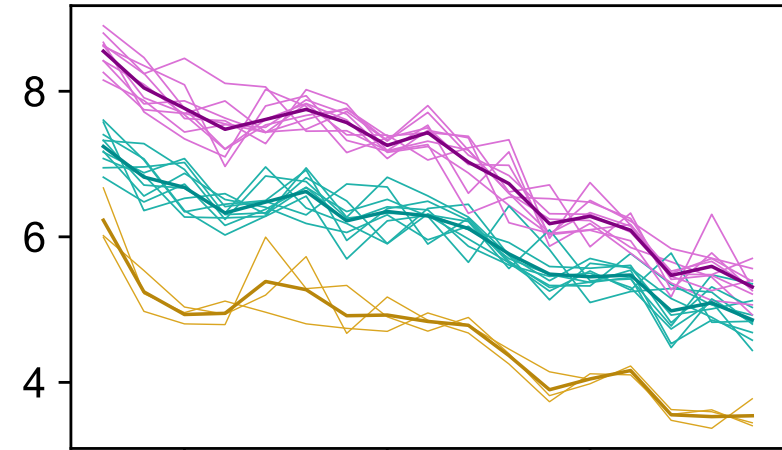
NE US



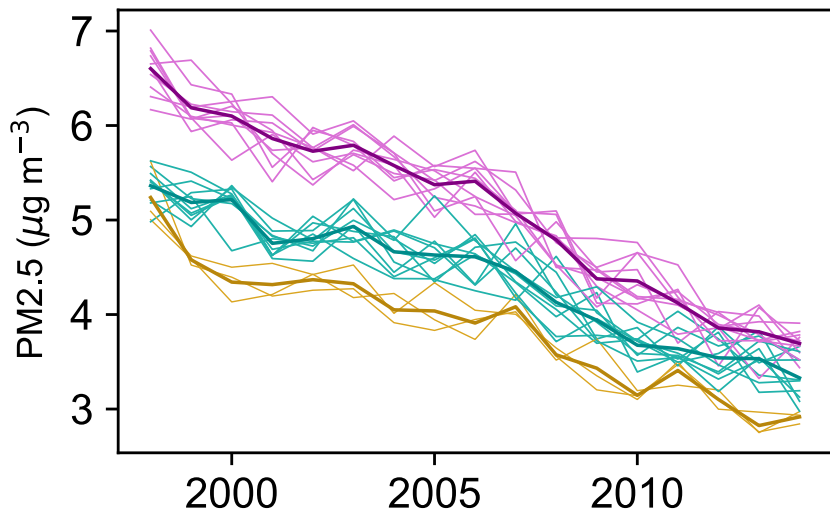
SE US



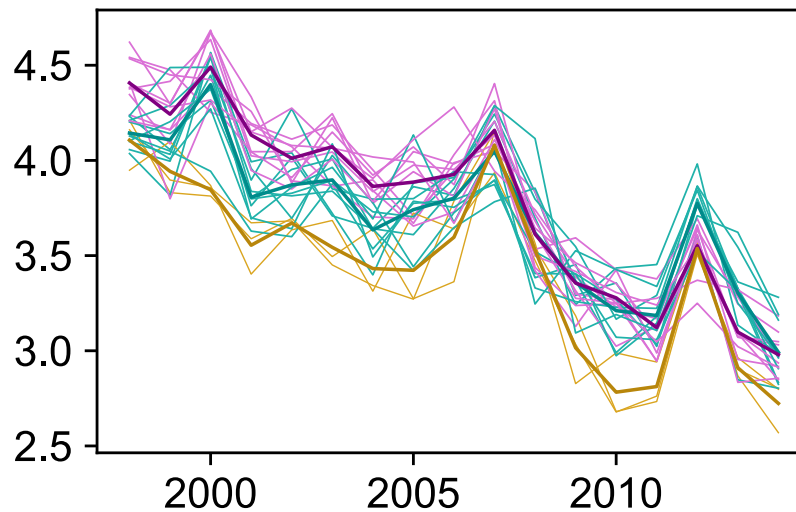
MidW US



