

# Use of VIIRS AOT in Hierarchical Autoregressive Model to Predict Daily PM<sub>2.5</sub>

Jim Szykman<sup>1</sup>

Joint work with Erin Schliep<sup>2</sup>, Alan Gelfand<sup>2</sup>, David Holland<sup>1</sup>

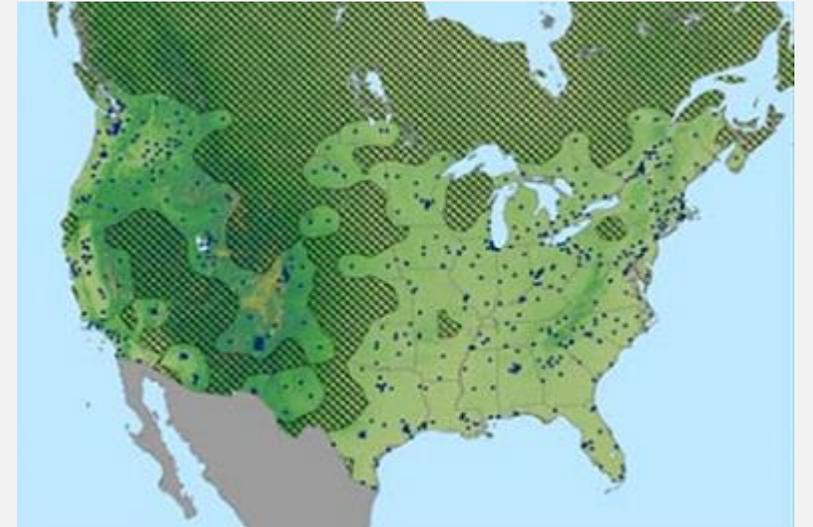
<sup>1</sup>National Exposure Research Laboratory  
U.S. EPA, Office of Research and Development, RTP, NC 27711

