

# Evaluating aerosols, clouds, and their interactions in three global climate models using satellite simulators and observations



**George Ban-Weiss**

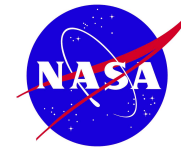
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October 1, 2014

# Acknowledgements

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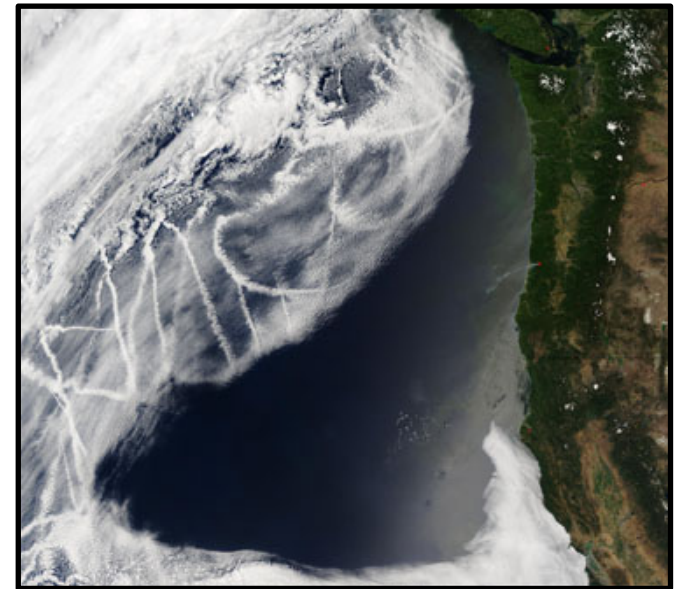
- Susanne Bauer (NASA GISS)
- Ling Jin (LBNL)
- Yi Ming (GFDL)
- Xiaohong Liu (Univ of Wyoming)
- Kai Zhang (PNNL)
- Ralf Bennartz (Vanderbilt Univ)
- Jonathan Jiang (JPL)
- Sponsor: Dept of Energy, Climate and Environmental Sciences Division, FASTER (FAst-physics System TEstbed and Research)



# Aerosol impacts on climate

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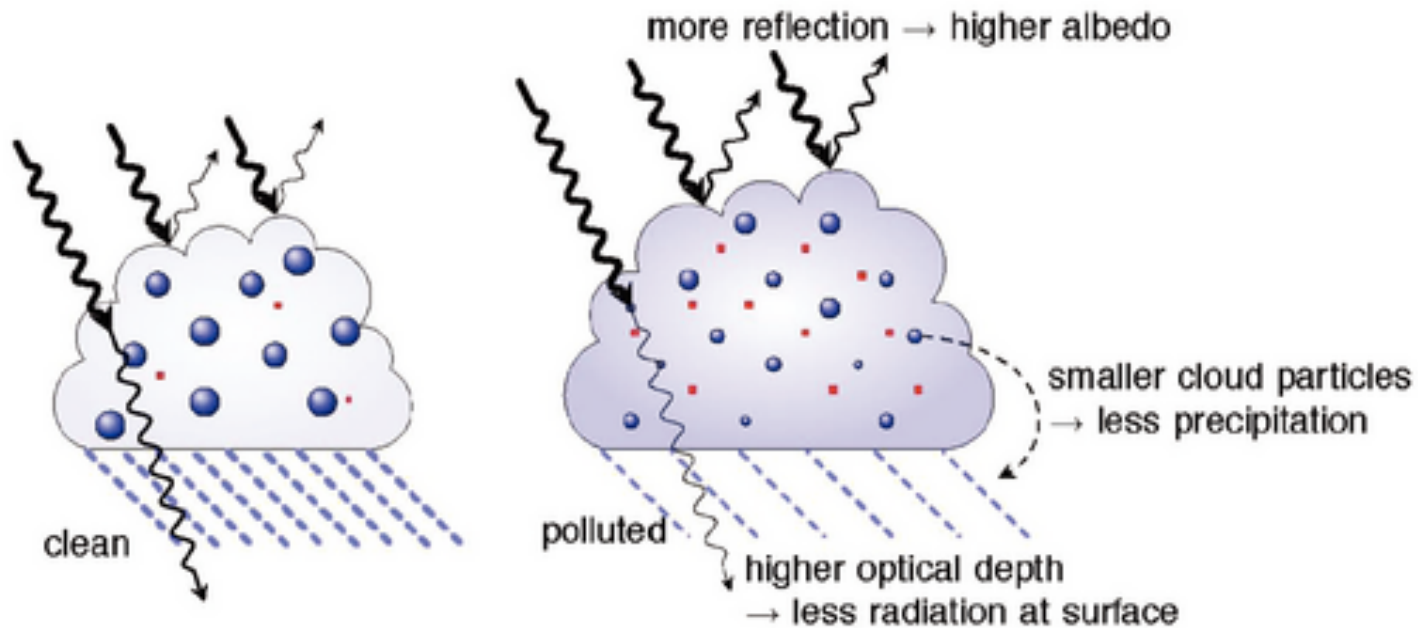
- Aerosols have profound impacts on climate
  - Scatter and absorb radiation
  - Alter macro- and micro-physical cloud properties
- Aerosol “indirect effects” on microphysical cloud properties are uncertain and difficult to quantify at global scale



[gfdl.noaa.gov](http://gfdl.noaa.gov)

# Aerosol “indirect effects” on warm clouds

Challenging to model these microphysical processes in a global climate model (with large grid cells)



source: IPCC AR4, WG1

# Research goals

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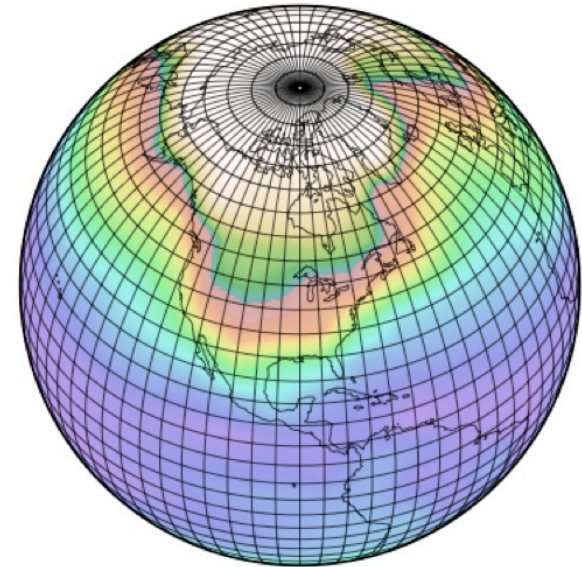
- (1) Assess global distributions of aerosol and cloud properties in three global climate models and satellite observations
  - Assure model vs observation comparison is as “apples to apples” as possible
  
- (2) Compare signatures of aerosol-cloud interactions
  - Focusing on regional analysis with specific cloud regimes (subtropical stratocumulus)
  - Focusing on effect of aerosol on cloud droplet number conc ( $N$ )
  - Accounting for meteorological influences

# Climate models and observations

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## Climate models

- CAM5 ( $1.9^\circ \times 2.5^\circ$ )
- GISS ModelE2 ( $2^\circ \times 2.5^\circ$ )
- GFDL AM3 ( $2^\circ \times 2.5^\circ$ )