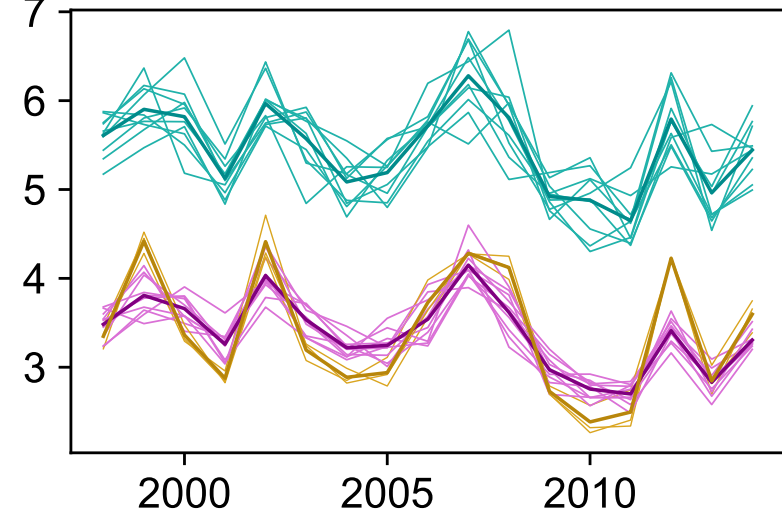
Central US



IMW US



NW US

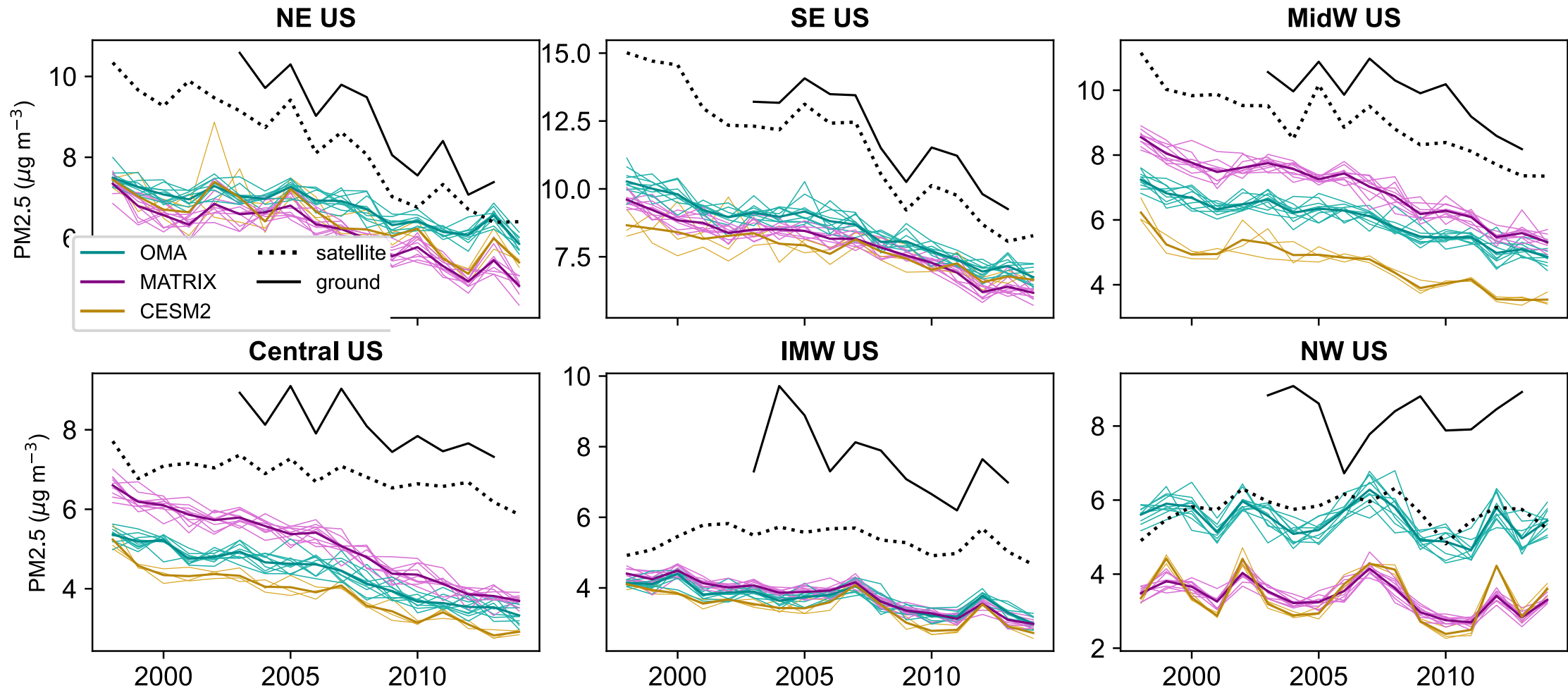


CMIP6 fully coupled historical sims

Thick – ensemble averages