<sup>2</sup>Duke University, Durham, NC 27708

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# Motivation

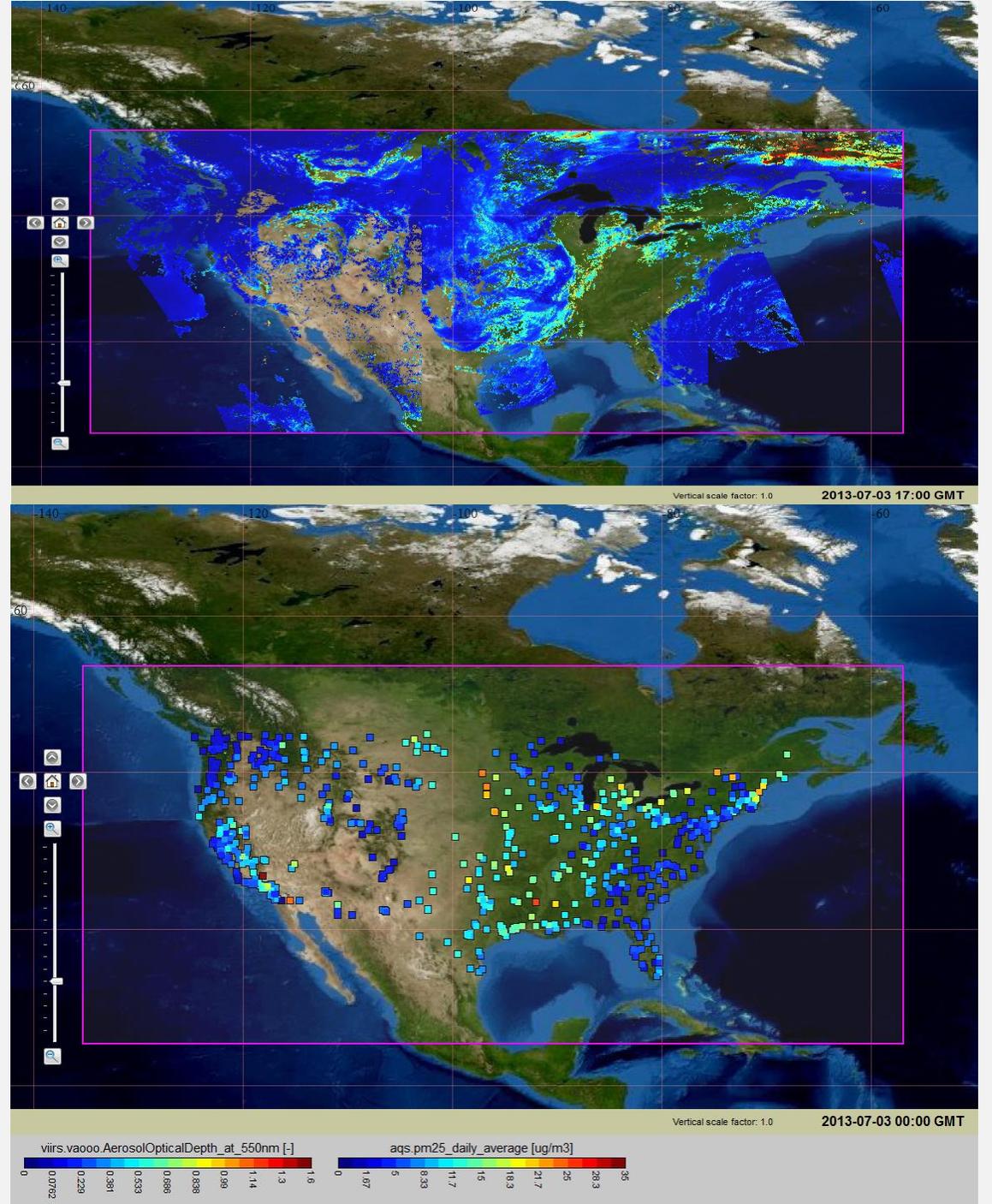
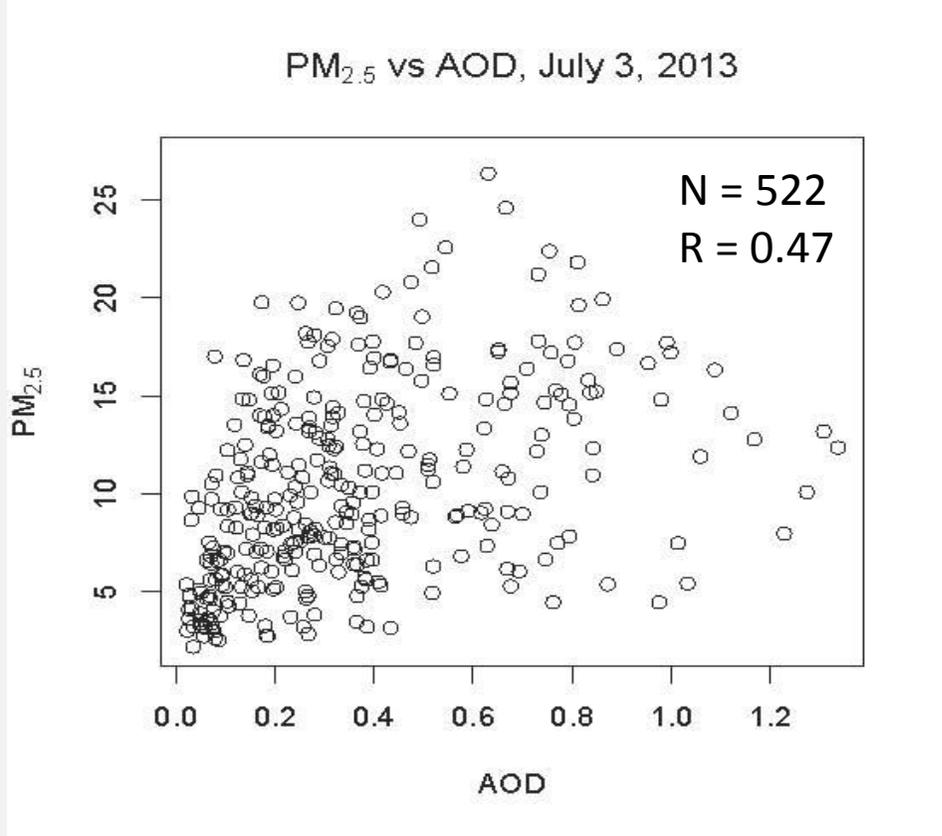
- Spatial and temporal coverage of existing PM<sub>2.5</sub> monitoring - significant data gaps resulting in over 36 million Americans (~40% of the area) not covered by a monitoring network
- Demand for accurate air quality characterization in community surveillance/human health analyses
- Chemical Transport Models require extensive emission inventories for model predictions – often do not capture high PM<sub>2.5</sub> concentrations associated with wildfires
- Daily AOT is a measure of the true state of the atmosphere for aerosols



