AERO-MAP: A data compilation and modelling approach to understand the fine and coarse mode aerosol composition

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10.

20.

50.

100.



~14,000 stations >20 million obs includes data outside of US and Europe

https://essd.copernicus.org/preprints/essd-2024-1/

Grided to 2x2 so we can see

Aerosols important for understanding climate



- Aerosols large source of uncertainty in current radiative forcing and PI/PD changes
- Different aerosol sources have different trends going into the future
- Different aerosols have different impacts (warming versus cooling, cloud interations, biogeochemistry)



IPCC, 2014

Methodology

- Compile total PM, SO4, EC/BC, OM/OC, Na, Al, NO3 and NH4 for PM2.5 and PM10 at the daily to weekly level globally (much more data available then in previous datasets (Szopa et al., 2021 2010)
- This is the first paper presenting annual means across 1986-202 time period. More papers will follow with temporal resolution
- Compare to CESM model (default)+ added NO3/NH4 from Vira et al., 2021
- Add in other sources:
 - Agricultural dust (tuned to Ginoux et al., 2012 satellite estimate)
 - Road dust, industrial emissions, coarse OC and BC (from Klimont et al. 2017)
 - Add in primary biogenic emissions (Mahowald et al., 2005; Burrows et al., 2009; Heald and Spracken 2009)
- Average to 2x2 grid for display (where there is data there is often too much to display)
- Convert modeled composition to observed composition when needed (e.g. dust to Al or sea salts to Na).







a. PM2.5 (ug/m³)



Gridded comparison

100

100

c. PM2.5 (ug/m³)

N=742 r= 0.72

100

Model too high over Asia?



Included a lot of new data from China and India: there seems to be a systematic bias between Chinese and and Indian air quality data versus US embassy data: unclear why.



Compare by constituent as well.

10.0

1.0 Model



0.005

0.001

0.01

0.02

0.05

0.10

0.5

Model SO4 too high BC/EC about right

Much less data than PM

Liu et al., 2011, 2016

PM2.5 constituents

e. OM PM2.5 ug/m³



- OM/OC: a little high: much less • data
- Na: model too flat compared to ٠ obs. (need industrial source of Na?

Al/dust



PM2.5 Nitrogen aerosols

k. NO₃ PM2.5 ug/m³



Needed to half NO3 (Vira et al., 2021) (likely because no thermodynamic model)

m. NH₄ PM2.5 ug/m³







Or new data not representative?

~ v =

c. PM10 ug/m³ (gridded) N= 687 r= 0.67

PM10 sulfate and BC/EC



Model over estimates sulfate in coarse mode

Added BC into coarse mode: maybe too much?

PM10 OC/OM and seasalts/Na



Dust/Al



PM10 NO3 and NH4



Model doesn't differentiate coarse versus fine, so this is all nitrogen

It's very hard to get N aerosols right without thermodynamic model which was not in Vira et al., 2021.

Most areas of the globe or even of land have no data

a. PM2.5 coverage (%)



b. PM10 coverage (%)



Red: only PM, blue/purple also composition

3% of land is covered by observations of aerosol.

Surface aerosol amount and important composition is not well measured.

Cannot get composition from remote sensing.

N aerosols are going up, sulfate down: don't know where they are (Adams et al., 2001; Bauer et al., 2007)

Need more in situ data.

CESM does not include many important aerosols

- PM2.5 concentration default sources (%) b. PM2.5 concentration new sources (%) c. PM10 concentration default sources (%) d. PM10 concentration new sources (%)
 - 0.0 10. 20. 30. 40. 50. 60. 70. 80. 90. 100.

- Most important: N aerosols: need for climate simulations
 - 30% of aerosols
 - Have different trends (upward due to land use) than sulfate (downward due to less fosssli fuels)
- Also: agricultural dust important to include
- (other additions didn't matter as much: road dust, coarse mode EC/BC, PBPs...)

Summary conclusions

- New annual compilation for use in comparing model aerosols to data
 - Much more data in non-US and non-European areas in this compilation
- Temporal variability will be next! (other ideas: glad to collaborate)
- Includes composition data as available
- Need more in situ data to constrain current distribution of aerosols
- Models need to include N aerosols and agricultural dust or they are missing important aerosol trends.
- (Aerosols even more unconstrained if we look farther back then 1980: see Mahowald et al., 2024, ACP).