*image source: serc.carleton.edu*

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## Observations

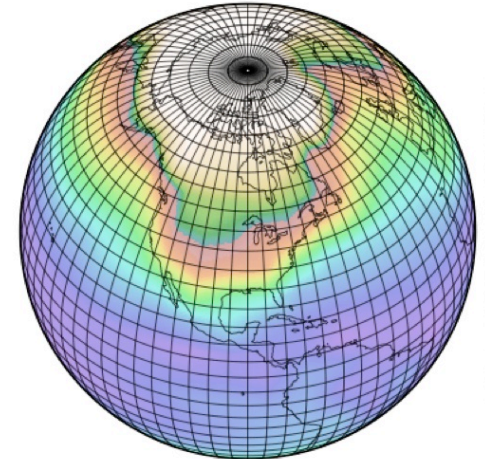
- MODIS L3C5 ( $1^\circ \times 1^\circ$ )
- ERA-INT reanalysis ( $0.5^\circ \times 0.5^\circ$ )



# GCM simulation details

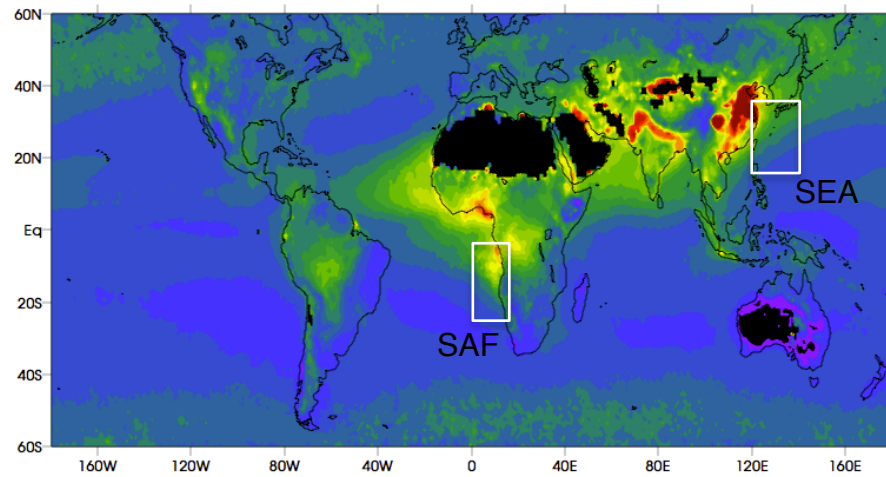
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- Simulation period: Jan 1 2007- Jan 1 2009
- CMIP5 emissions
- To maximize comparability of GCM and observations:
  - Prescribed sea surface temperatures
  - Horizontal winds are nudged to reanalysis
  - High frequency (3 hourly) GCM output
  - Extract satellite overpass times
  - CFMIP Observation Simulator Package (COSP)
  - Satellite-observed and simulated cloud drop number concentration ( $N$ ) use same algorithm

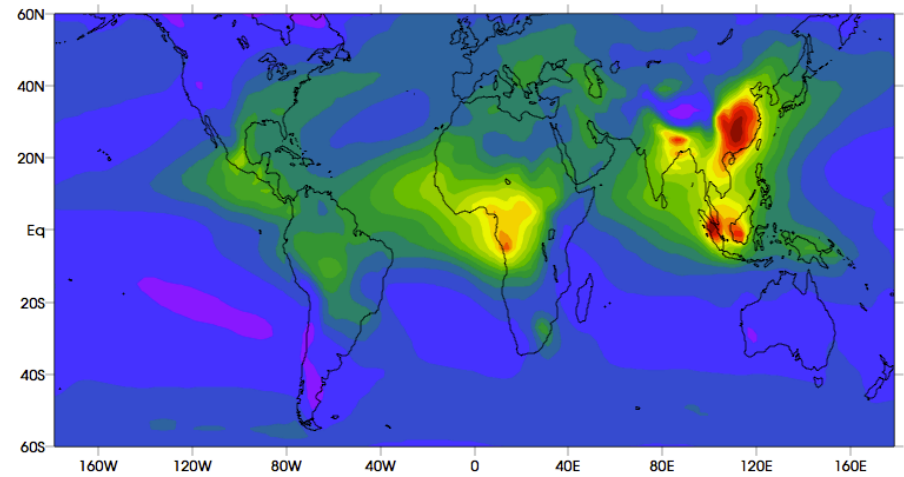


# Aerosol optical depth – 2yr annual mean

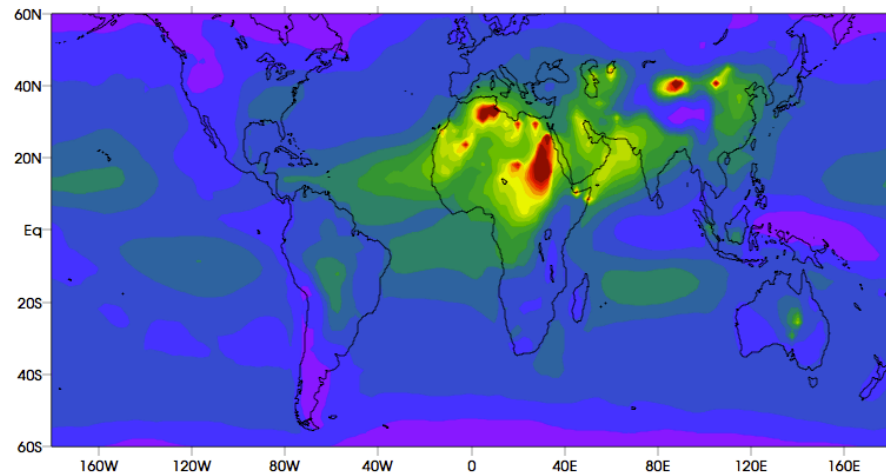
MODIS (observation)



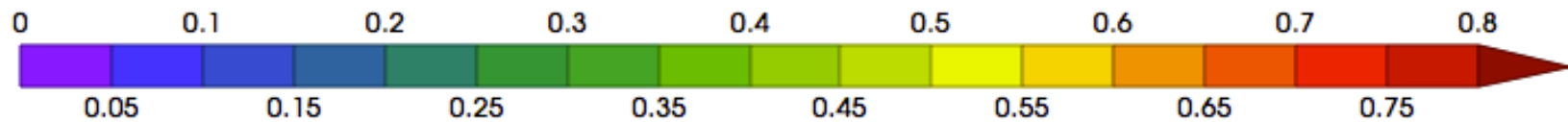
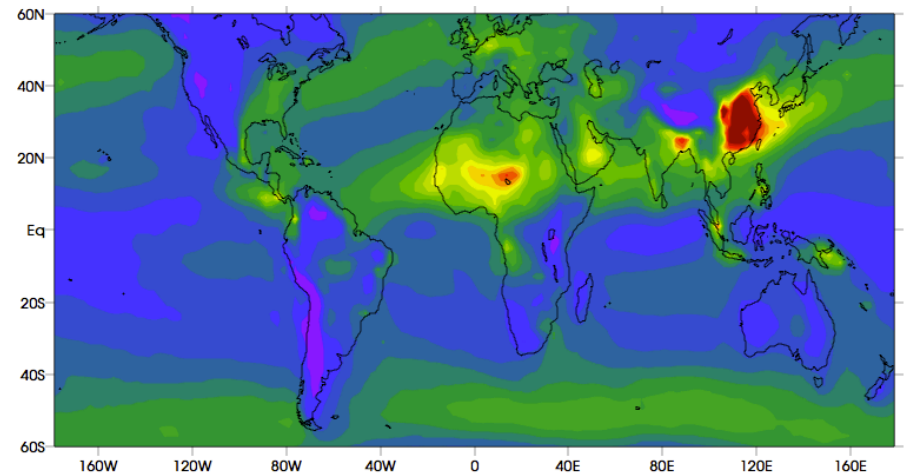
AM3 (model)



