

# **PM2.5 vs AOD on Fine Time Scale**

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**<sup>2</sup>NASA GSFC**

# PM2.5 vs AOD

## ❖ From satellite observable to air quality application

### Global monitoring of air pollution over land from the Earth Observing System-Terra Moderate Resolution Imaging

#### Global Estimates of Ambient Fine Particulate Matter Concentrations from Satellite-Based Aerosol Optical Depth: Development and Application

Aaron van Donkelaar,<sup>1</sup> Randall V. Martin,<sup>1,2</sup> and Paul J. Villeneuve<sup>5,6</sup>

<sup>1</sup>Department of Physics and Atmospheric Science, Department of Physics and Atmospheric Science, David Dunlap Observatory, University of Toronto, Toronto, Ontario, Canada; <sup>2</sup>Department of Physics and Atmospheric Science, David Dunlap Observatory, University of Toronto, Toronto, Ontario, Canada; <sup>3</sup>NASA Goddard Space Flight Center, Greenbelt, Maryland, USA; <sup>4</sup>University of Toronto, Toronto, Ontario, Canada; <sup>5</sup>Population Studies Division, Statistics Canada, Ottawa, Ontario, Canada; <sup>6</sup>Department of Environmental Health Sciences, Dalla Lana School of Public Health, University of Toronto, Toronto, Ontario, Canada

#### and PM<sub>2.5</sub> mass: Implications for air quality forecasts

Jun Wang and Sundar A. Christopher

Department of Atmospheric Sciences, University of Washington, Seattle, Washington, USA

### IMPROVING NATIONAL AIR QUALITY FORECASTS WITH SATELLITE AEROSOL OBSERVATIONS

BY JASSIM AL-SAADI, JAMES SZYKMAN,\* R. BRADLEY PIERCE, CHIEKO KITTAKA, DOREEN NEIL, D. ALLEN CHU, LORRAINE REMER, LIAM GUMLEY, ELAINE PRINS,<sup>†</sup> LEWIS WEINSTOCK, AND JACK FISHMAN

## Remote Sensing of Particulate Pollution from Space: Have We Reached the Promised Land?

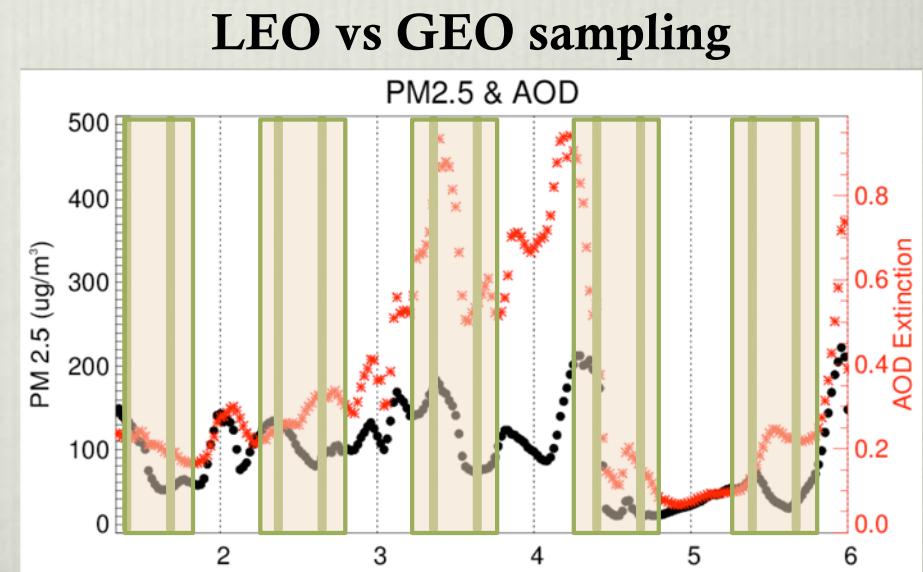
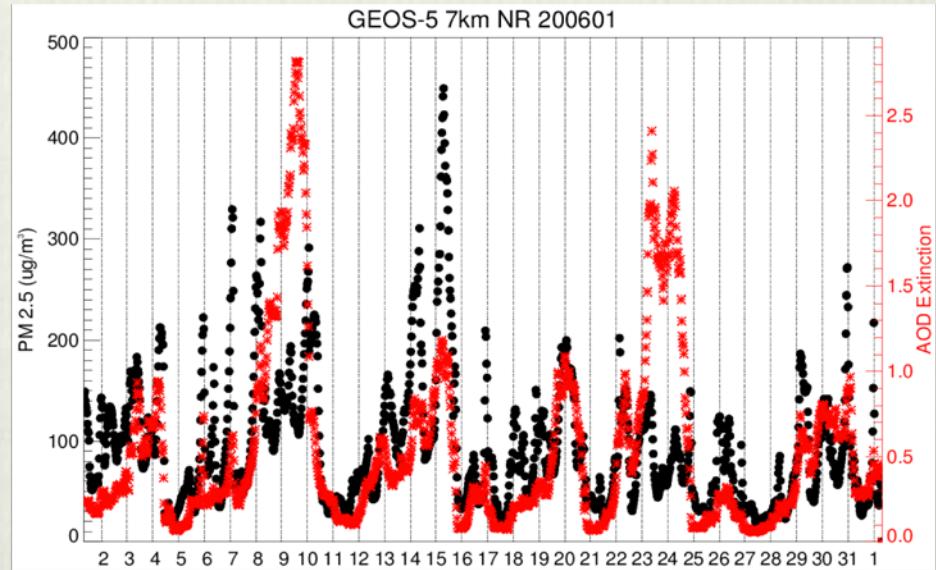
Raymond M. Hoff & Sundar A. Christopher

air

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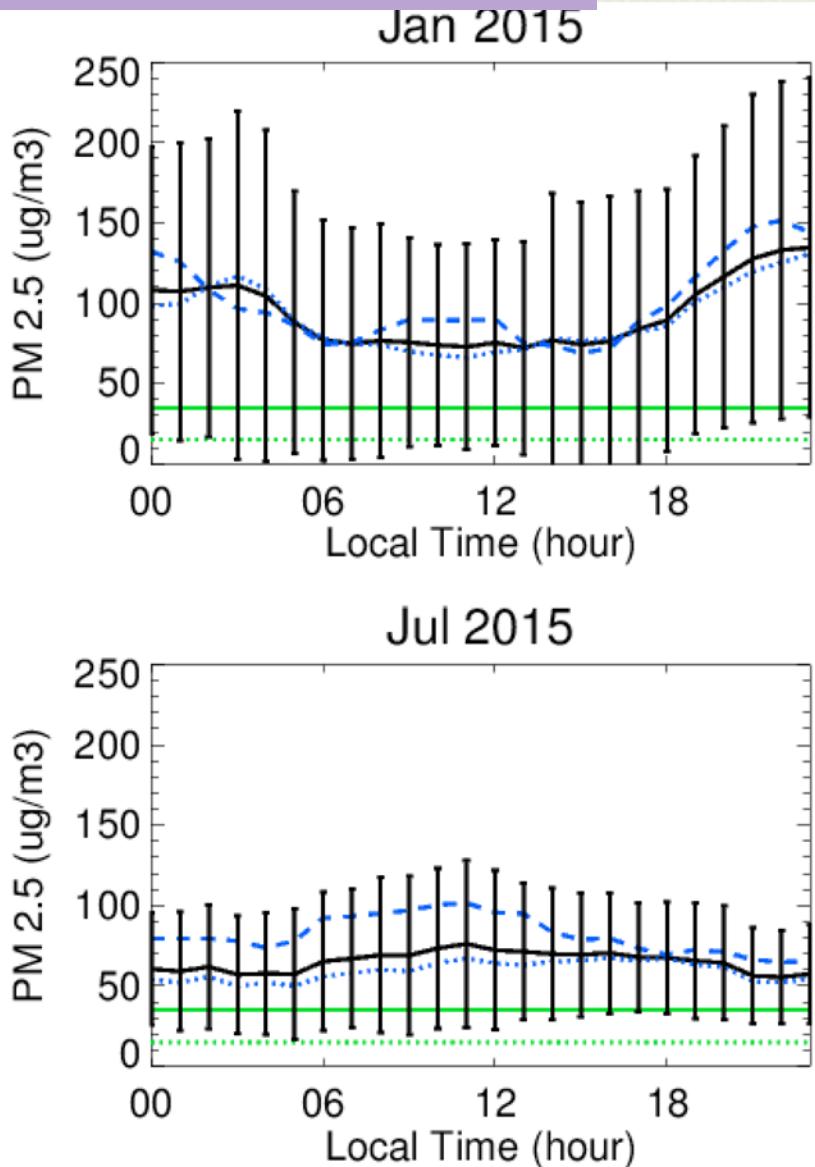
# Variation of PM2.5 & AOD

- ❖ Common factors
  - ❖ Synoptic systems
- ❖ Diurnal variation
  - ❖ Surface PM2.5
    - ❖ Emissions
    - ❖ Photochemical reactions
    - ❖ Boundary layer mixing
  - ❖ AOD
    - ❖ Mass loading of aerosols
    - ❖ Aerosol vertical distribution
    - ❖ H<sub>2</sub>O/RH diurnal cycle

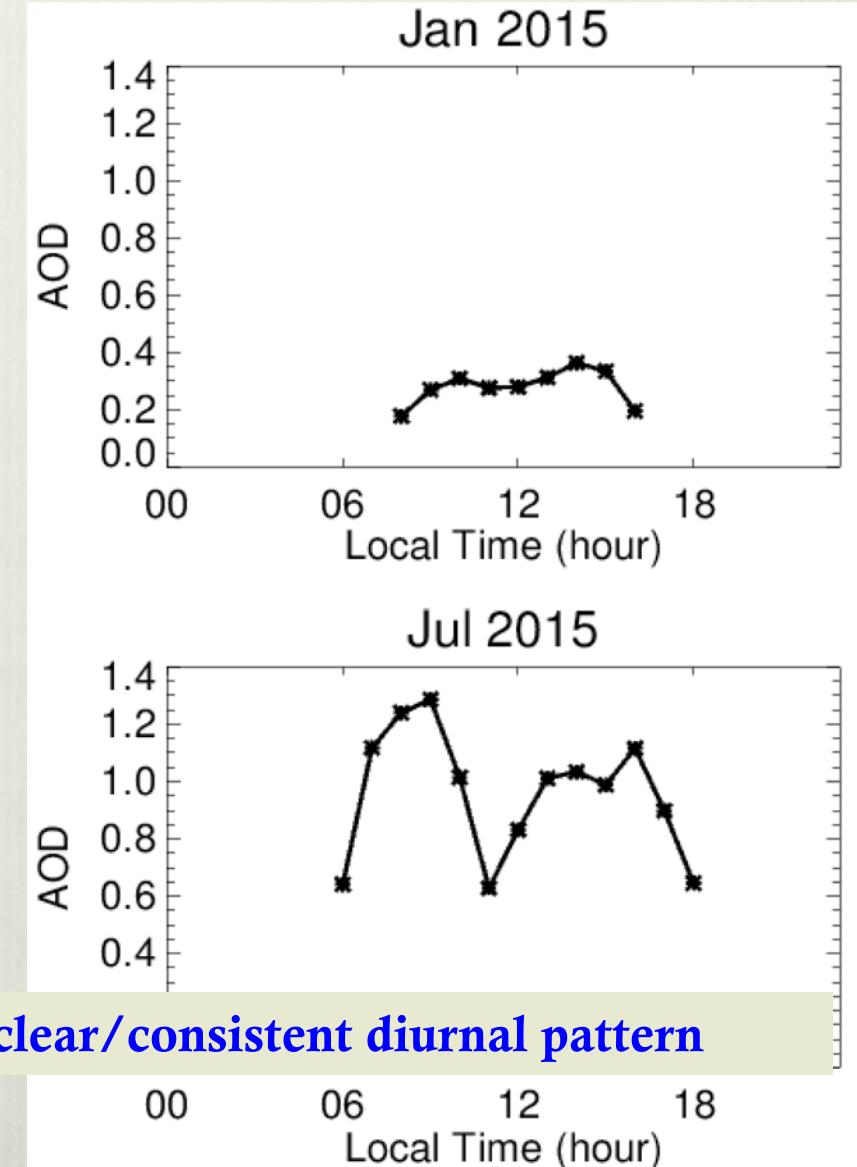


# Diurnal variations: Observed PM<sub>2.5</sub> & AOD

Result of PBL change?