**Additional information needed for spatial prediction of PM<sub>2.5</sub> in the hatched areas**

# Challenges with AOT and surface PM<sub>2.5</sub> in fusion models

- Correlation between the two data sources varies both in time and in space
- Data sources are temporally and spatially misaligned
- Extensive missing data in both the monitoring data and satellite data
  - AOT observed at 64% of grid cells with monitoring stations
  - Daily observations rate for study period 45% - 83%



# Hierarchical Autoregressive Model

Model consecutive day average  $PM_{2.5}$  across CONUS using daily spatially-varying coefficients:

- VIIRS AOT data - day-specific spatially-varying intercept and coefficient
- Account for missingness in AOT data via model-based imputation at missing grid cells
- Autoregressive term based on previous day surface  $PM_{2.5}$  concentrations
- Meteorological covariates (daily avg. T and RH)

# Autoregressive Model

$$P_t(s) = \alpha_{0,t} + \beta_{0,t}(s) + (\alpha_{1,t} + \beta_{1,t}(s))A_{i,t} + X_t(s)\gamma + \rho P_{t-1}(s) + E_t(s) \quad (M1)$$

$\alpha_{0,t}$  and  $\alpha_{1,t}$  - global intercept and AOT coefficients for day t

$X_t(s)$  - vector of location and day specific meteorological covariates

$\gamma$  - vector of coefficients

$\beta_{0,t}(s)$  and  $\beta_{1,t}(s)$  - spatially varying intercept and AOT coefficients for day t

$E_t(s)$  - error

Schliep E. M., A. E. Gelfand, and D. M. Holland, *Autoregressive spatially-varying coefficient models for predicting daily PM<sub>2.5</sub> using VIIRS satellite AOT*, Adv. Stat. Clim. Meteorol. Oceanogr, submitted Aug 2015

# Model Comparison

Competing submodels nested within model

Global intercept:

$$P_t(s) = \alpha_{0,t} + \beta_{0,t}(s) + (\alpha_{1,t} + \beta_{1,t}(s))A_{i,t} + X_t(s)\gamma + \rho P_{t-1}(s) + E_t(s) \quad (S1)$$

Non-autoregressive:

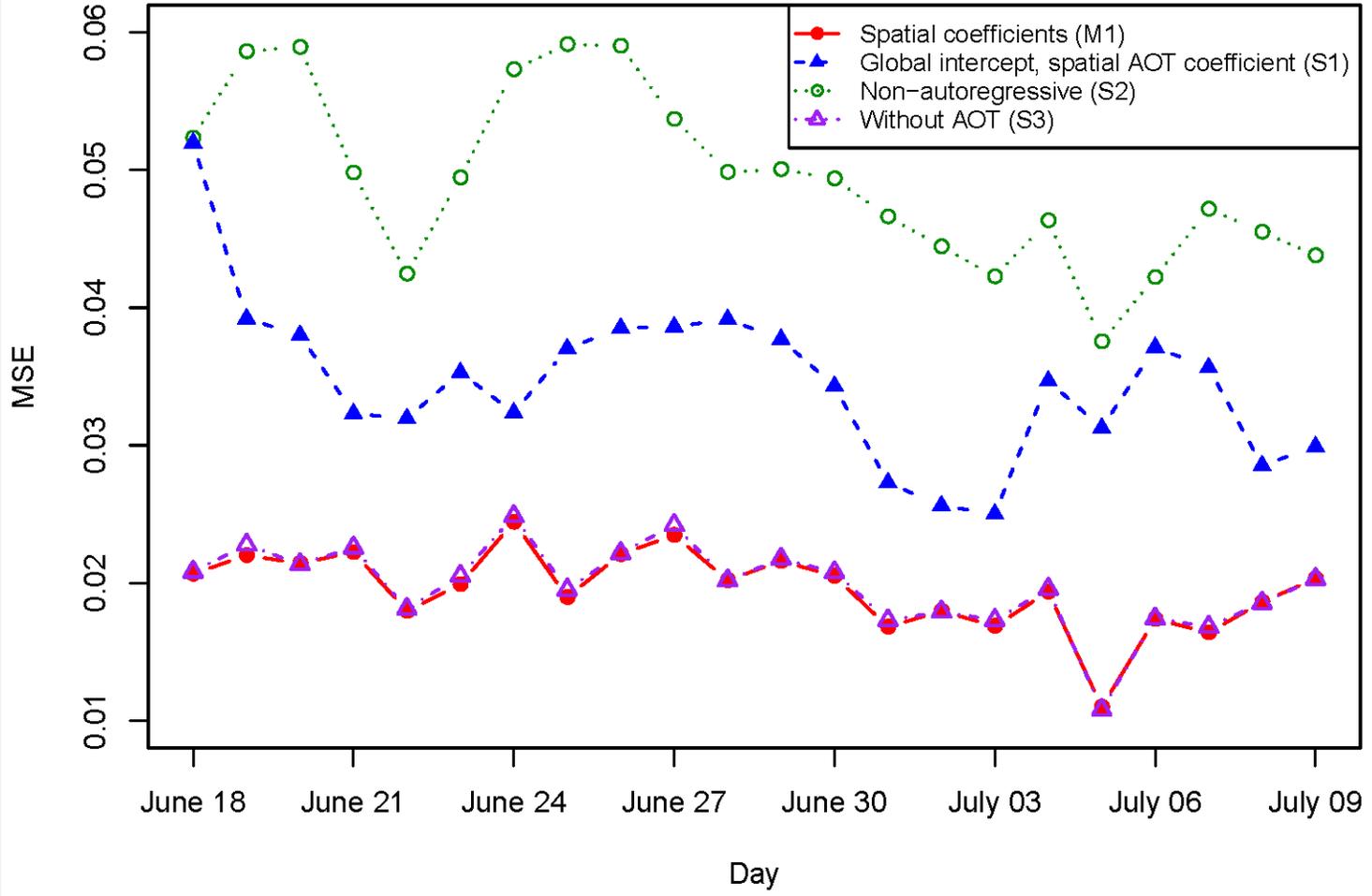
$$P_t(s) = \alpha_{0,t} + \beta_{0,t}(s) + (\alpha_{1,t} + \beta_{1,t}(s))A_{i,t} + X_t(s)\gamma + \rho P_{t-1}(s) + E_t(s) \quad (S2)$$

Without AOT:

$$P_t(s) = \alpha_{0,t} + \beta_{0,t}(s) + (\alpha_{1,t} + \beta_{1,t}(s))A_{i,t} + X_t(s)\gamma + \rho P_{t-1}(s) + E_t(s) \quad (S3)$$