CAM5 (model)



ModelE2 (model)

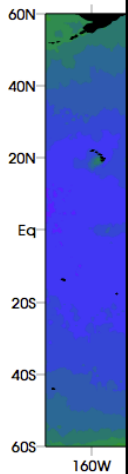




# Cloud droplet number concentration (# cm<sup>-3</sup>)

MODIS (observation)

AM3 (MODIS simulator)



- Observed  $N$ :

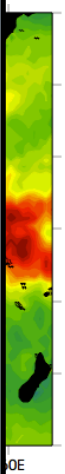
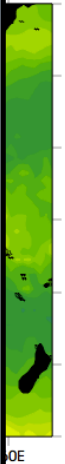
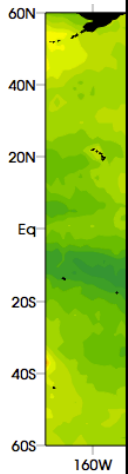
- Calculated from MODIS liquid  $\tau_c$  and  $r_e$  (Bennartz 2007)
- Assumes clouds are adiabatically stratified (LWC increases linearly and  $N$  is constant with height)
- $N \sim \tau_c^{1/2} r_e^{-5/2}$

- Modeled  $N$ :

- Computed from MODIS-simulated values using the same algorithm as MODIS observations

- Comparison:

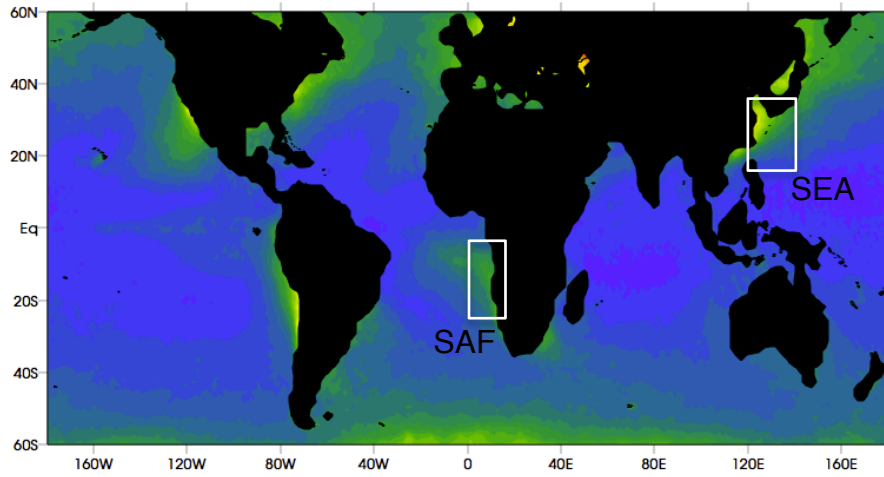
- Liquid clouds only
- Pixels with liquid cloud fraction  $< 0.3$  are screened
- **In-cloud  $N$**



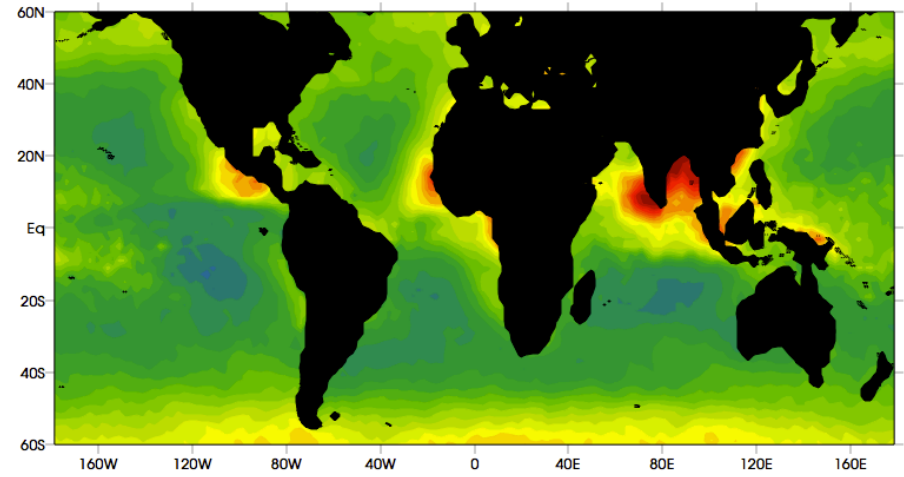
10 30 50 70 90 110 130 150 170 190 210 230 250

# Cloud droplet number concentration (# cm<sup>-3</sup>)

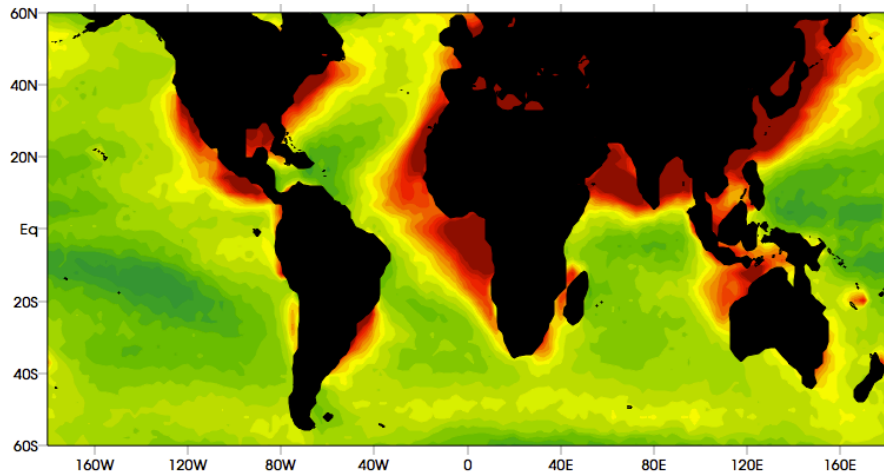
MODIS (observation)