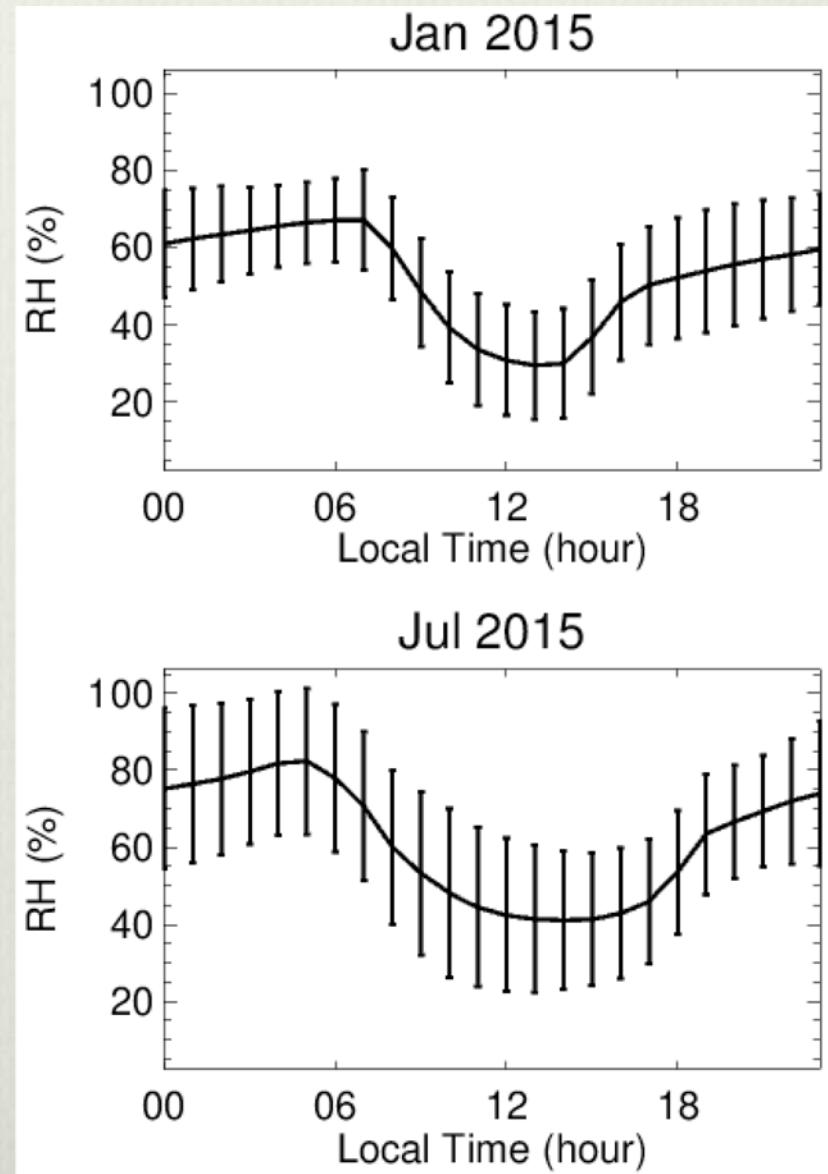
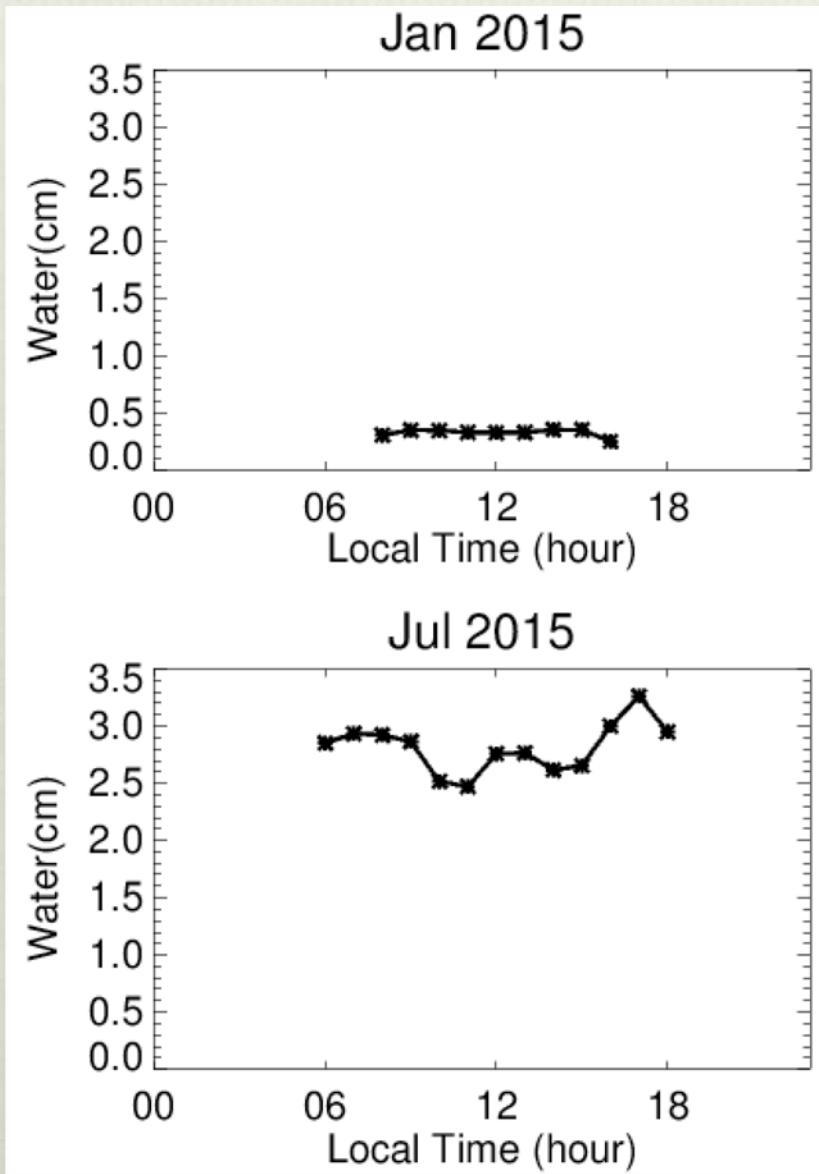


$$AOD = \int_{\text{surface}}^{\text{TOA}} \text{extinction}(z) dz$$



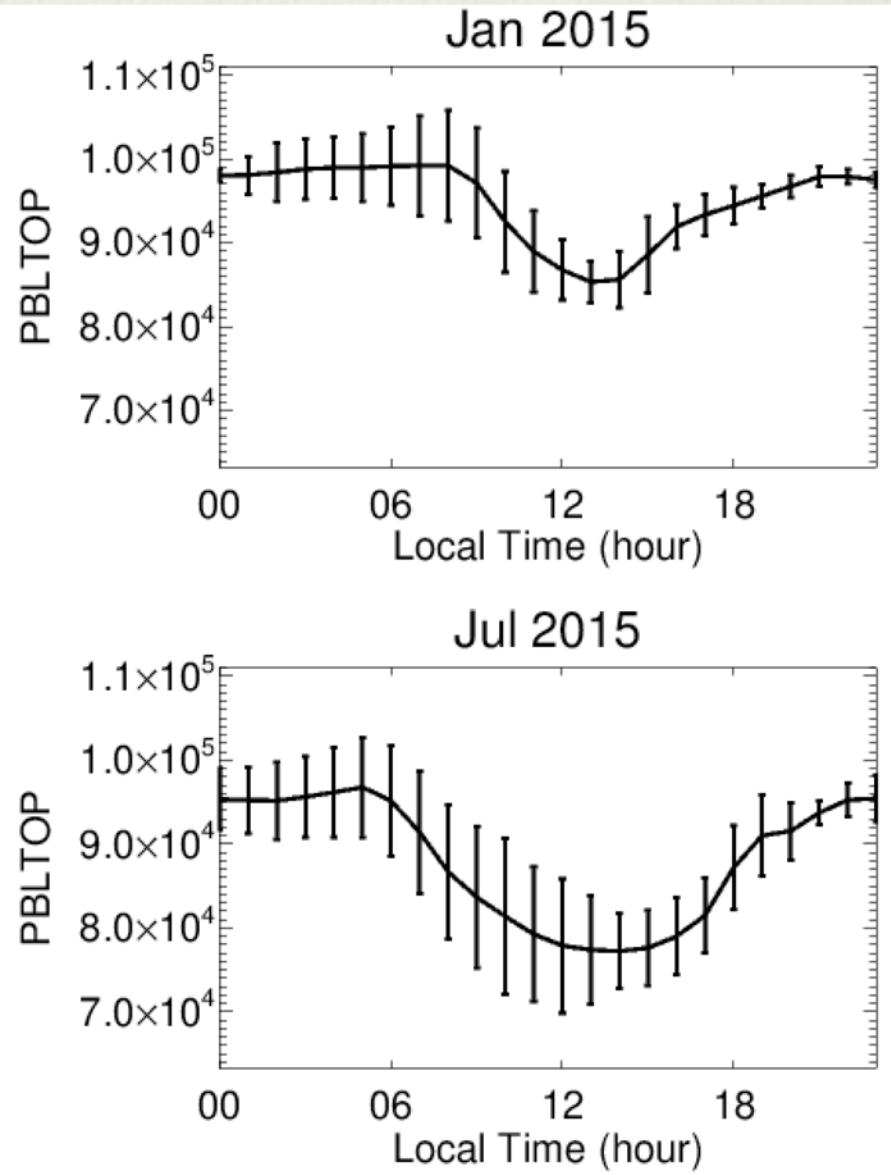
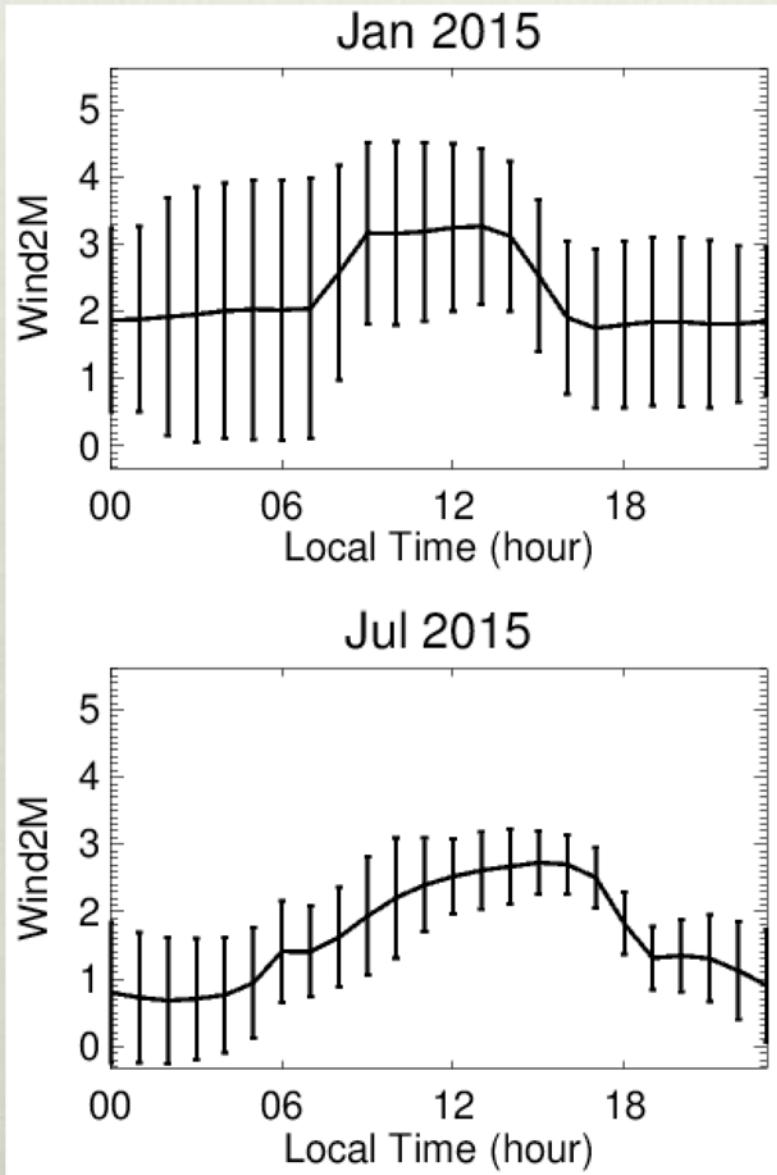
$$AOD \cong F(PM_{2.5_{surface}}, PM_{3D}, Optical, RH_{surface}, H_2O_{column}, PBL, rain, \text{cloud fraction, wind, etc.})$$

## Diurnal cycle: water column & surface RH



$$AOD \cong F(PM_{2.5_{surface}}, PM_{3D}, Optical, RH_{surface}, H_2O_{column}, PBL, rain, cloud fraction, wind, etc.)$$

## Diurnal cycle: wind & PBL



# PM2.5 vs AOD on higher temporal resolution

## ❖ Observations

- ❖ PM2.5 hourly observations
- ❖ AOT hourly observations
- ❖ Paired Sites: USA: GSFC, St. Louis, Fresno, Houston; Beijing (China)