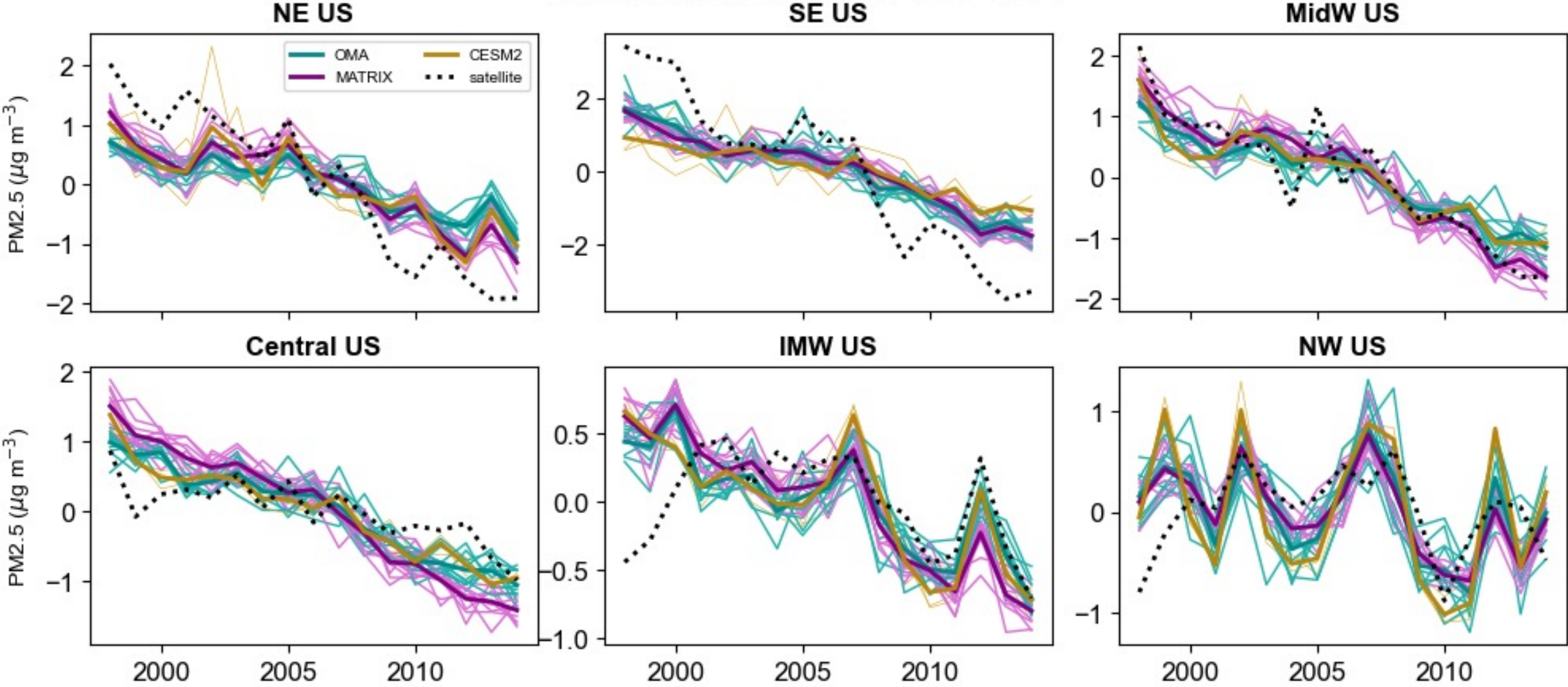
Thin – individual ensemble member

Annual PM_{2.5} over U.S. regions, from 1998 to 2014



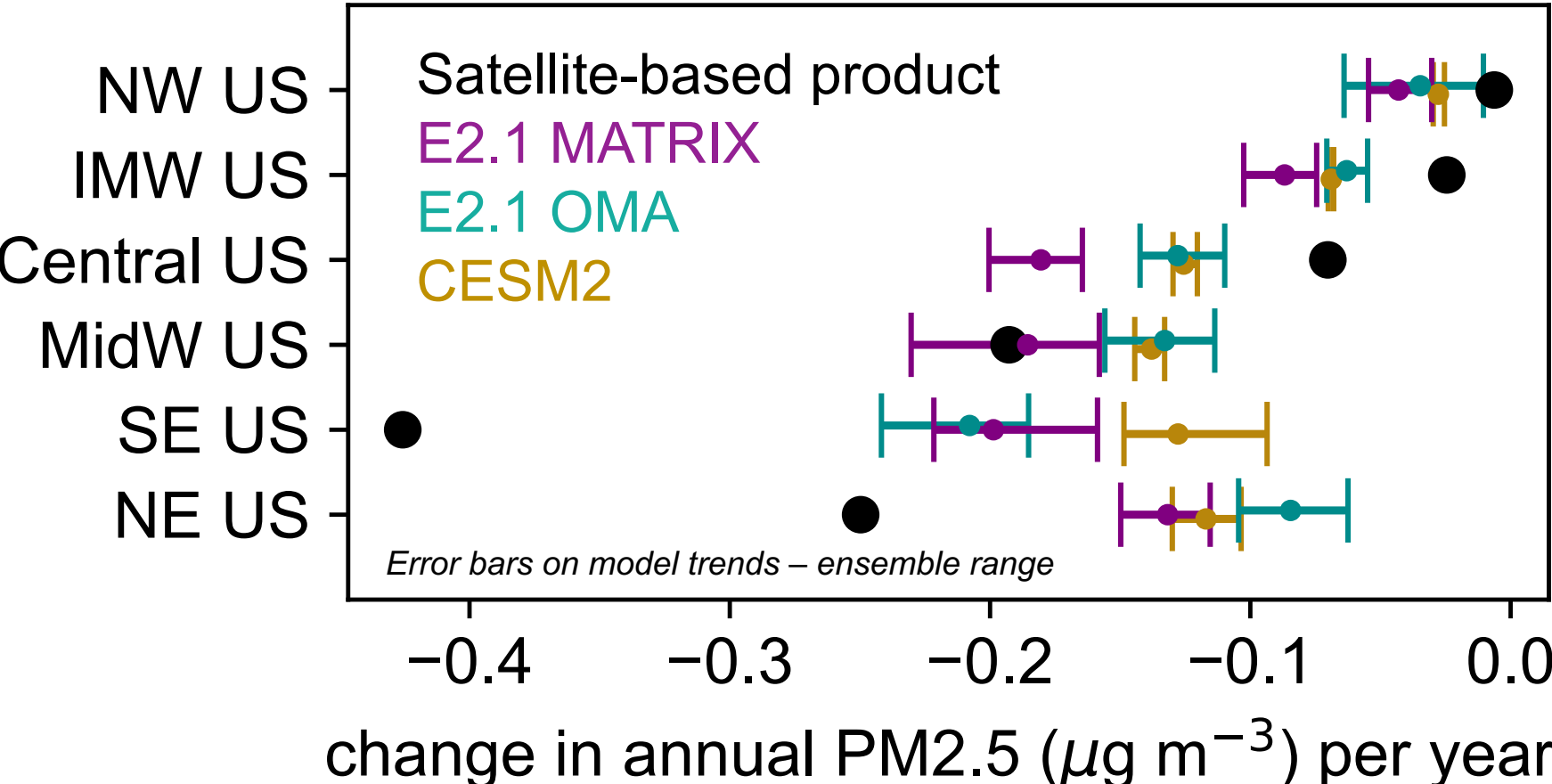
- Observation-based products similar for NE, SE, & MidW; different for Central & western US
- Low PM_{2.5} in models relative to observation-based products