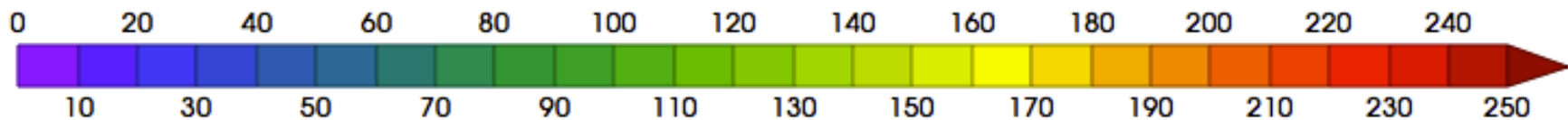
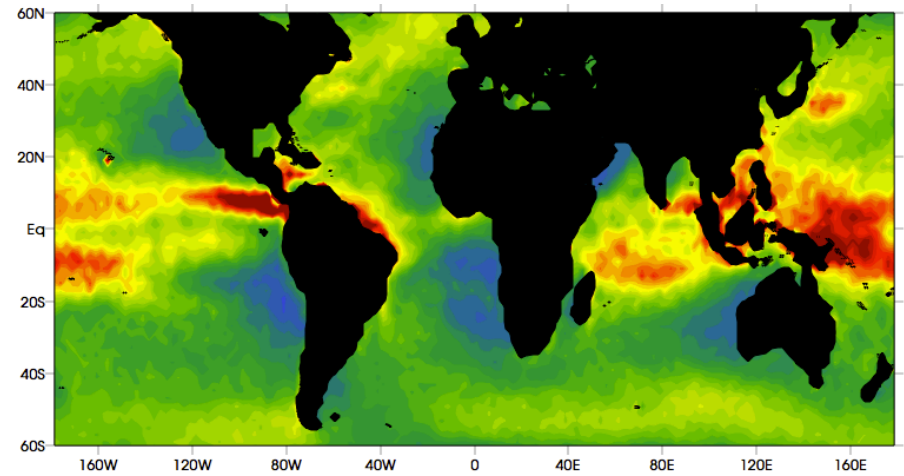
AM3 (MODIS simulator)



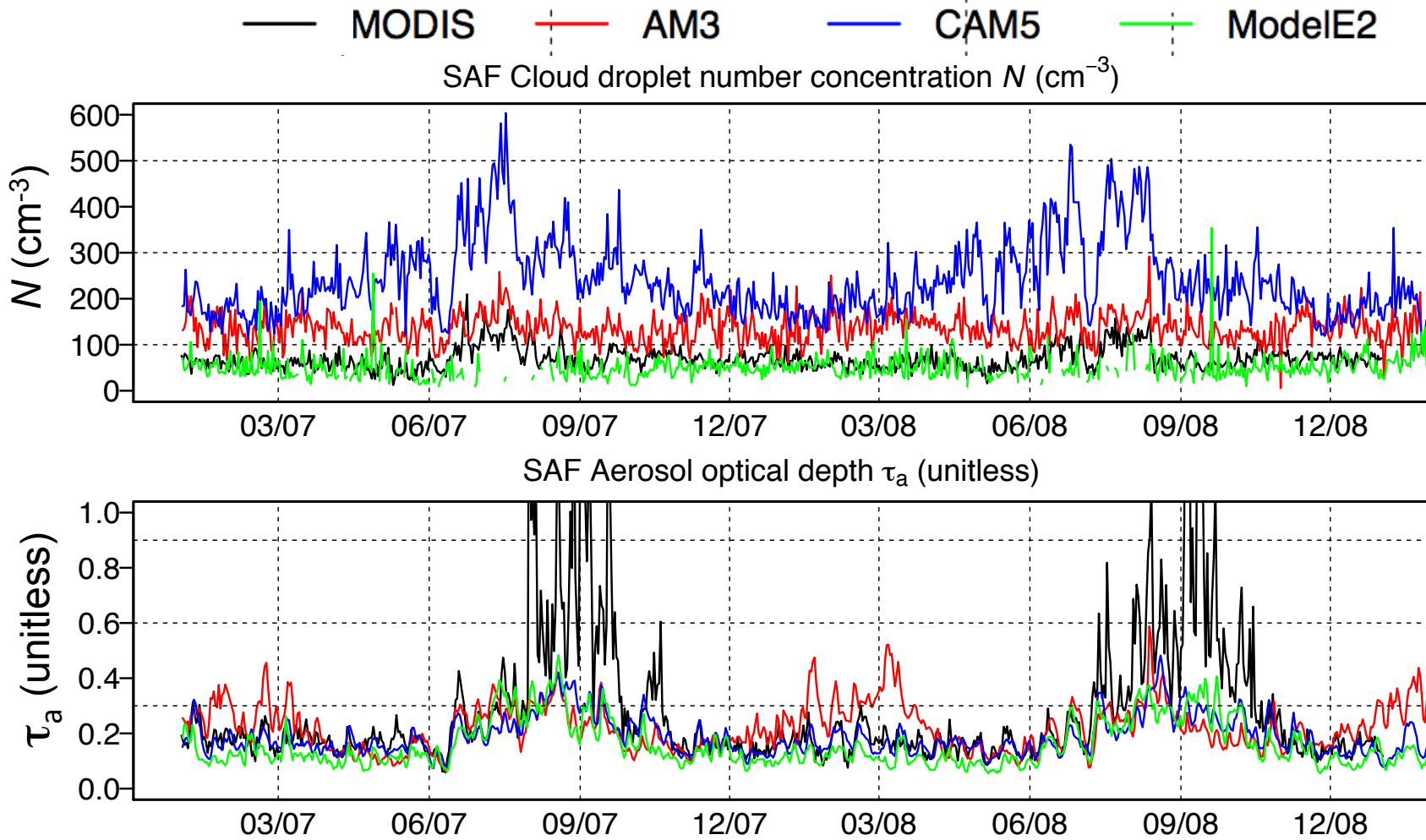
CAM5 (MODIS simulator)



ModelE2 (MODIS simulator)



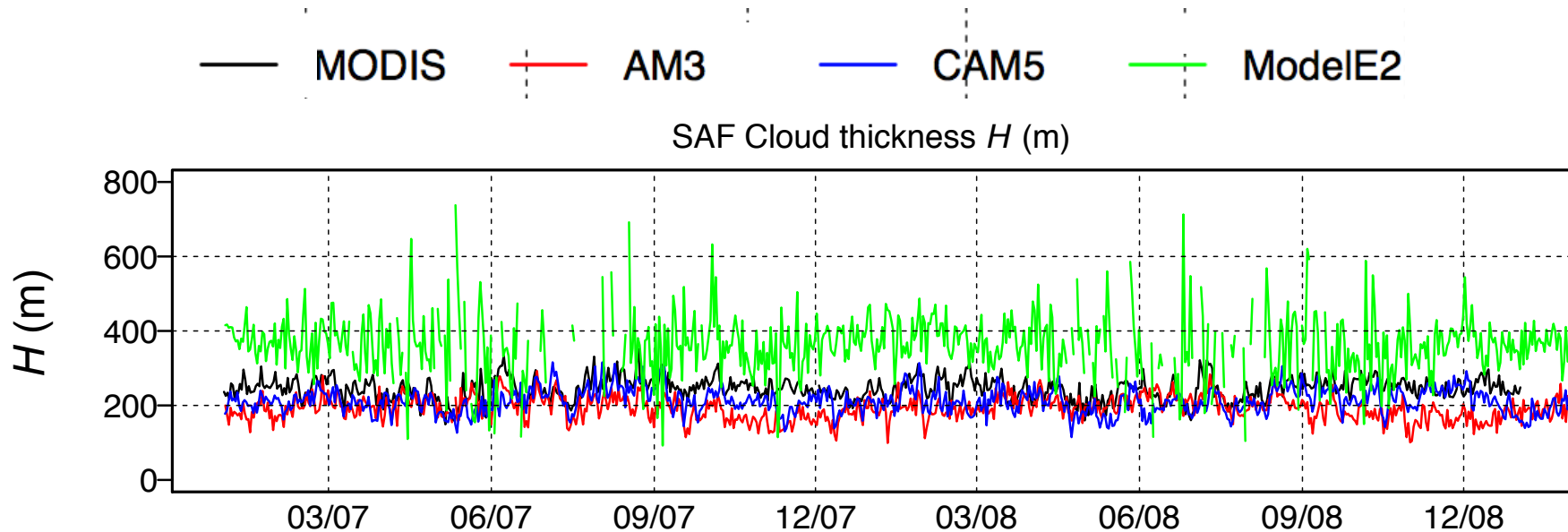
# Spatial-mean dailies (SAF)