## ❖ Model

- ❖ MERRA-2
- ❖ GEOS-5 7km

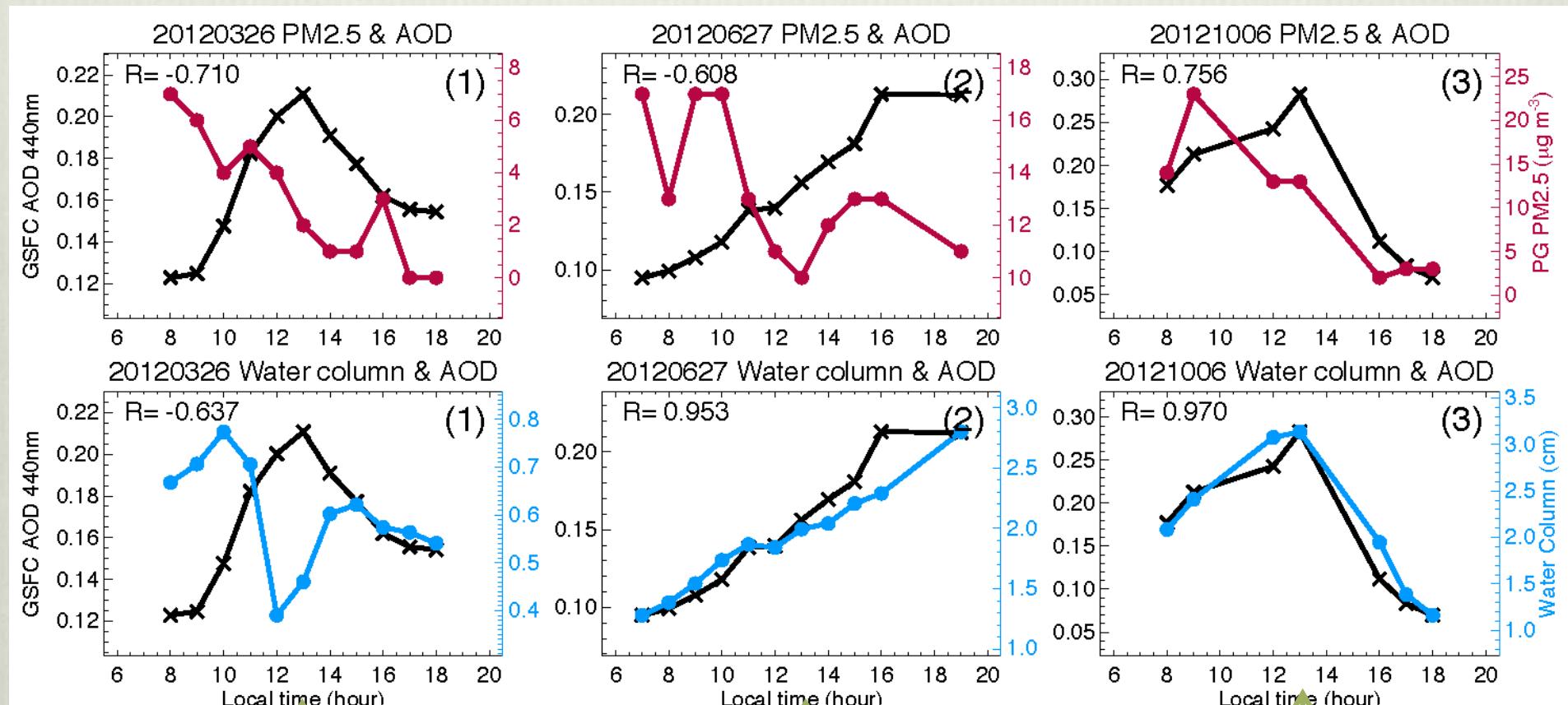
## ❖ Diurnal cycle

- ❖ PM2.5, AOD, meteorological fields.



# How are AOD and PM<sub>2.5</sub> correlated within a day?

- ❖ Example of daytime AERONET AOD and EPA PM<sub>2.5</sub> near GSFC



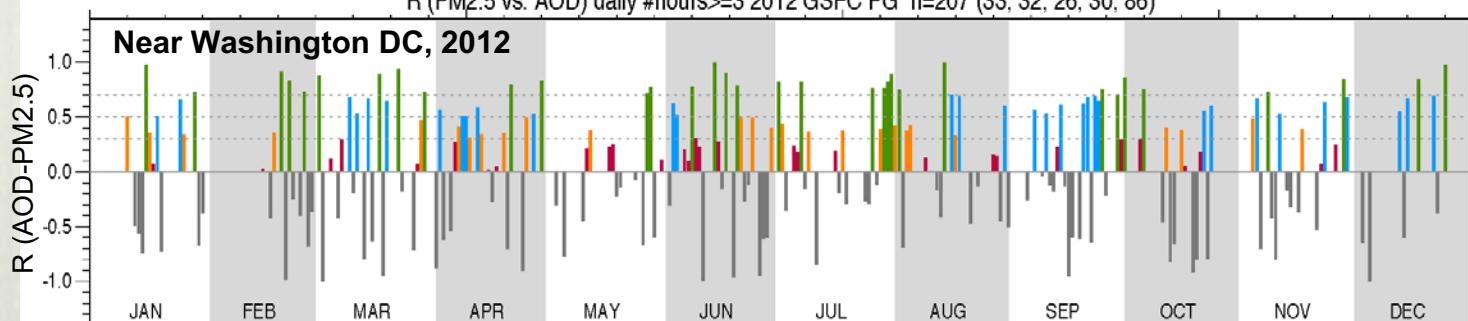
AOD negatively correlated with both **PM<sub>2.5</sub>** and **H<sub>2</sub>O(g)**

AOD negatively correlated with **PM<sub>2.5</sub>** but positively with **H<sub>2</sub>O(g)**

AOD positively correlated with both **PM<sub>2.5</sub>** and **H<sub>2</sub>O(g)**

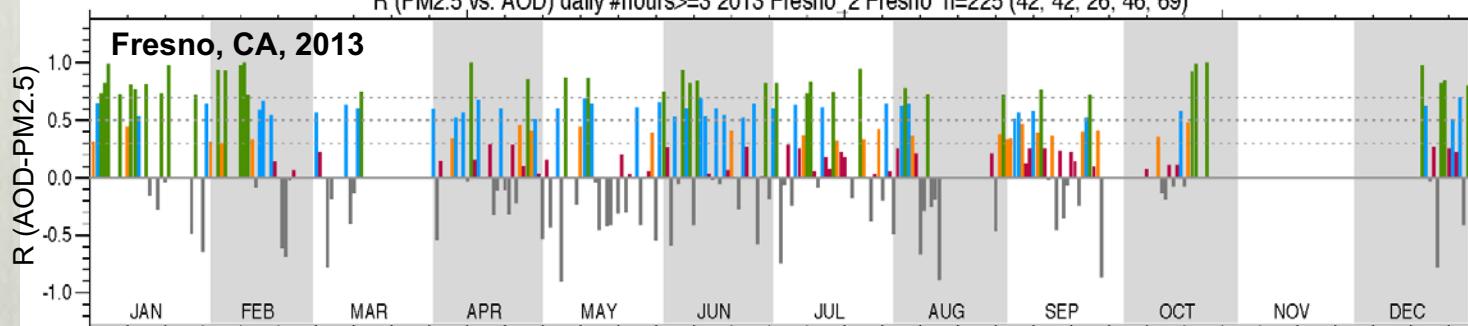
# Daily AOD-PM2.5 correlation coefficients in the four locations

R (PM2.5 vs. AOD) daily #hours>=3 2012 GSFC PG n=207 (33, 32, 26, 30, 86)



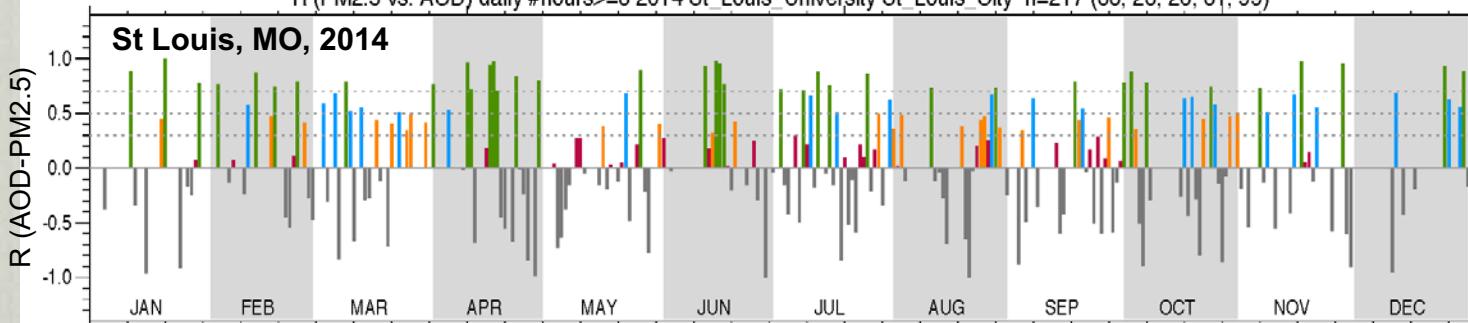
**Near Washington DC, 2012:**  
Total available days: 207  
→ 16% ( 33):  $R \geq 0.7$   
15% ( 32):  $0.5 \leq R < 0.7$   
13% ( 26):  $0.3 \leq R < 0.5$   
14% ( 30):  $0.0 \leq R < 0.3$   
42% ( 86):  $R < 0.0$

R (PM2.5 vs. AOD) daily #hours>=3 2013 Fresno\_2 Fresno n=225 (42, 42, 26, 46, 69)



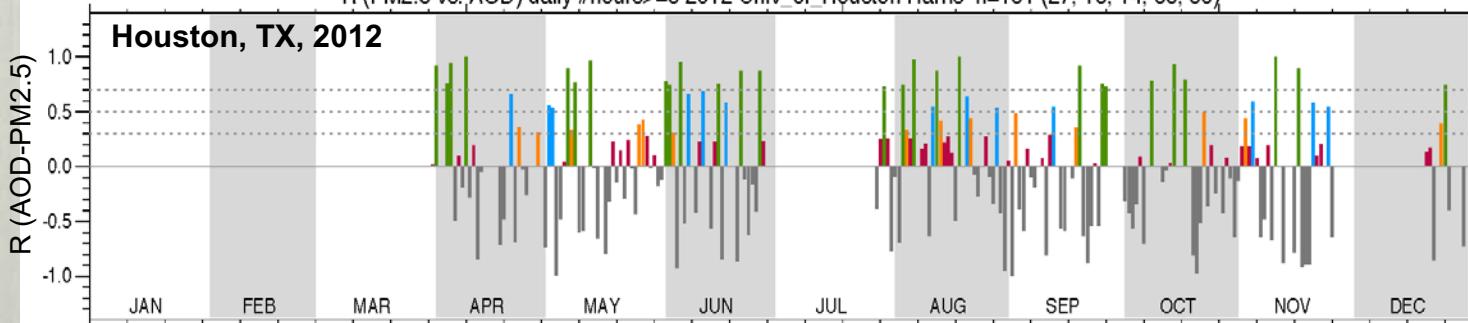
**Fresno, 2013:**  
Total available days: 225  
→ 19% ( 42):  $R \geq 0.7$   
19% ( 42):  $0.5 \leq R < 0.7$   
11% ( 26):  $0.3 \leq R < 0.5$   
20% ( 46):  $0.0 \leq R < 0.3$   
31% ( 69):  $R < 0.0$

R (PM2.5 vs. AOD) daily #hours>=3 2014 St\_Louis\_University St\_Louis\_City n=217 (38, 23, 26, 31, 99)



**St. Louis, 2014:**  
Total available days: 217  
→ 17% ( 38):  $R \geq 0.7$   
11% ( 23):  $0.5 \leq R < 0.7$   
12% ( 26):  $0.3 \leq R < 0.5$   
14% ( 31):  $0.0 \leq R < 0.3$   
46% ( 99):  $R < 0.0$

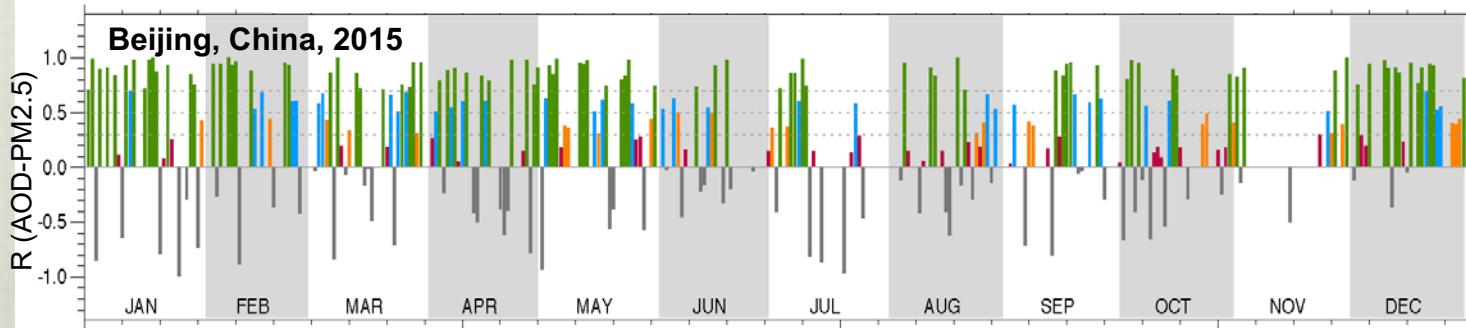
R (PM2.5 vs. AOD) daily #hours>=3 2012 Univ\_of\_Houston\_Harris n=181 (27, 13, 14, 38, 89)