Annual anomalies in PM_{2.5} over U.S. regions, from 1998 to 2014



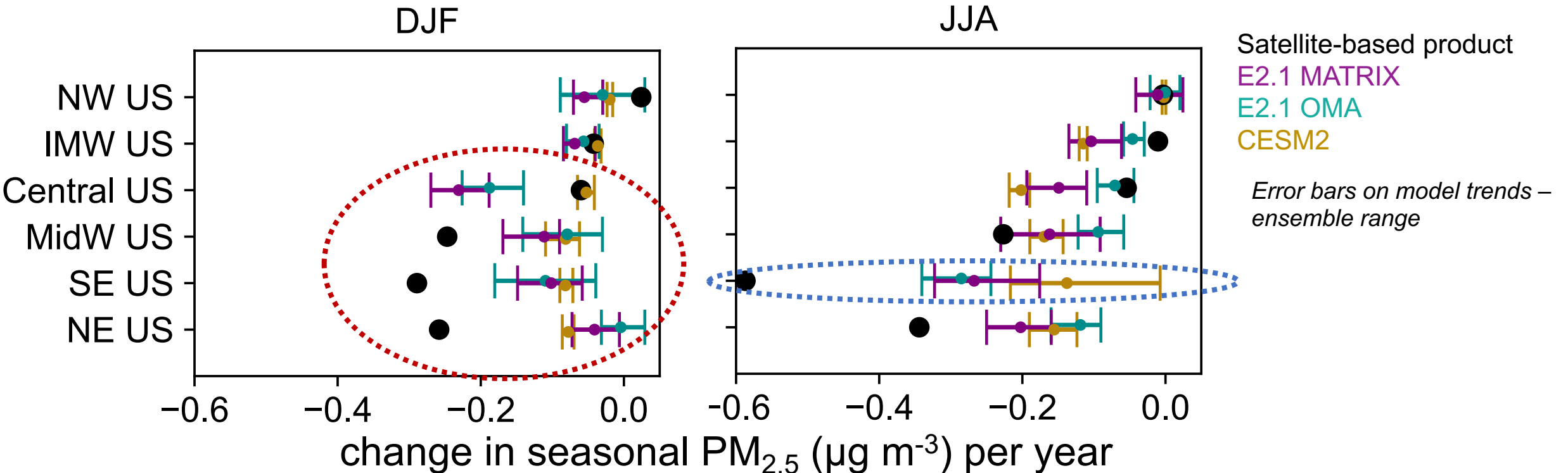
- Models largely capture degree of IAV in satellite product (but not always trend)
- Similar finding for ground-based product (not shown)

Annual trend in PM_{2.5} over U.S. regions, from 1998 to 2014



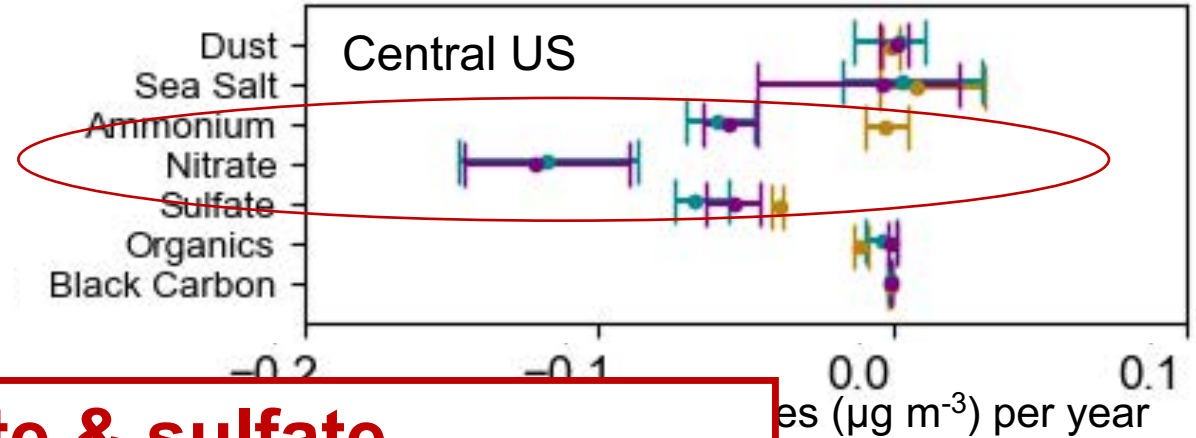
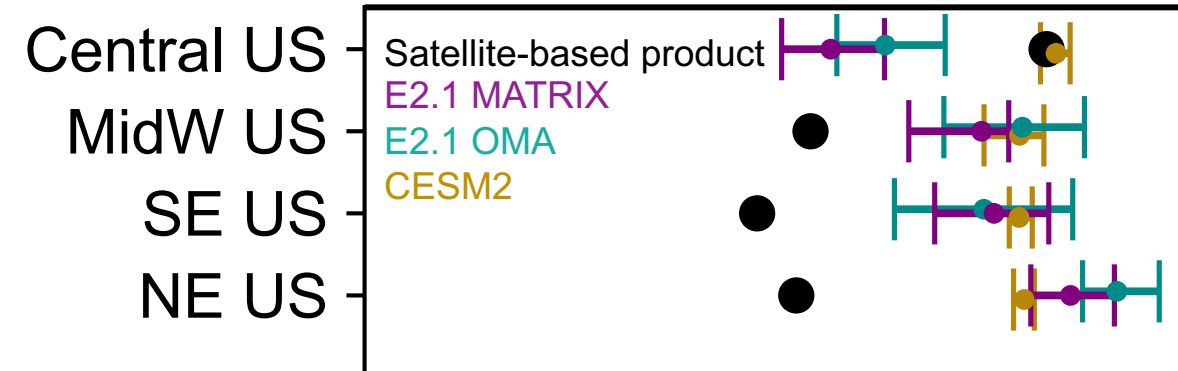
- Models slightly overestimate satellite-based trend over NW, IMW, & Central US
- MATRIX captures satellite-based trend for MidW
- Models strongly underestimate satellite-based trend over SE & NE