Ban-Weiss et al. (2014)

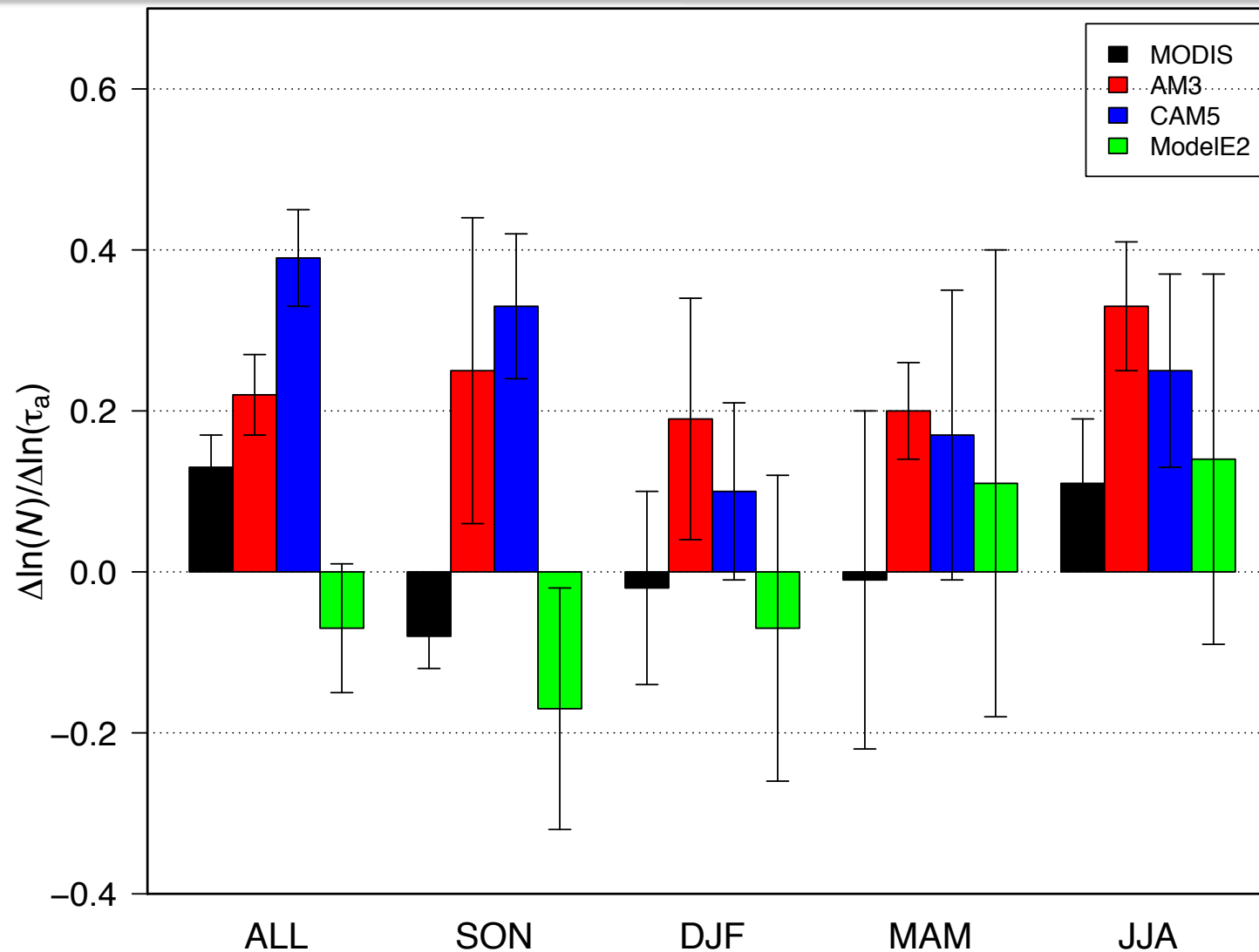
# Spatial-mean dailies (SAF)

Quick aside: Interesting method for evaluating cloud thickness



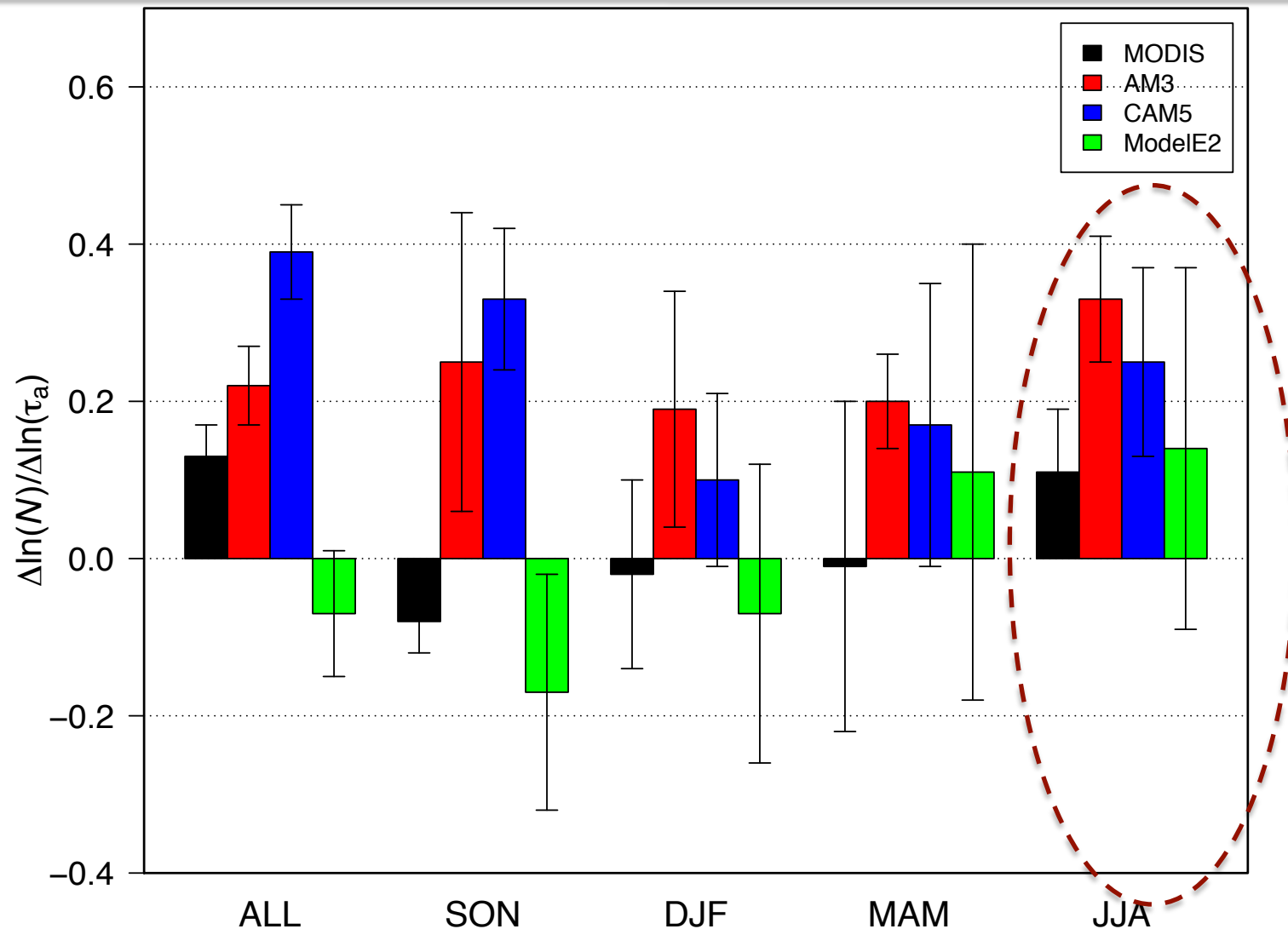
Ban-Weiss et al. (2014)