**Houston, 2012:**  
Total available days: 181  
→ 15% ( 27):  $R \geq 0.7$   
7% ( 13):  $0.5 \leq R < 0.7$   
8% ( 14):  $0.3 \leq R < 0.5$   
21% ( 38):  $0.0 \leq R < 0.3$   
49% ( 89):  $R < 0.0$

# Daily correlation coefficients in Beijing, China, 2015

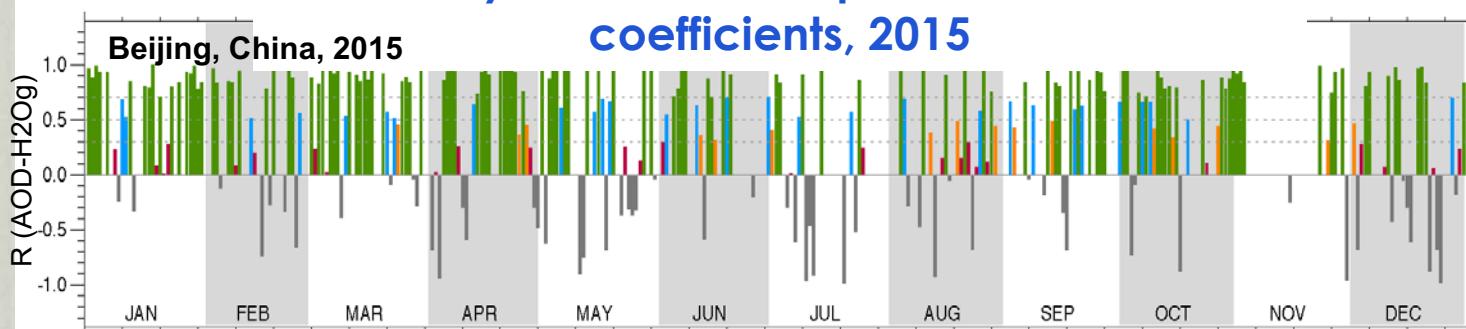
**Daily AOD-PM2.5 vapor correlation coefficients, 2015**



**Beijing, AOD-PM2.5, 2015:**  
**Total available days: 212**

- 36% ( 92):  $R \geq 0.7$
- 14% ( 35):  $0.5 \leq R < 0.7$
- 10% ( 25):  $0.3 \leq R < 0.5$
- 15% ( 38):  $0.0 \leq R < 0.3$
- 25% ( 64):  $R < 0.0$

**Daily AOD-water vapor correlation coefficients, 2015**



**Beijing, AOD-water, 2015:**  
**Total available days: 256**

- 50% (127):  $R \geq 0.7$
- 11% ( 29):  $0.5 \leq R < 0.7$
- 6% ( 16):  $0.3 \leq R < 0.5$
- 10% ( 26):  $0.0 \leq R < 0.3$
- 22% ( 58):  $R < 0.0$

- For all three years (2014-2016) and two pairs of Beijing sites, 30-40% days AOD and PM2.5 are correlated with  $R \geq 0.7$  and 25-33% days  $R < 0$  – better correlations than sites in the US
- For AOD-water vapor, 40-60% days they are correlated with  $R \geq 0.7$  and 20-25% days  $R < 0$  – Comparable the sites in the US

**Closer to the large sources, PM2.5 plays more role in AOD changes**

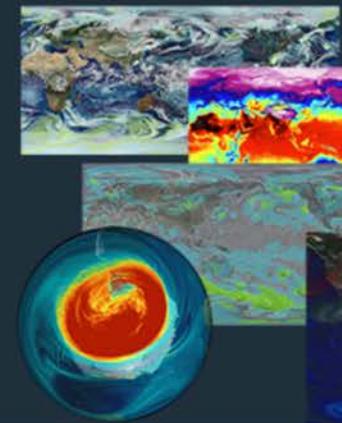
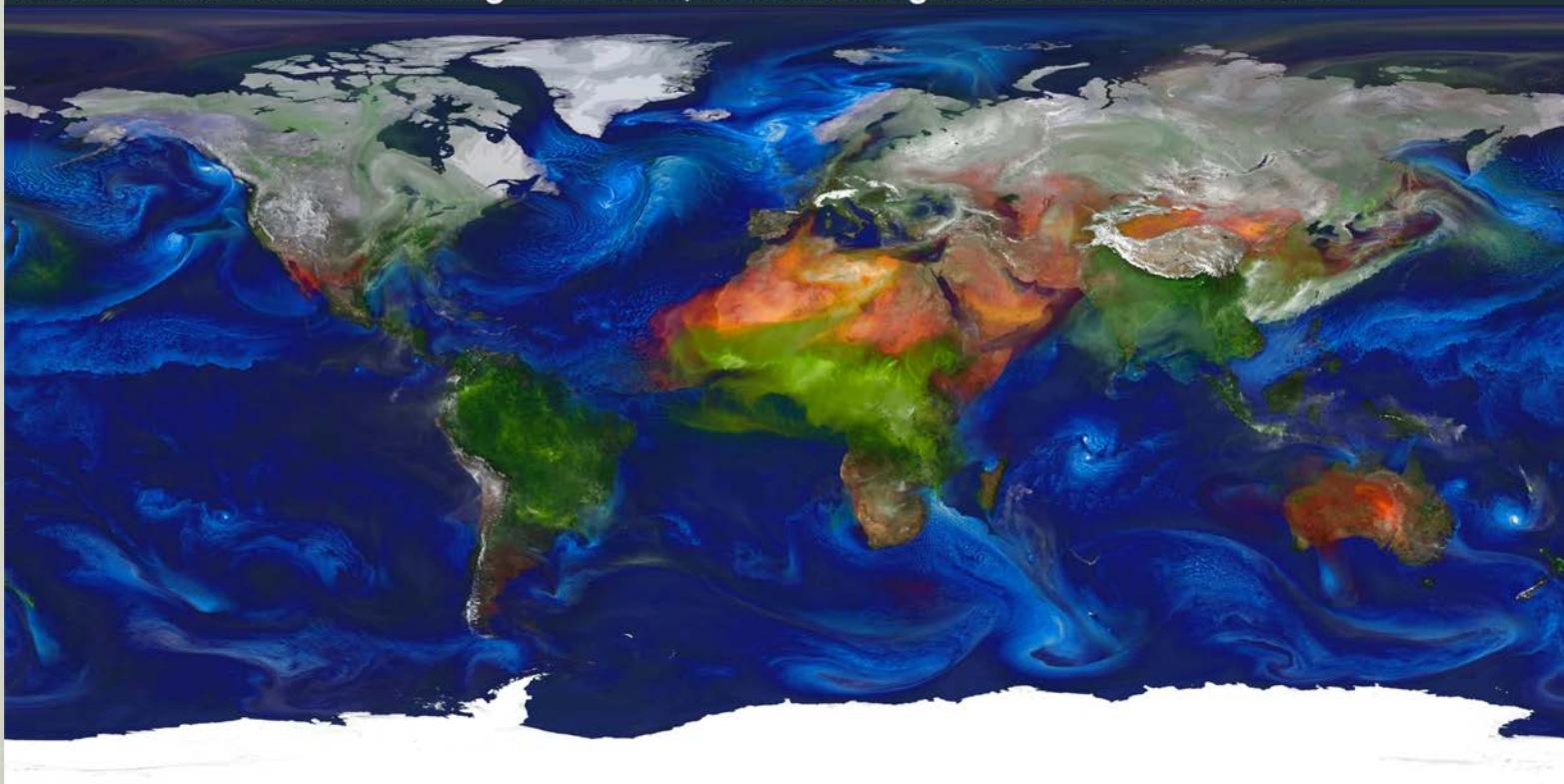
## GEOS-5 Nature Run, Ganymed Release

## What does model say?

### Non-hydrostatic 7 km Global Mesoscale Simulation

The GEOS-5 Nature Run (Ganymed Release) is a 2-year global, non-hydrostatic mesoscale simulation for the period June 2005 through May 2007 with a 7 km horizontal resolution. In addition to standard meteorological parameters (wind, temperature, moisture, surface pressure), this simulation includes 15 aerosol tracers (dust, seasalt, sulfate, black and organic carbon), O<sub>3</sub>, CO and CO<sub>2</sub>.

This model simulation is driven by prescribed sea-surface temperature and sea-ice, daily volcanic and biomass burning emissions, as well as high-resolution inventories of

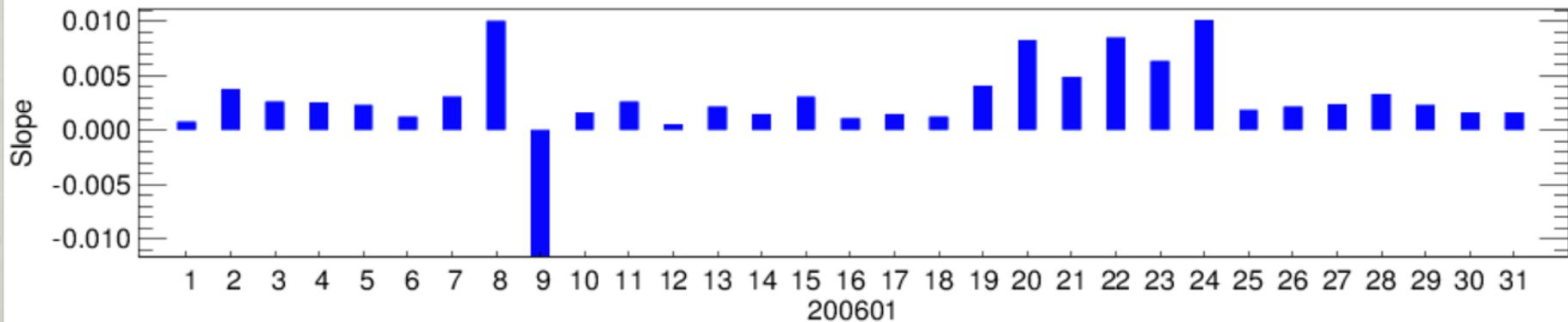
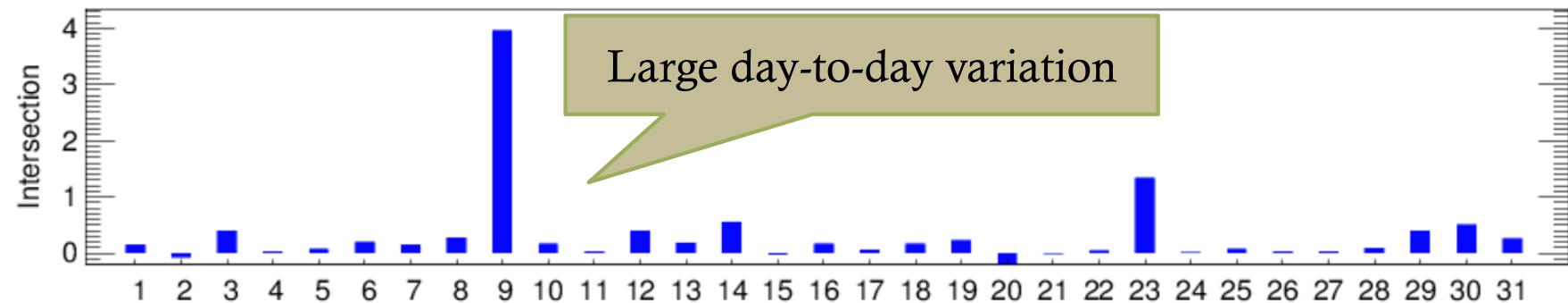
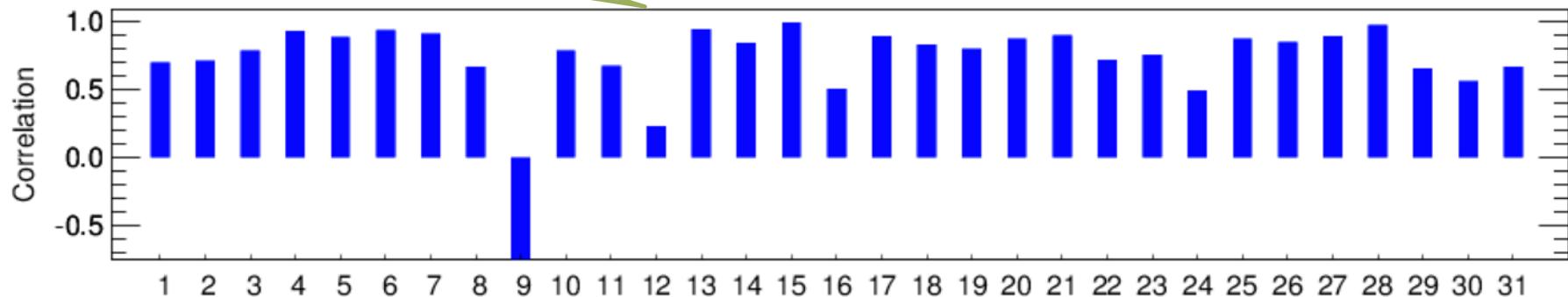


