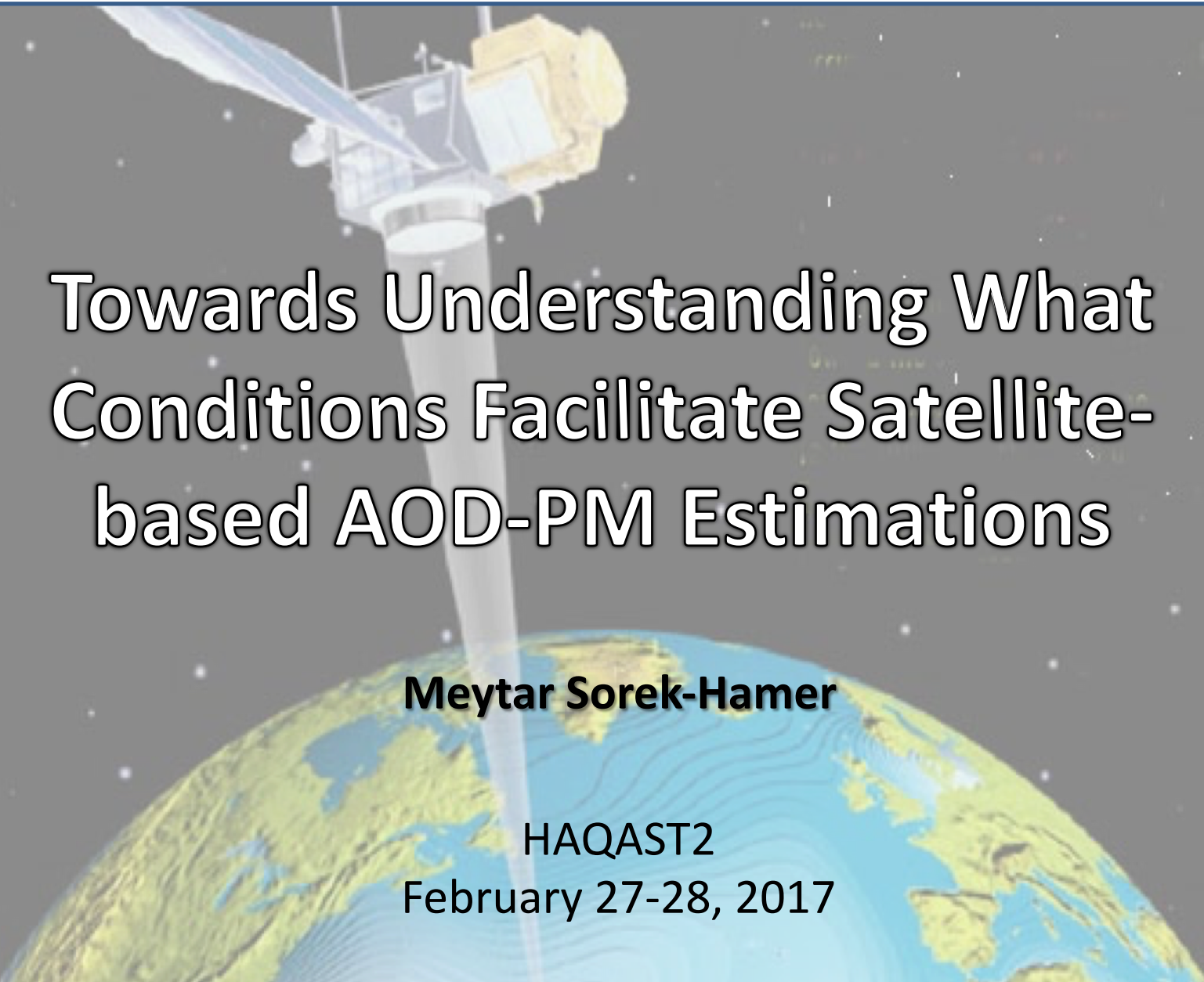
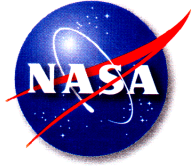


A satellite is shown in space, emitting a bright beam of light towards the Earth. The Earth is visible at the bottom of the frame, showing blue oceans and green landmasses. The background is a dark grey space filled with small white stars.

Towards Understanding What Conditions Facilitate Satellite- based AOD-PM Estimations

Meytar Sorek-Hamer

HAQAST2

February 27-28, 2017

Study area - Israel



ISRAEL and...

The United States



*Maps drawn to same scale.
Israel shown in blue.*

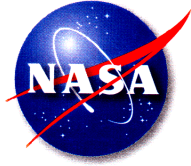
ISRAEL and...

California



*Maps drawn to same scale.
Israel shown in blue.*





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Ben Gurion University, Israel

Outline

- Background and summary of published work
- The contribution of AOD to PM prediction models
- Future work

This is the first study, to our knowledge, that quantitatively presents the AOD spatiotemporal contribution to the MEM which has been found to be < 2% in different regions.