# Sensitivity of $N$ to $\tau_a$ in SAF region



Ban-Weiss  
et al. (2014)

# Sensitivity of $N$ to $\tau_a$ in SAF region



Ban-Weiss  
et al. (2014)

But  $N$  is also sensitive to meteorology. And meteorology and  $\tau_a$  covary

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$$\Delta \ln(N) = \Delta \ln(\tau_a) \left. \frac{\partial \ln(N)}{\partial \ln(\tau_a)} \right|_{\text{met}} + (\Delta \text{met}) \left. \frac{\partial \ln(N)}{\partial (\text{met})} \right|_{\tau_a}$$

Cloud-aerosol  
sensitivity under  
constant  
“meteorology”

Cloud-“meteorology”  
sensitivity under  
constant aerosol

But  $N$  is also sensitive to meteorology. And meteorology and  $\tau_a$  covary

---

$$\Delta \ln(N) = \Delta \ln(\tau_a) \left. \frac{\partial \ln(N)}{\partial \ln(\tau_a)} \right|_{\text{met}} + (\Delta \text{met}) \left. \frac{\partial \ln(N)}{\partial (\text{met})} \right|_{\tau_a}$$

$$\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)} = \left. \frac{\partial \ln(N)}{\partial \ln(\tau_a)} \right|_{\text{met}} + \frac{d(\text{met})}{d \ln(\tau_a)} \left. \frac{\partial \ln(N)}{\partial (\text{met})} \right|_{\tau_a}$$



But  $N$  is also sensitive to meteorology. And meteorology and  $\tau_a$  covary

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$$\Delta \ln(N) = \Delta \ln(\tau_a) \left. \frac{\partial \ln(N)}{\partial \ln(\tau_a)} \right|_{\text{met}} + (\Delta \text{met}) \left. \frac{\partial \ln(N)}{\partial (\text{met})} \right|_{\tau_a}$$

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Total sensitivity

Partial sensitivity

Confounding meteorological effects

But  $N$  is also sensitive to meteorology. And meteorology and  $\tau_a$  covary

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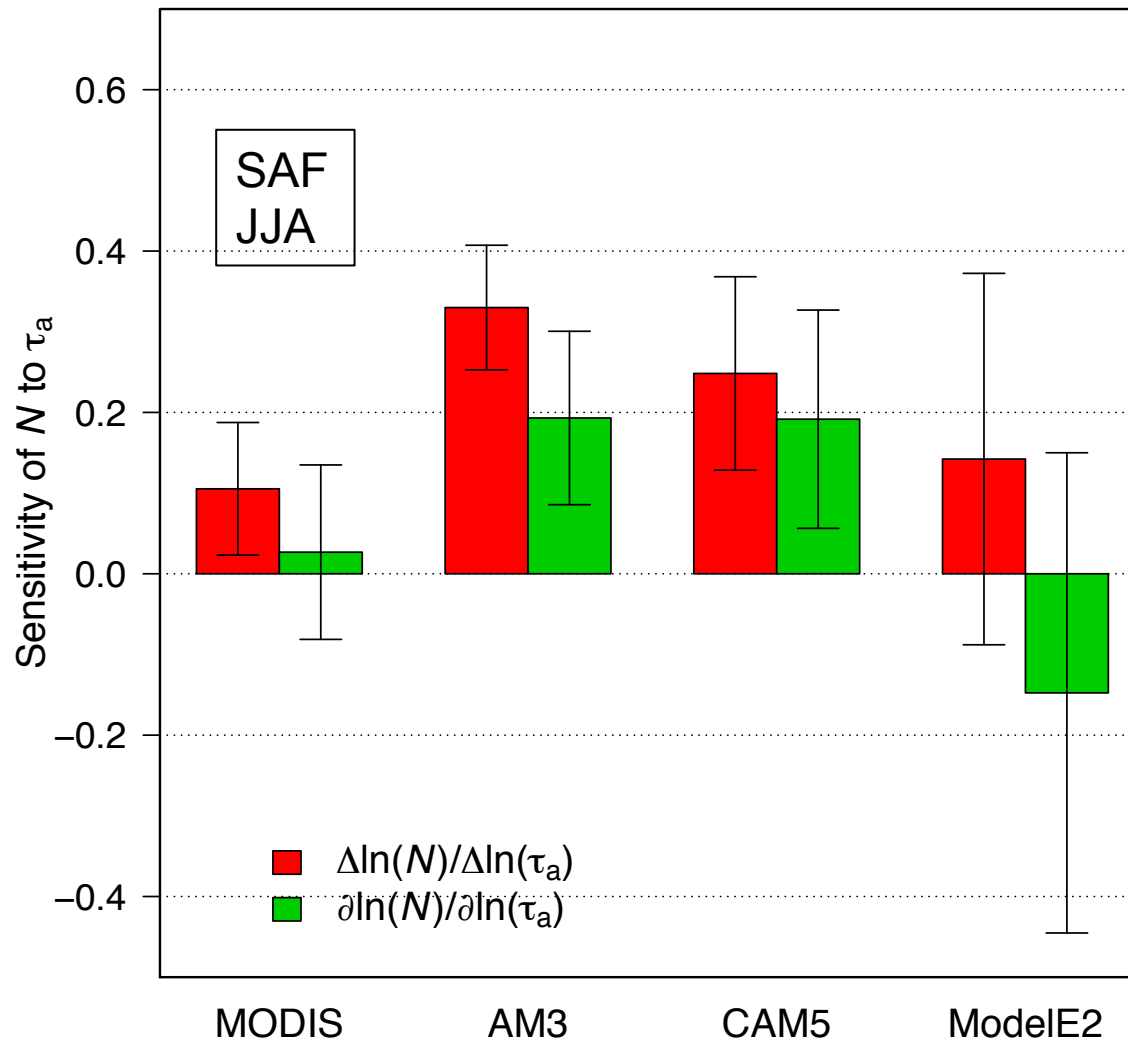
Total sensitivity