7km, 0.0625°  
2005/05-2007/06  
30 min  
Natural run, no assimilation

# R/intersection/slope of daily PM2.5 vs AOD

R change day-to-day  
It can be anti-correlated

2006-01

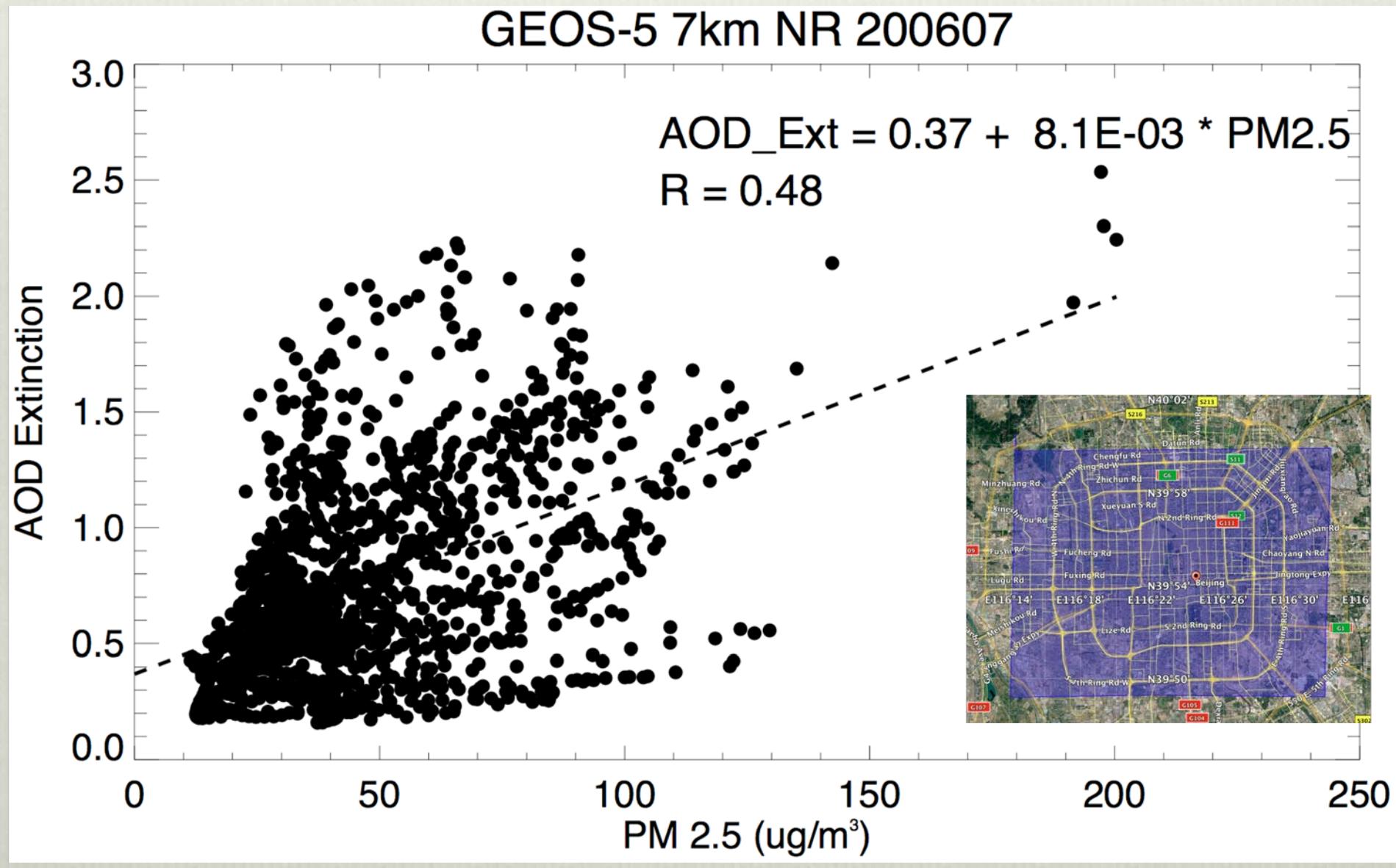


$$AOD \cong F(PM_{2.5_{surface}}, PM_{3D}, Optical, RH_{surface}, H_2O_{column}, PBL, rain, cloud fraction, wind, etc.)$$

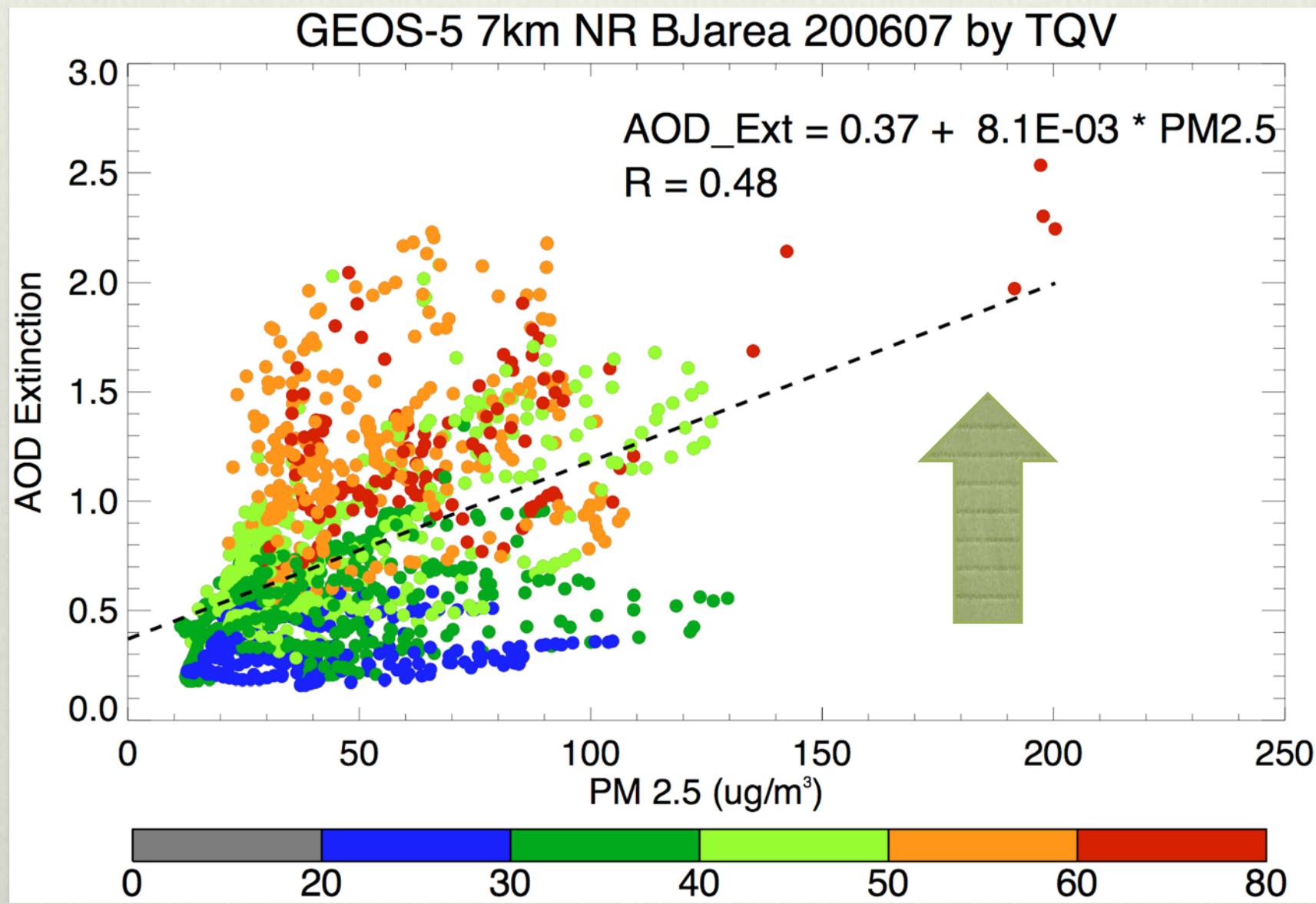
# Variables checked

CAPE	CLDTOT	CWP	IWP	LWP	PBLH
convective available potential energy	cloud fraction	condensed water path	ice water path	liquid water path	PBL height
PRECTOT	PS	QLML	QV10M	RH	SLP
total precipitation	surface pressure	surface specific humidity	10 m specific humidity	relative humidity	sea level pressure
T10M	TAUTOT	TQI	TQL	TQV	U10M
10 meter temperature	total cloud optical depth	total precipitable ice water	total precipitable liquid water	total precipitable water vapor	10 meter wind
U50M	V10M	V50M	W850	Wind10M	Wind50M
50 meter wind	10 meter wind	50 m wind	Wind @ 850mb	10 meter wind speed	50 meter wind speed