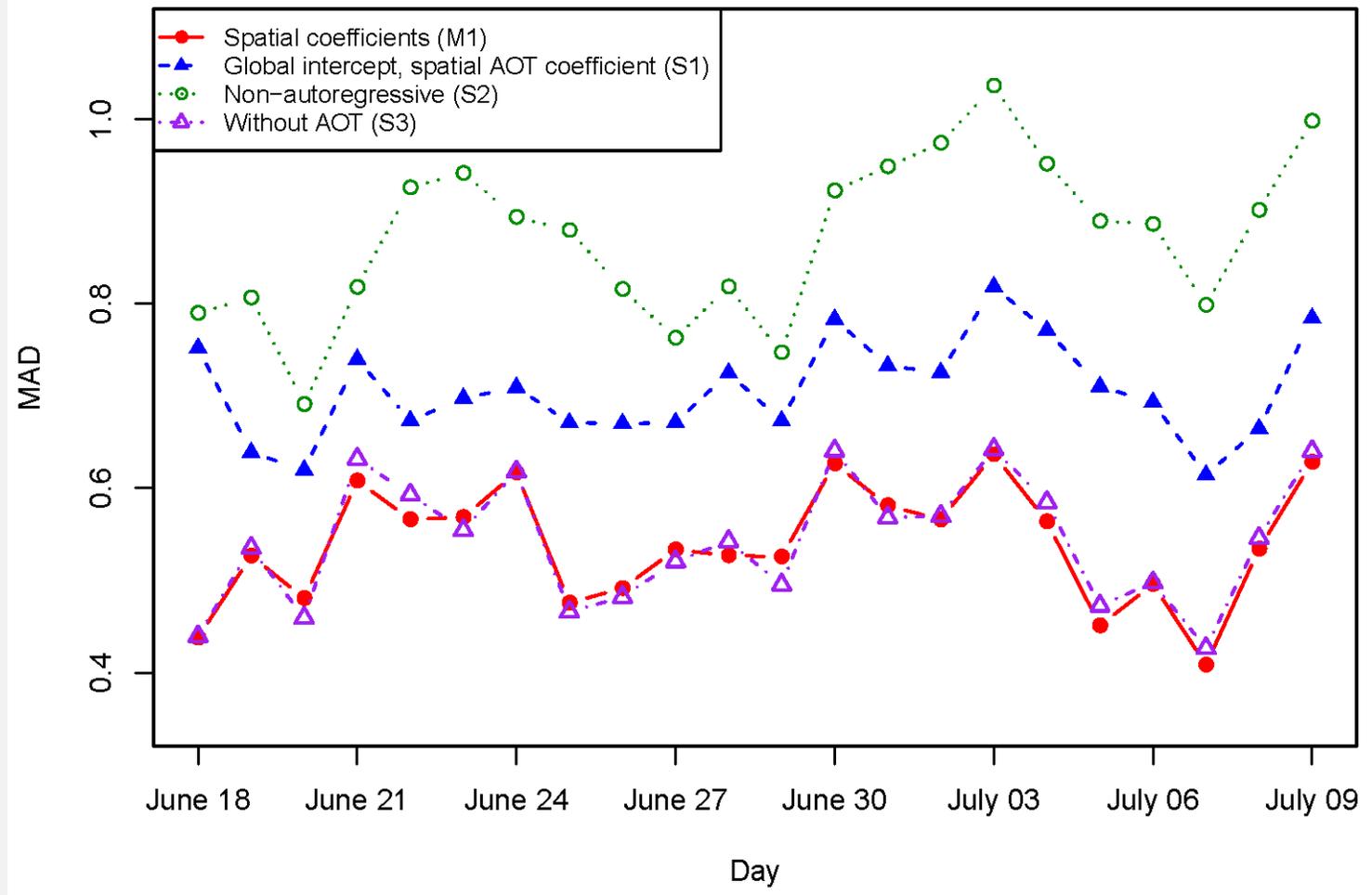
# Model Comparison

Daily MSE for the 510 in-sample locations



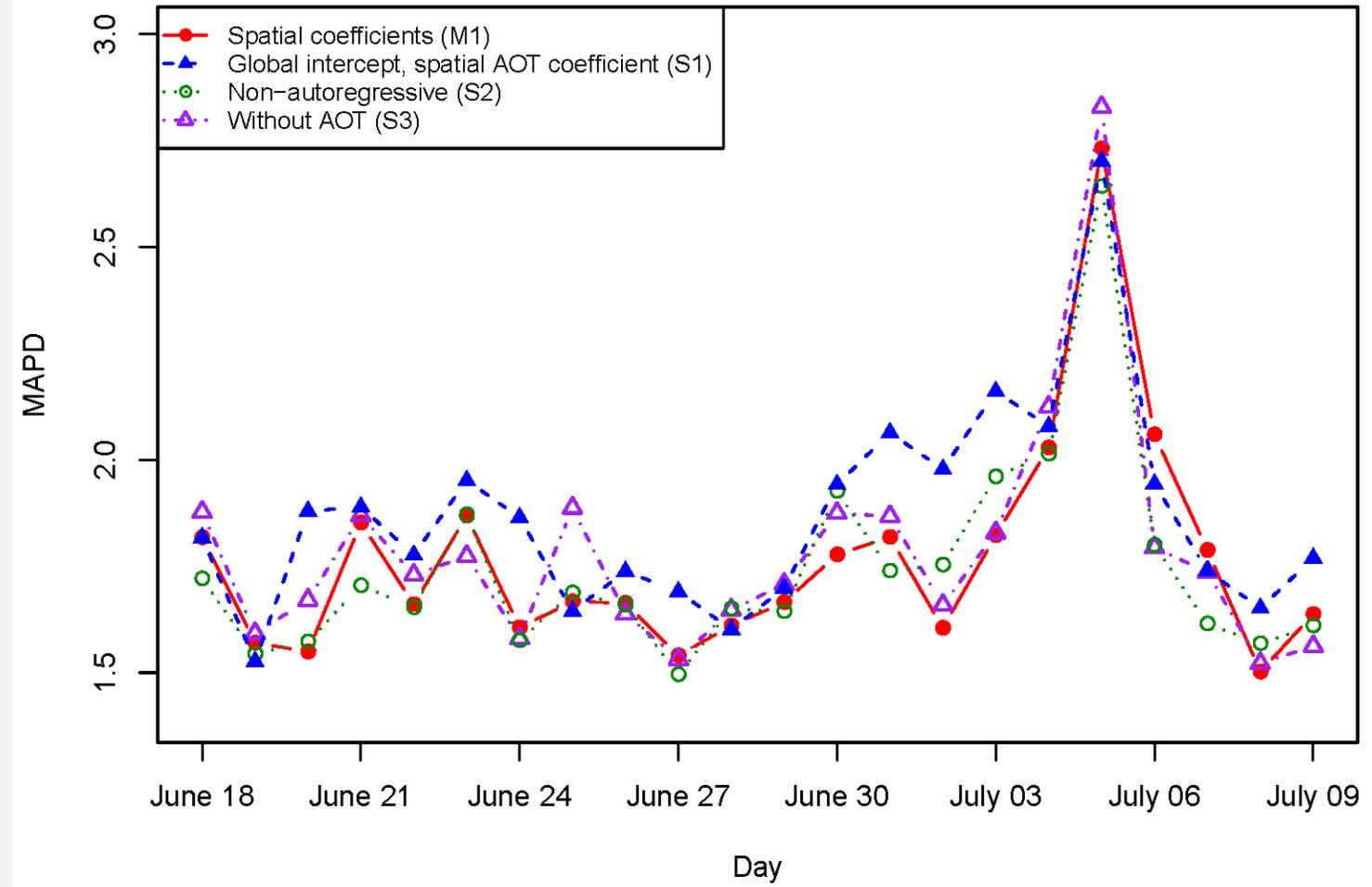
# Model Comparison

Daily MAD for the 510 in-sample locations



# Model Comparison

Daily MAPD for the 209 out-of-sample locations



# Summary and Conclusion

- Use of VIIRS AOT in hierarchical autoregressive model to model daily average  $PM_{2.5}$  concentration across CONUS
- Several submodels considered to quantify improvement in daily  $PM_{2.5}$  prediction using AOT
- Model comparison results show limited predictive capability with AOT, results consistent Paciorek and Liu (2009)
- Factors likely influencing use of AOT in model
  - Missing AOT data
  - Vertical structure of aerosols – need to develop improved scaling of AOT for aerosol aloft.



## EPA considering use of ceilometer (CL-51) as viable technology for PAMS mixing layer measurement

### Vaisala CL-51 Ceilometer Stated Characteristics:

- Cloud reporting range: 0...43,000 ft (0...13km)
- Backscatter profiling range: 0...49,200 ft (0...15km)
- Can operate in all weather
- Fast measurement - 6 second measurement cycle
- Reliable automatic operation
- Good data availability
- Eye safe diode laser (LIDAR)