Partial sensitivity

Confounding meteorological effects

$met$  = relative humidity @ 1000 and 700 hPa

# Comparing sensitivity of $N$ to $\tau_a$ with and without considering met (RH in two layers)



*Ban-Weiss et al. (2014)*

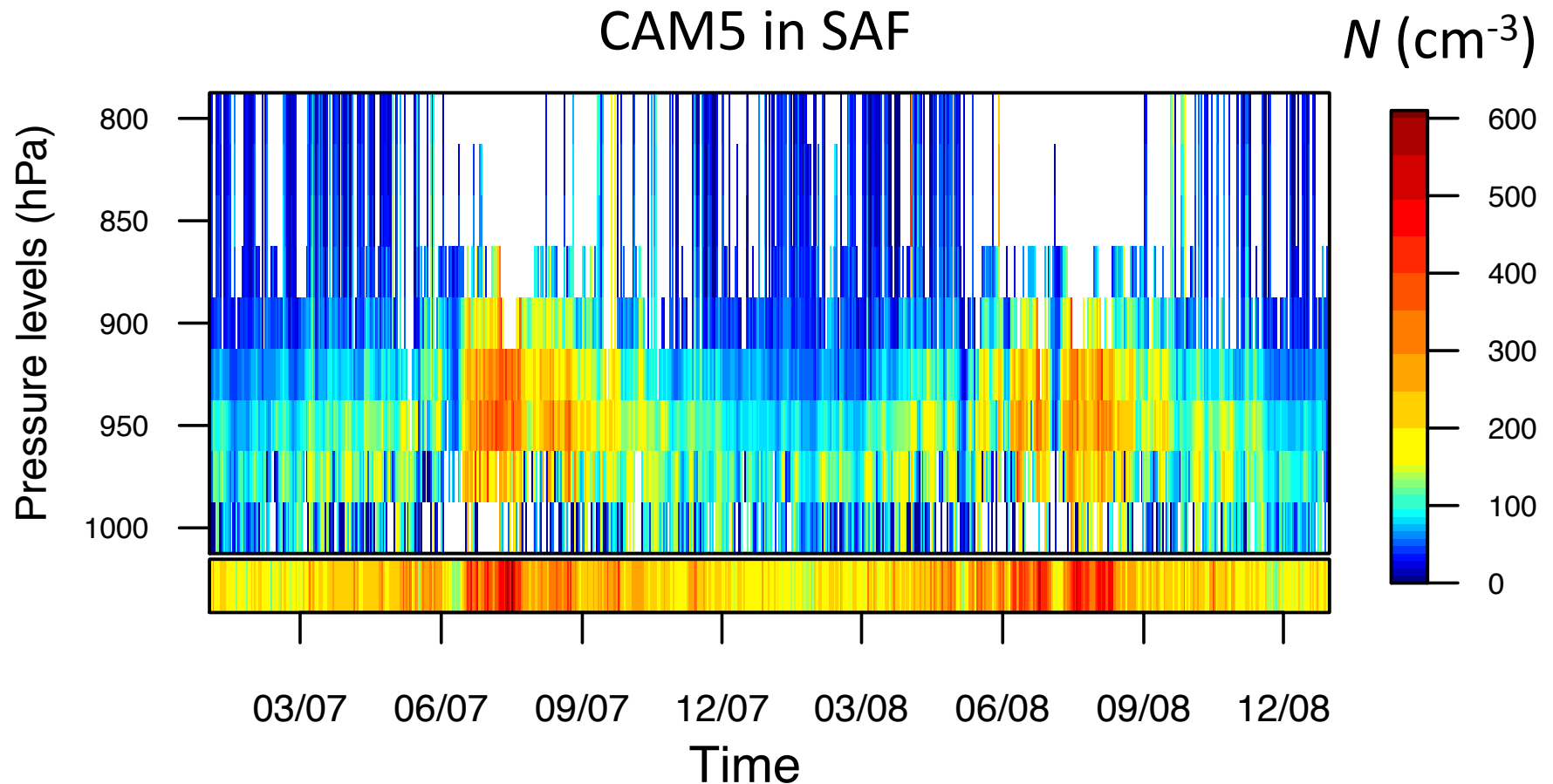
# Implications of the differences for total versus partial sensitivity

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- We should expect differences between observations and models in the “total” sensitivity  $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$  since models do not represent all confounding meteorological processes
- As aerosol-cloud parameterizations move away from empirical relationships, e.g.  $N = -598 + 298 \log(N_a)$ , and toward more physically realistic process descriptions, total sensitivity will change even for constant partial sensitivity  $\left. \frac{\partial \ln(N)}{\partial \ln(\tau_a)} \right|_{\text{met}}$