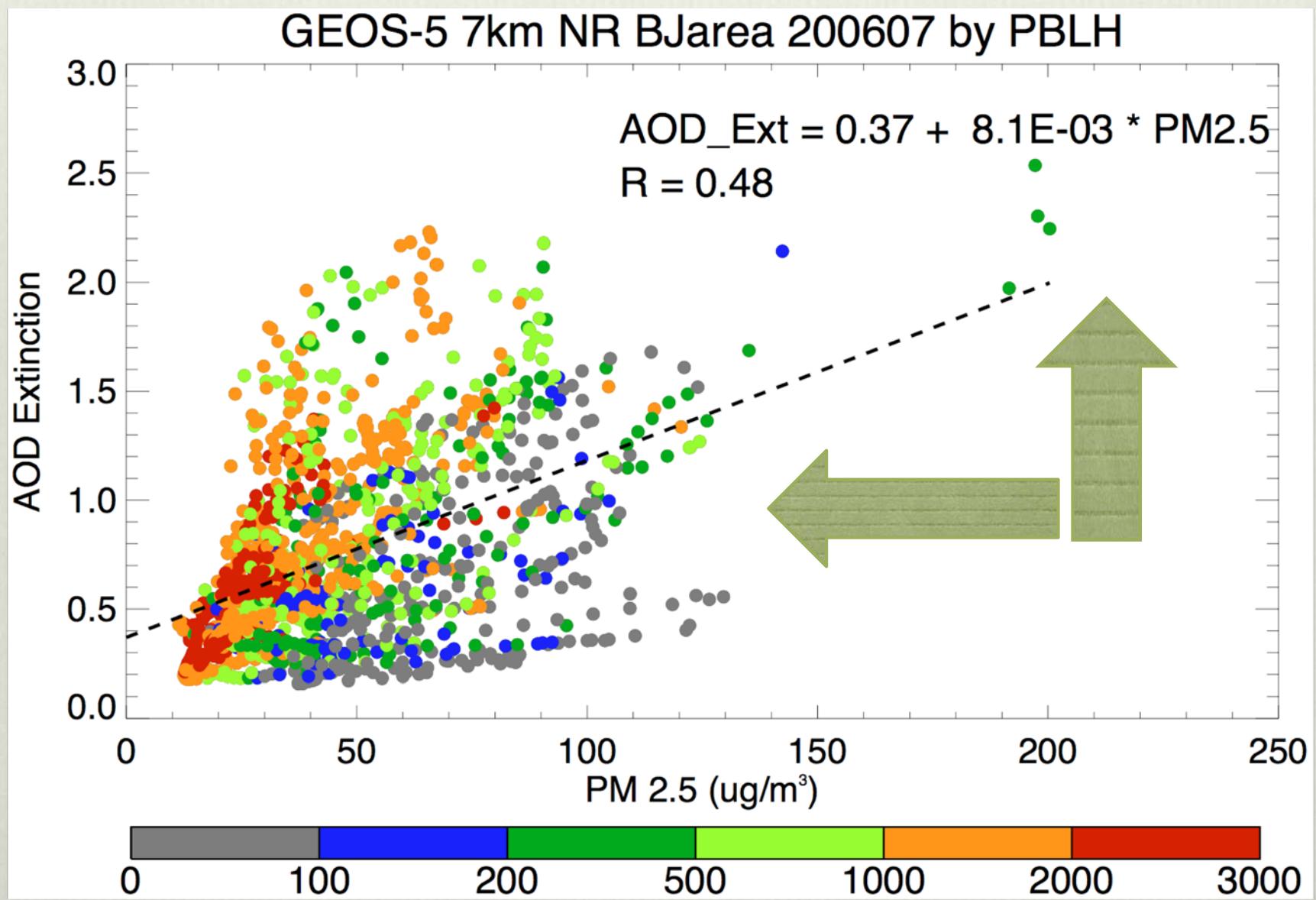
# Regional average around Beijing



Same PM2.5, more total column H<sub>2</sub>O vapor, higher AOD

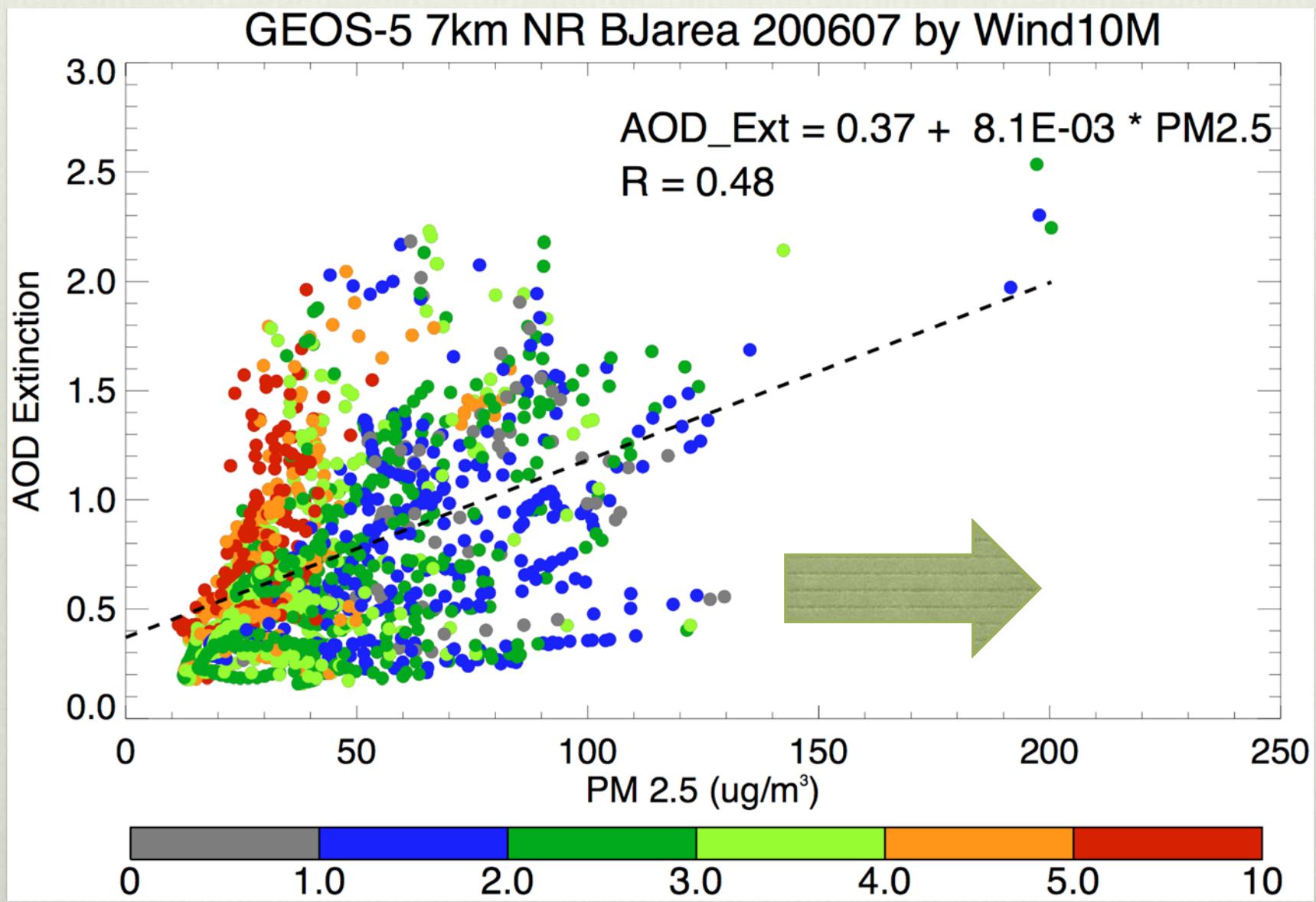


Same AOD: higher PBL → less PM<sub>2.5</sub>  
Same PM<sub>2.5</sub>: higher PBL → higher AOD