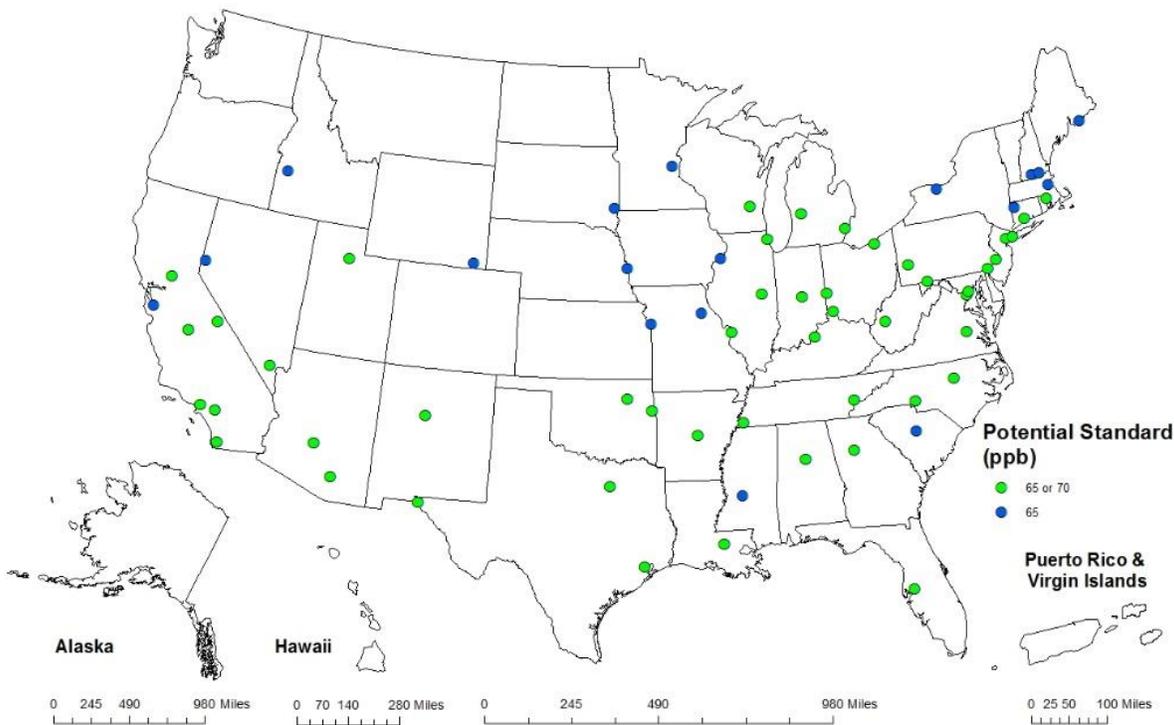
CL-51 positioned next to Space Science and Engineering Center, University of Wisconsin Mobile Lab



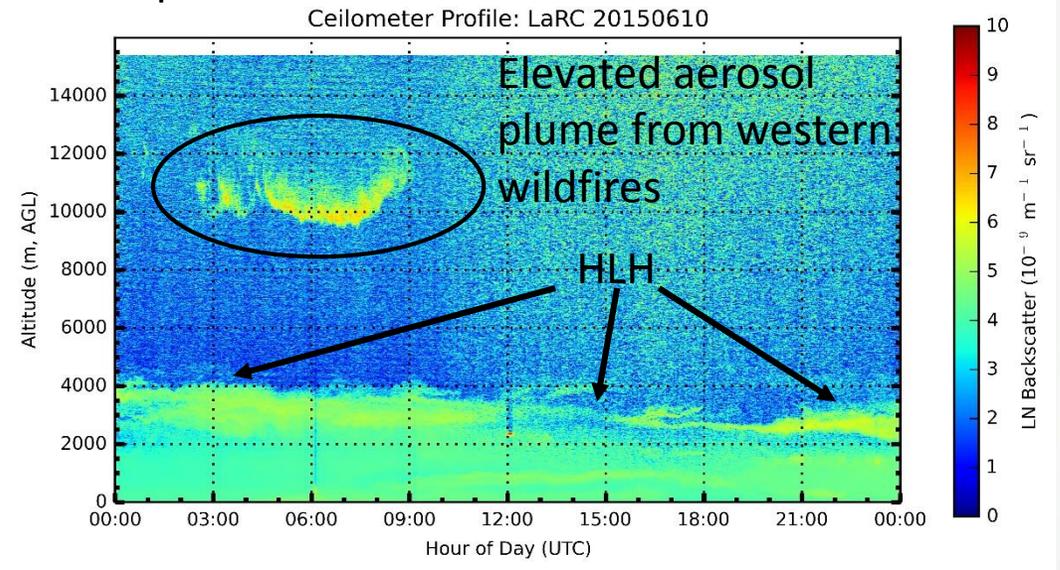


# EPA Photochemical Assessment Monitoring Station (PAMS) program

## Potential Site Locations for ceilometer (CL-51) Network



- Future CL-51 Network would allow for continuous aerosol profile measurement to define HLH on a regional basis
- CL-51 provides backscatter profile (~910 nm) up to 15.4 km



Map based on 2011-2013 ozone design values  
PAMS requirements will be based on 2014-2016 data

***Disclaimer: Although this work was reviewed by EPA and approved for presentation, it may not necessarily reflect official Agency policy. Mention of products or trade names does not indicate endorsement or recommendation for use by the Agency.***