Seasonal trend in $\text{PM}_{2.5}$ over U.S. regions, from 1998 to 2014



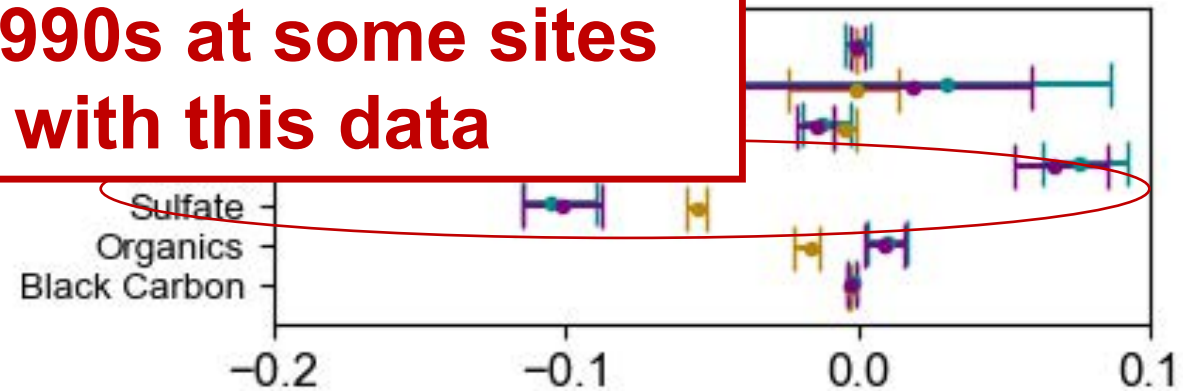
- Models do well for western US during DJF but issue for Central, MidW, SE, & NE (**CASE STUDY #1**)
- Large uncertainty during JJA over SE points to role for vegetation emissions & secondary organic aerosol (**CASE STUDY #2**)

Case Study # 1: What is going on during winter over eastern, Midwest & Central US?

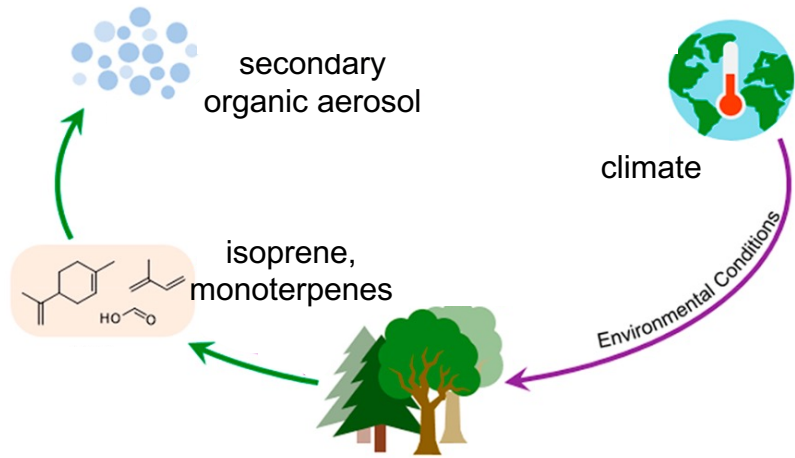


There are ground-based nitrate & sulfate measurements since the late 1990s at some sites across the US – next: evaluate with this data

- Central US
 - Nitrate & sulfate decreases in $\text{PM}_{2.5}$
 - CESM does not have nitrate & shows no change in ammonium so weak decreases



- MidW, SE, & NE US
 - E2.1 increases in nitrate *tempor* decreases in $\text{PM}_{2.5}$ driven by sulfate
 - CESM does not have nitrate & decreases in sulfate weaker than E2.1



Case Study # 2: What is going on during summer over SE US?