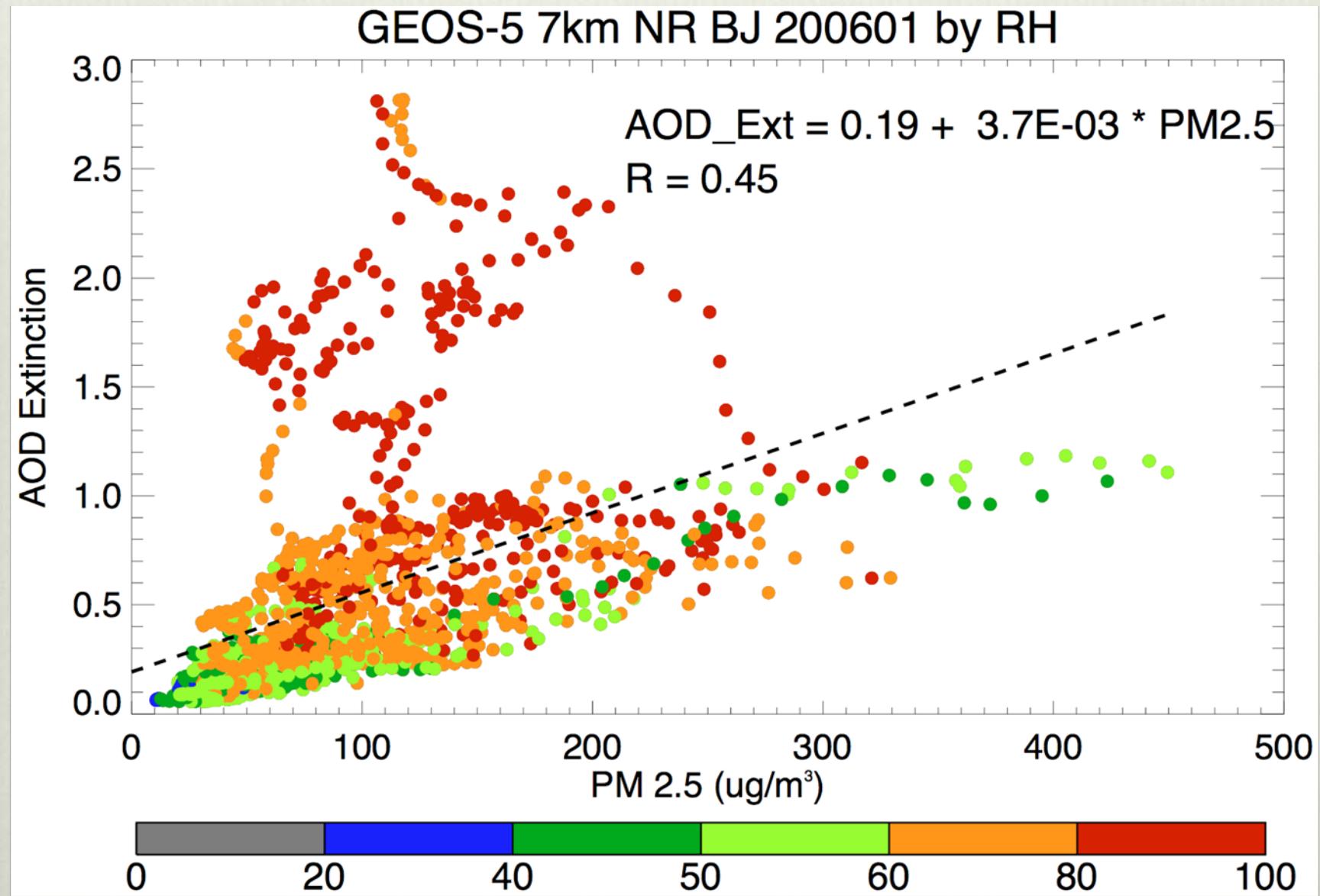


# More stagnation → more PM2.5

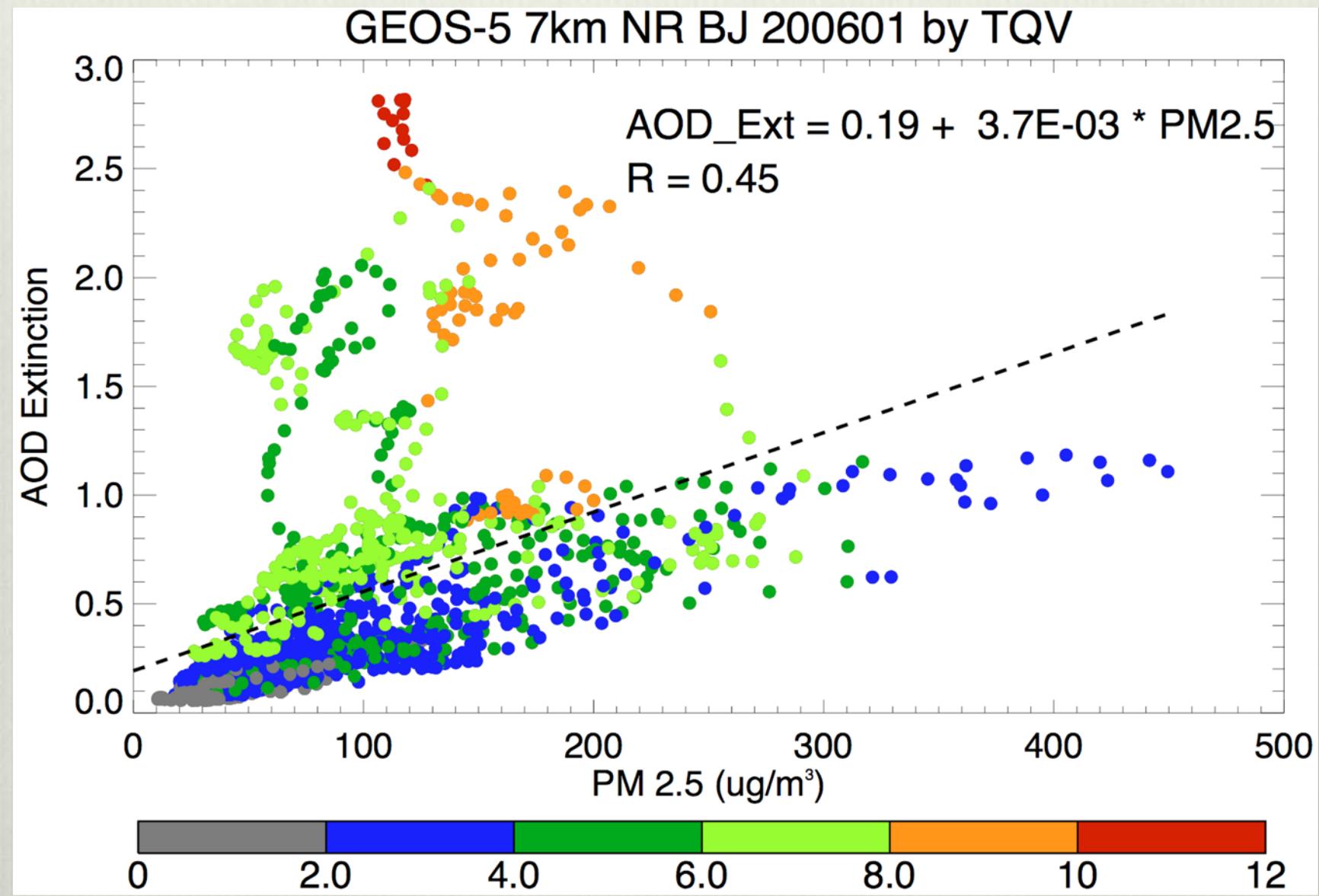
Wind speed is correlated to RH & TQv



# Total column Qv vs Surface RH

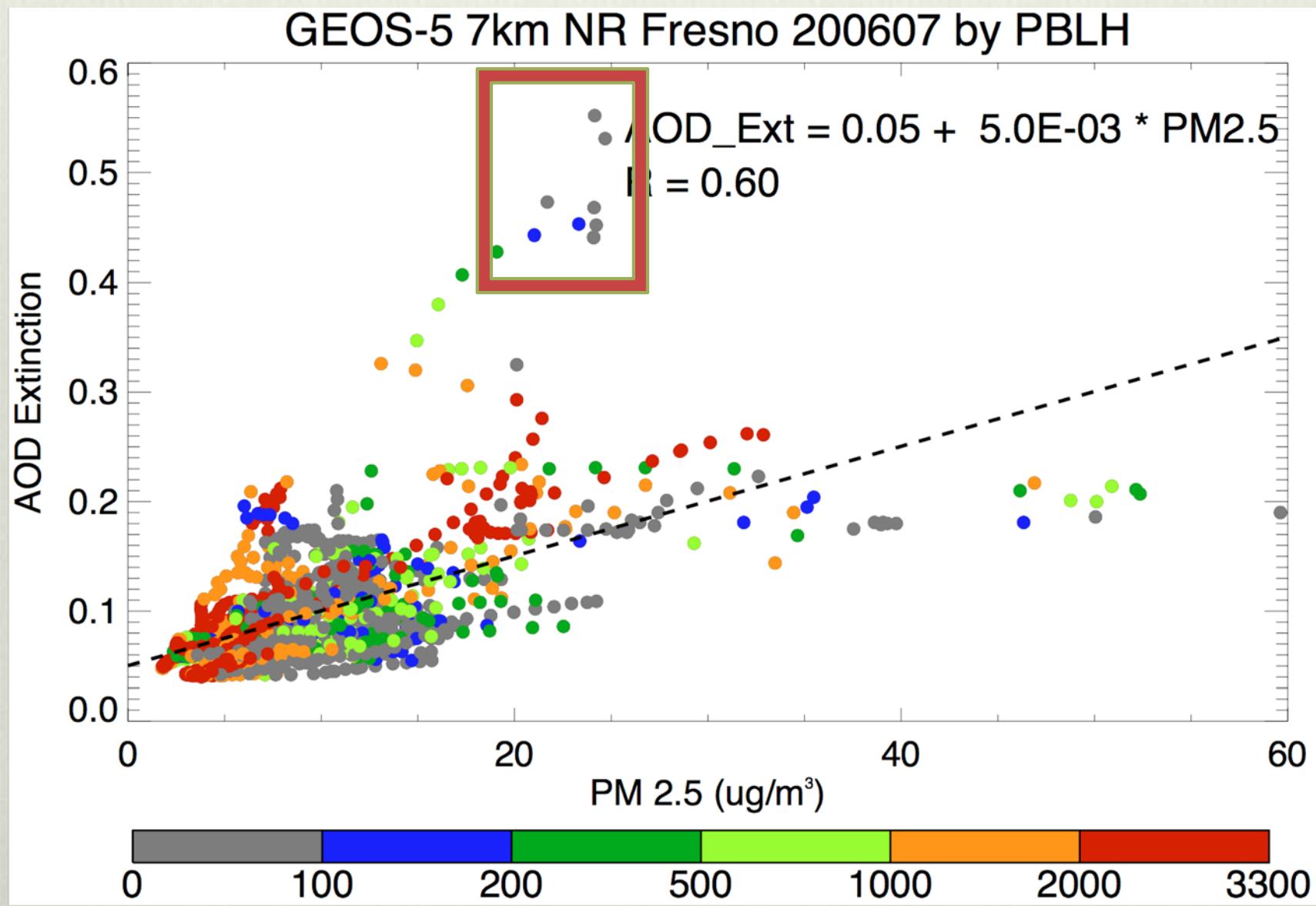


# Total column Qv vs Surface RH



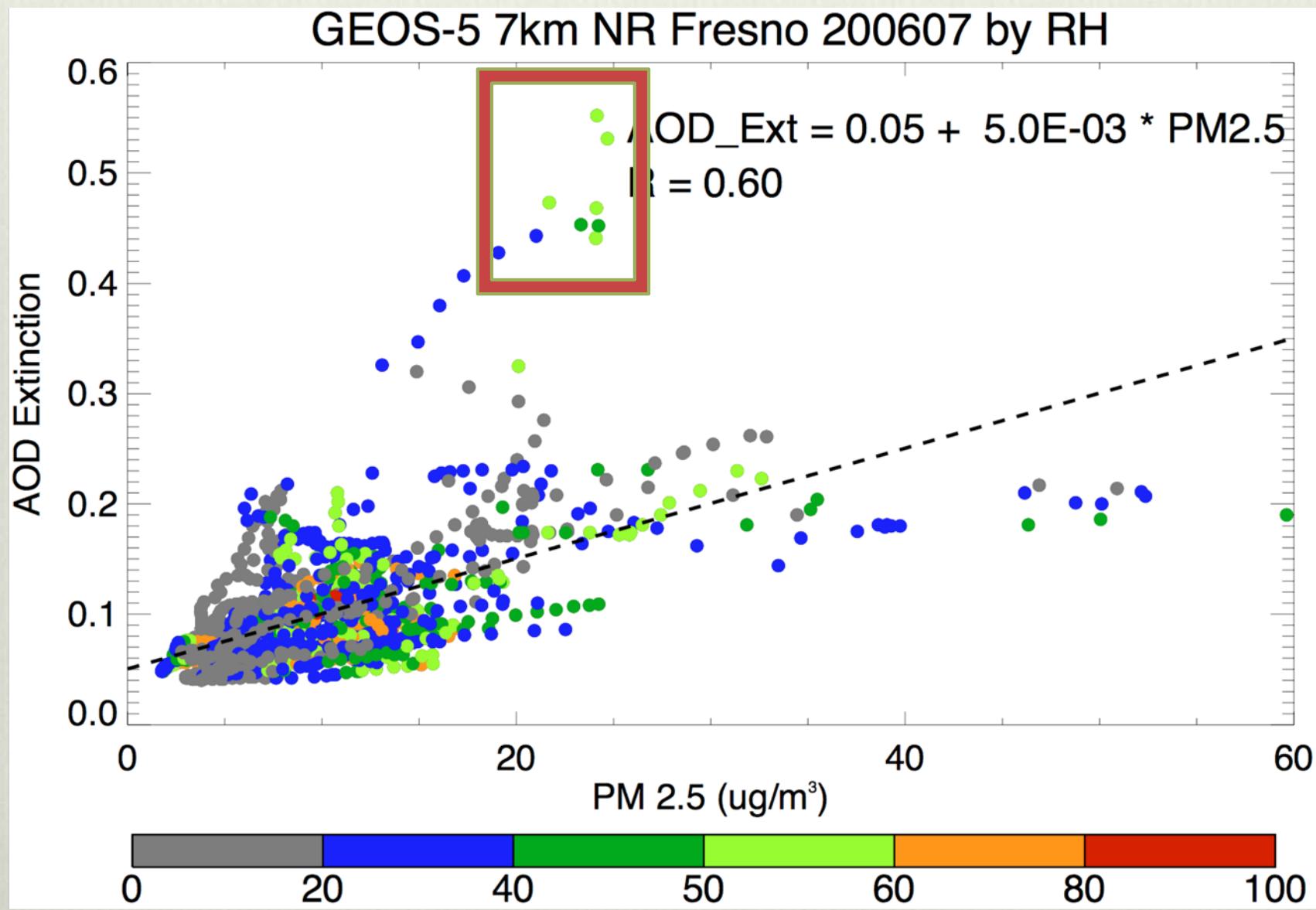
# Fresno, summer (dry)

## Same PM2.5, lower PBL, higher AOD (?)



# Fresno, summer (dry)

## RH plays bigger role in AOD for this case



# Conclusion

- ❖ AOD is affected by aerosol loading and multiple meteorological conditions
- ❖ Within a day, AOD vs PM2.5 correlation change with time and location.
- ❖ Near major source regions, relative humidity + H<sub>2</sub>O column, PBL height, wind speed, are needed to estimate PM2.5 from AOD.
- ❖ In other regions, information on aerosol composition, vertical distribution etc, are required.

# *Estimate PM2.5 Using Mixed Effects Models*

Robert Chatfield<sup>1</sup>, Meytar H. Sorek<sup>1</sup>

<sup>1</sup>NASA Ames Research Center

# Using MAIAC Aerosol Optical Thickness and Water Vapor

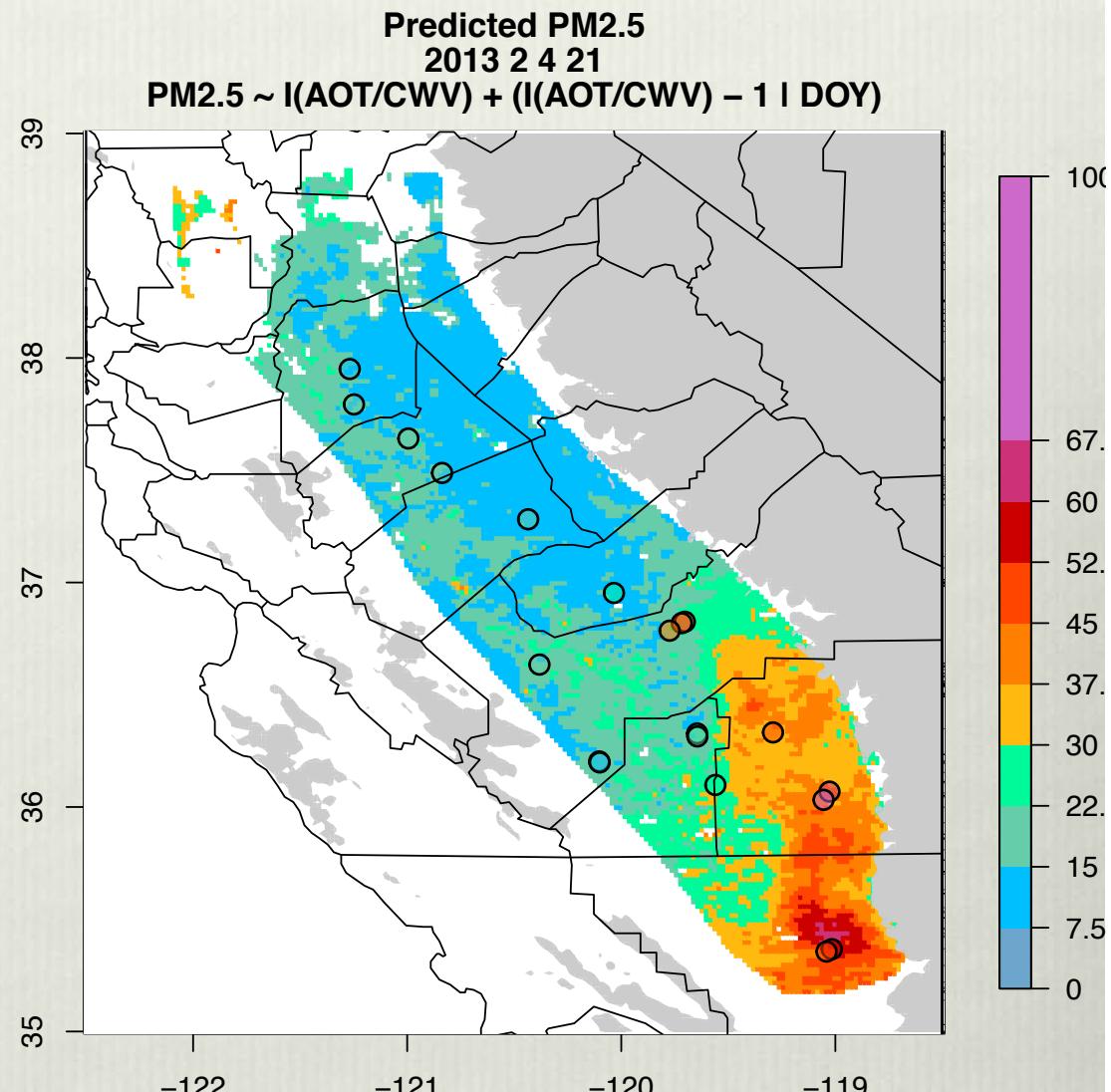
Shown is a map of a major smog-aerosol episode in the San Joaquin Valley.

*Colors inside the circles show that PM<sub>2.5</sub> stations are generally close to the mapped values, within rms error.*

For the winter, 11/2012–03/2013  
Area-wide estimation of PM<sub>2.5</sub>, in  $\mu\text{g m}^{-3}$   
at 1 km resolution, has ...