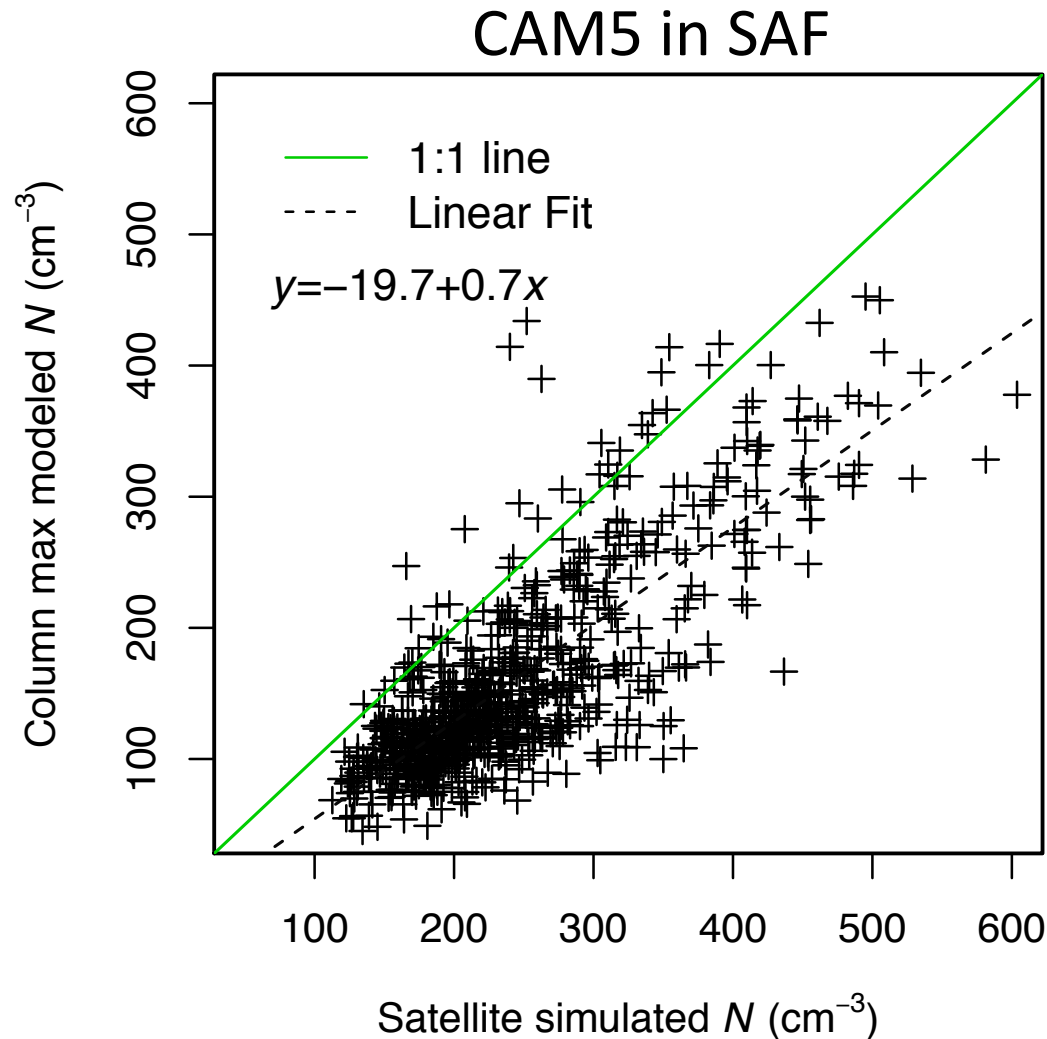
*Ban-Weiss et al. (2014)*

# Comparing satellite-simulated model values versus standard model output



*Ban-Weiss et al. (2014)*

# Standard model values are markedly lower than satellite-simulated values for $N$



Bias =  $83 \text{ cm}^{-3}$  \*\*

RMS =  $98 \text{ cm}^{-3}$

Bias ranges from  $55$  to  $115 \text{ cm}^{-3}$  for other models and regions

\*\*this bias is roughly half the difference between satellite simulated and observed values

*Ban-Weiss et al. (2014)*

# Implications: standard model values are nearer to observations than satellite-simulated values

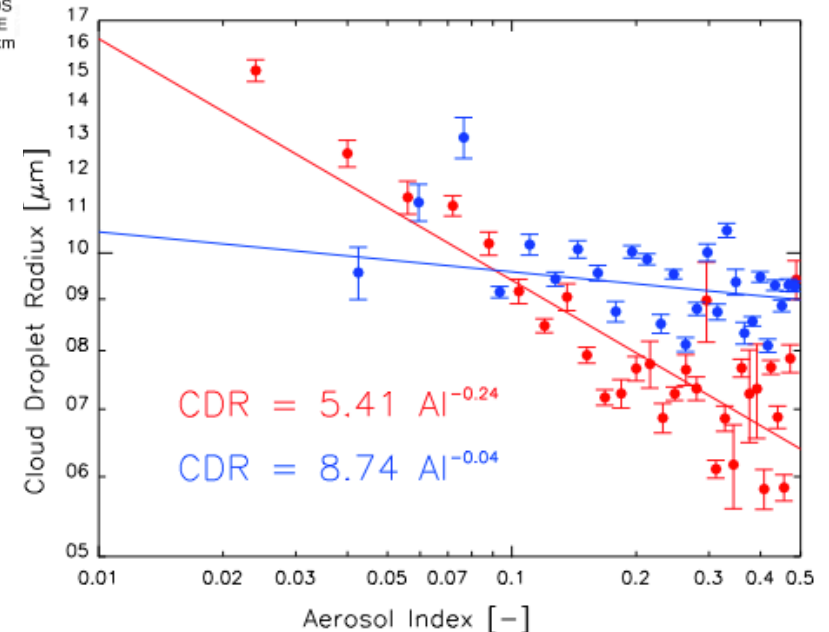
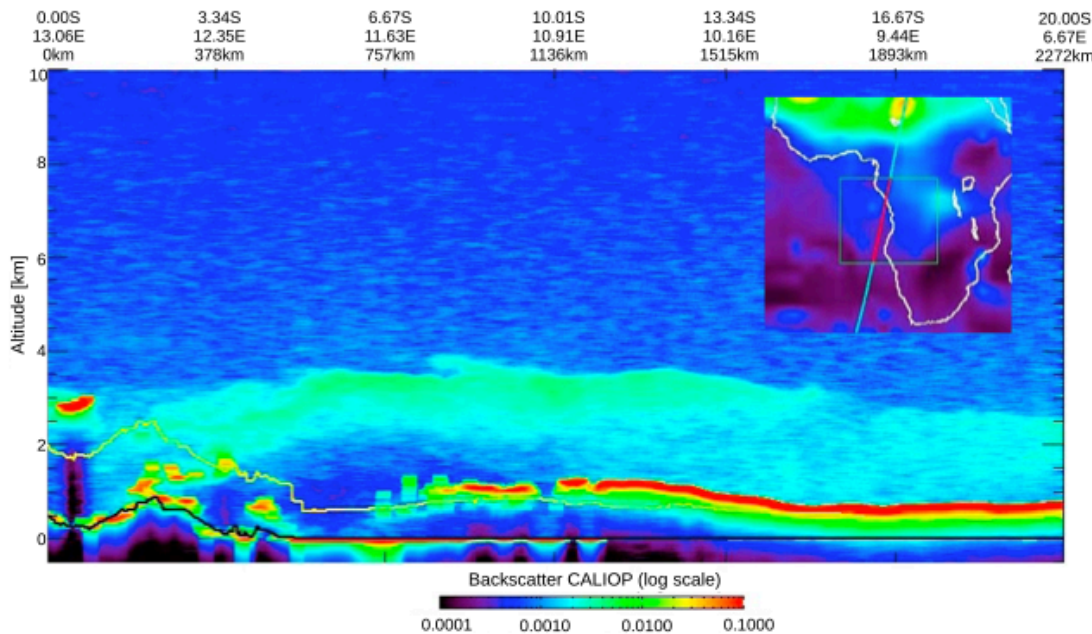
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- Unclear whether difference between satellite-simulated and standard model values are driven by differences in deriving  $\tau_c$  and  $r_e$ , or the algorithm for deriving  $N$
- May suggest that comparing  $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$  for observations versus standard model output is biased
- What is the right metric for comparing to observations?

*Ban-Weiss et al. (2014)*

# Cavaets and study limitations

- Is aerosol optical depth a good proxy for cloud condensation nuclei (CCN)?
- Are aerosols and clouds vertically collocated?



Source: Costantino and Breon (2010)



## See Ban-Weiss et al. (2014) JGR

- more regions and interpretations
- more detailed comparison of the differences in aerosol and cloud parameterizations between the models
- comparison to previous generations of the model

# Conclusions

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- Aerosol optical depth spatial pattern
  - CAM5, AM3, and ModelE2 match MODIS observed aerosol optical depth quite well
- Cloud droplet number concentration spatial pattern
  - CAM5 and AM3 capture the MODIS observed spatial pattern of  $N$
  - ModelE2 pattern is reversed
- Magnitude of  $N$  in 'SAF':
  - AM3 and ModelE2 are near MODIS observations
  - CAM5 overpredicts  $N$  (because of corresponding overpredictions in liquid  $\tau_c$  and underpredictions in  $r_e$ )

# Conclusions

- Sensitivity of  $N$  to  $\tau_a$   $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$  in 'SAF'
  - CAM5 and AM3 more sensitive than observations
  - ModelE2 has high uncertainty
- Covariation between  $N$ ,  $\tau_a$ , and meteorology confounds  $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$
- Differences in  $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$  between observations and models should be expected given the different degrees to which they represent these confounding factors.
- As parameterizations evolve,  $\frac{\Delta \ln(N)}{\Delta \ln(\tau_a)}$  may change even for constant aerosol-cloud interactions
- Satellite-simulated model values for  $N$  are much higher than standard model values (bias =  $83 \text{ cm}^{-3}$ ).