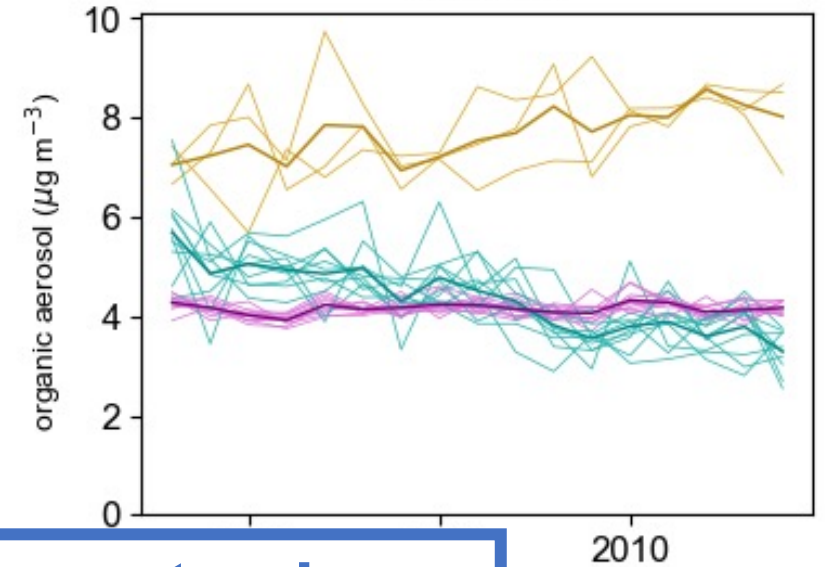
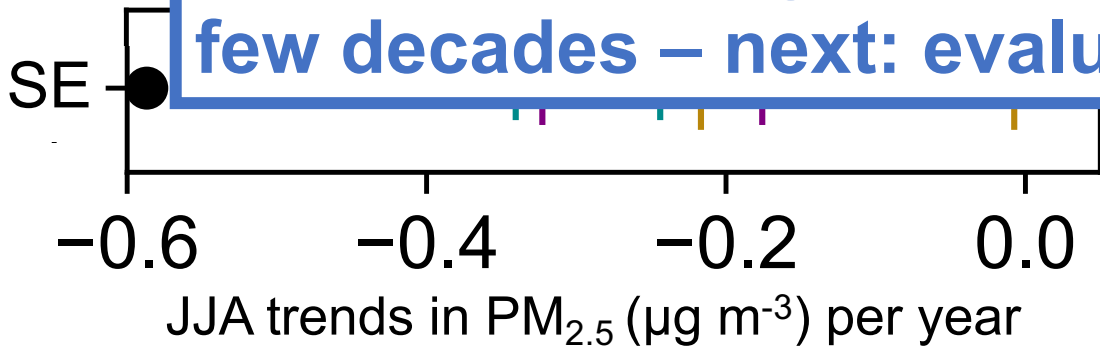


figure adapted from [unclear] 2014

Ground-based organic aerosol measurements show decreases during summer over the SE US over past few decades – next: evaluate with this data



Satellite-based product

CEM2 E2.1 MATRIX E2.1 OMA

weakenes the decreasing trend in $\text{PM}_{2.5}$ in CEM2 and increases ensemble spread

- Lack of variability in organic aerosol in MATRIX due to strong tie to static monoterpene emissions

Takeaways

Ugly – PM_{2.5} observational products are imperfect and uncertain as well as rather short-term

Bad – Representation of nitrate, sulfate, & secondary organic aerosol needs to be scrutinized

Bad/Good – Models too low in terms of magnitude of PM_{2.5} but generally capture interannual variability and trends

Good – We can still learn from model-“observation” comparisons, esp. with a focus on seasons & species as well as inclusion of model structural uncertainty, even if observational products are imperfect

Good – We may be able to use climate models to understand extent to which observed PM_{2.5} trends are imprinted by climate variability