*rms residual error = 7.3  $\mu\text{g m}^{-3}$  and  $R^2 = 0.73$*

**Limitations:**  
*accuracy of satellite AOT, CWV,  
also very local effects  
and peculiarities of the reference  
method (e.g. response to volatiles)*



*Mixed effects constrained by many stations*

*GEO instruments allow Mixed Effects Models to be constrained by observations many times of day  
(multiple observations help distinguish “signal” from “noise”)*

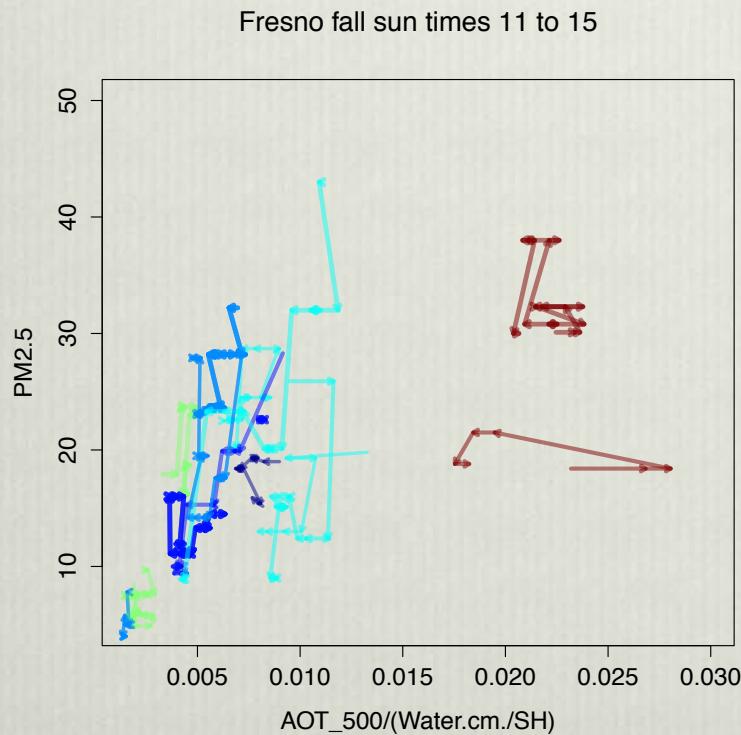
Arrows showing time variation of the predictor and PM2.5

By making a separate calibration each day, the accuracy improves (bottom right)

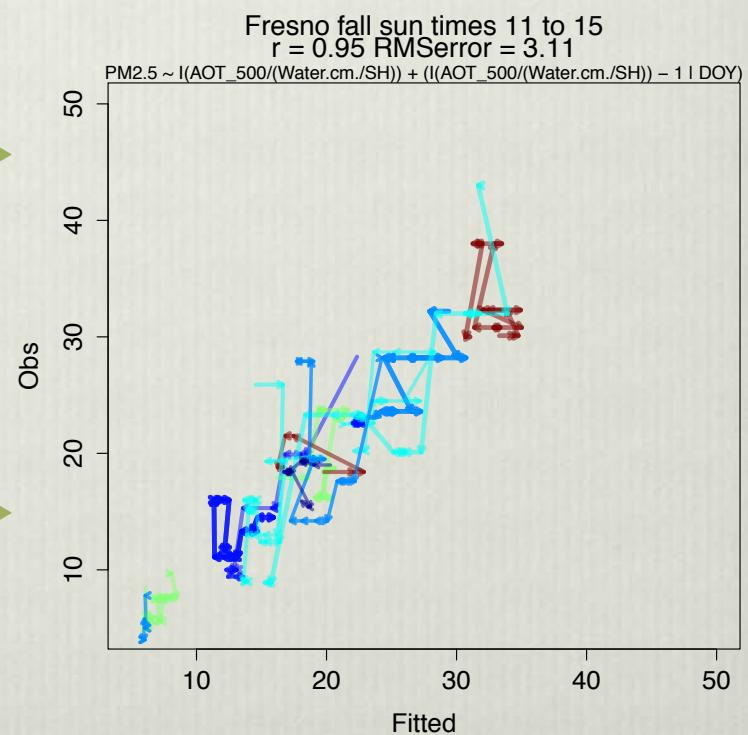
*Why one day?* Meteorology (subsidence, winds, mixing) is often correlated over a day's period, and composition is too.

**AERONET** estimates of AOT and CWV do have better precision than current MAIAC.

These **Fresno, CA, 2007–2016** estimates have *rms residual error of  $\sim 3 \text{ } \mu\text{g m}^{-3}$*  over many years and seasons rather than  $\sim 7$  seen for MODIS MAIAC for this and (likely) other reasons.

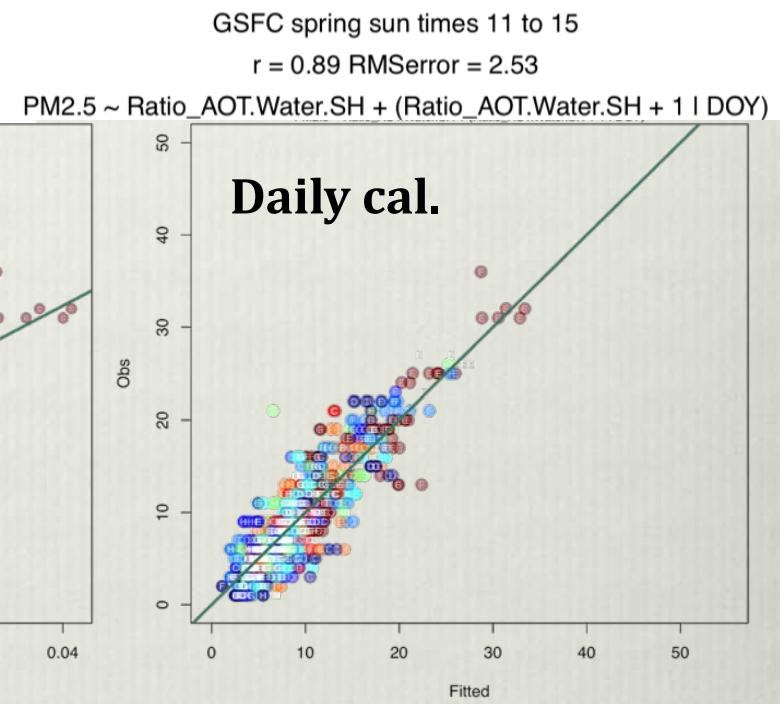
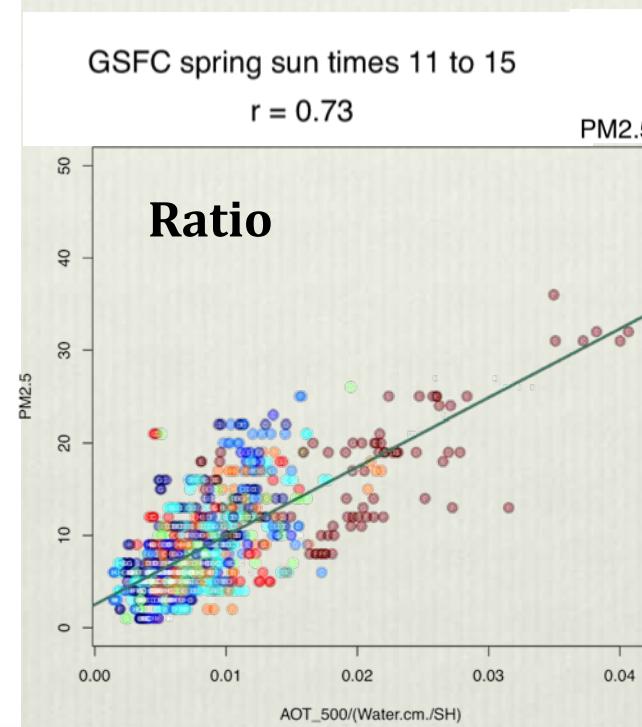
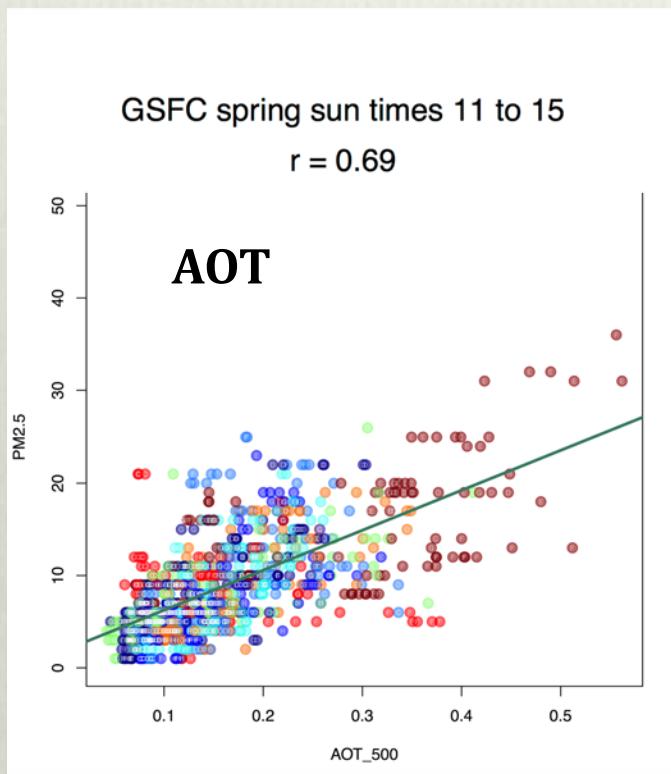


Calibrate sensitivity every day (mixed effects model)



**Note the progressive improvement as we estimate mixed depth using CWV, then use daily calibration of sensitivity**

**Comparing one AERONET at GSFC to one monitor at Beltsville, MD.**



**Moving from afternoon hours comparison shown (fully mixed PBL) to all hours of observation, ... still remarkably good results.  
i.e., 0.63 (AOT) -> 0.69 (Ratio) -> 0.82 (fitted)  
(Why? There should be layers not mixing to surface.)**

## **Limitations of multiple observations per day**

As we have seen, there is variation during the day, especially RH and very local processes adding new particles to several-day loadings.

A definition of “homogenous region” is needed to make maps for each part of the country.

## **Benefits:**

Although detailed GEO imagery *may not* (*yet!*) **improve** our desired **within-day estimation** of PM<sub>2.5</sub> variation, the additional observations **may still help** distinguish signal from noise day to day.

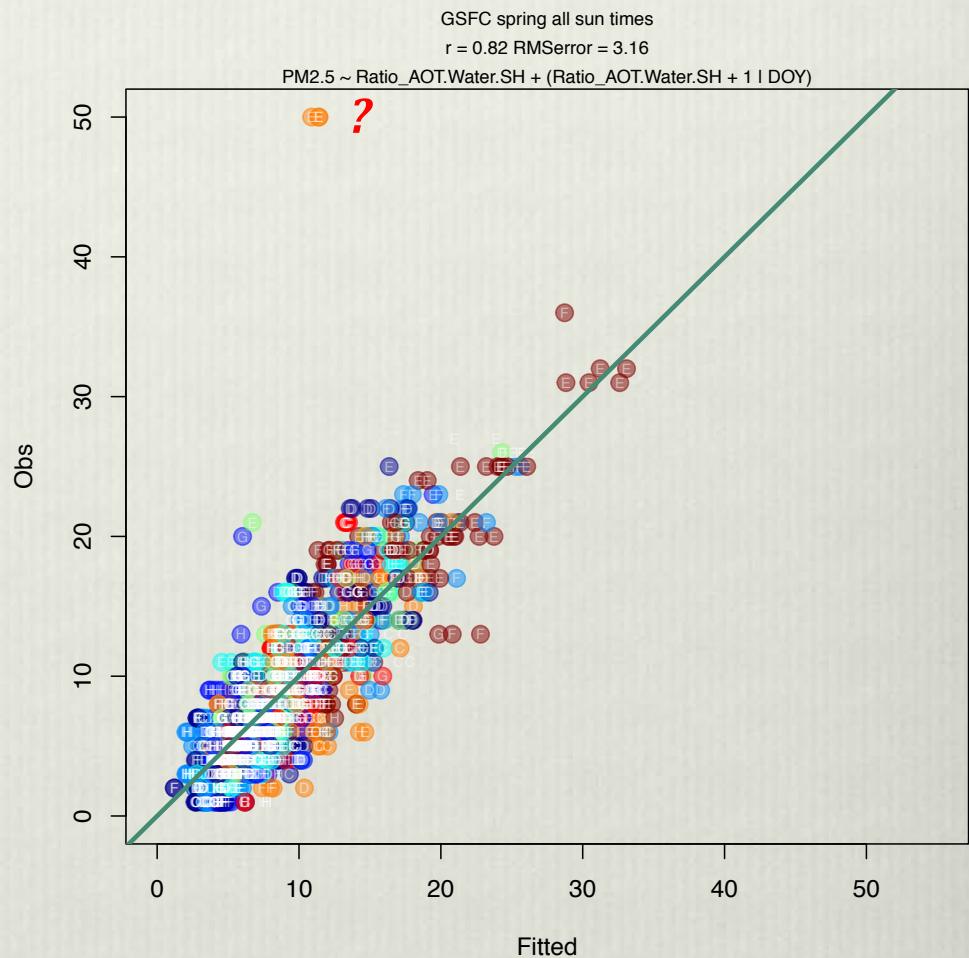
**TEMPO CWV** will aid in these regions;

For *these* regions, there is a **correlated history** of water input and then particle (precursor) input, with correlated mixing patterns.

**Not always true:** in those cases, use whole meteorological PBL cycle carefully.

## **GSFC / Beltsville, MD, 2007-2016**

also provides good relationships, although we know not all aerosol layers mix to the surface. (Need to understand theory.)



(Colors of points help allow us to identify the year and date of observation.)  
Changes during one day are not obvious.

# Sample of Current Research Around the World

	R <sup>2</sup>	% Precision		R <sup>2</sup>	%		R <sup>2</sup>	% Precision
Van Donkelaar et al. (2015)	0.69	15%	Poggia et al. (2016)	0.72		Ma et al. (2016)	0.7	28%
Kloog et al. (2014)	0.88	21%	et al. (2015)	0.72		Li et al. (2015)	0.8	20%
Chudnovsky et al. (2012)	0.92	21%	Jamer et al. (2015)	0.45		Sai Suman et al. (2014)	0.77	NA
Lee et al. (2011)	0.92	20%						