Why estimate air pollution?

What do satellites tell us?

- Fine particulate matter ($PM_{2.5}$) has a large and growing literature of associations with adverse health effects
- Measurements of $PM_{2.5}$ for epidemiological studies are often based on assigning exposures based on a sparse network of ground monitors
- Satellite measures, collected daily with fine resolution, can help reconstruct the spatial and temporal exposure patterns in $PM_{2.5}$

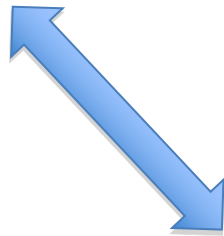
The Highest Quality of Data Rests on a Tripod



Satellite Data

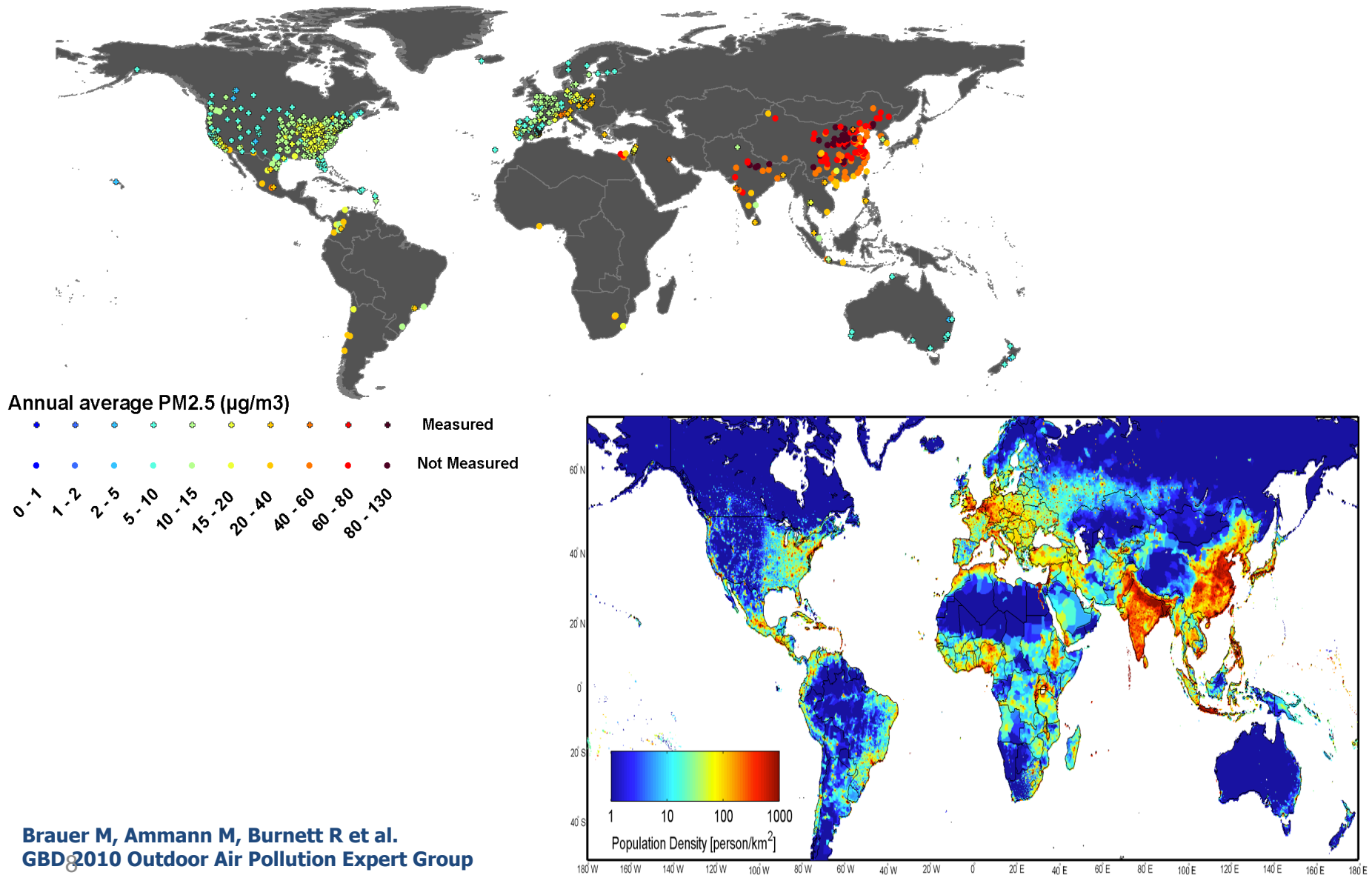


Ground Measurements
and In-Situ Data



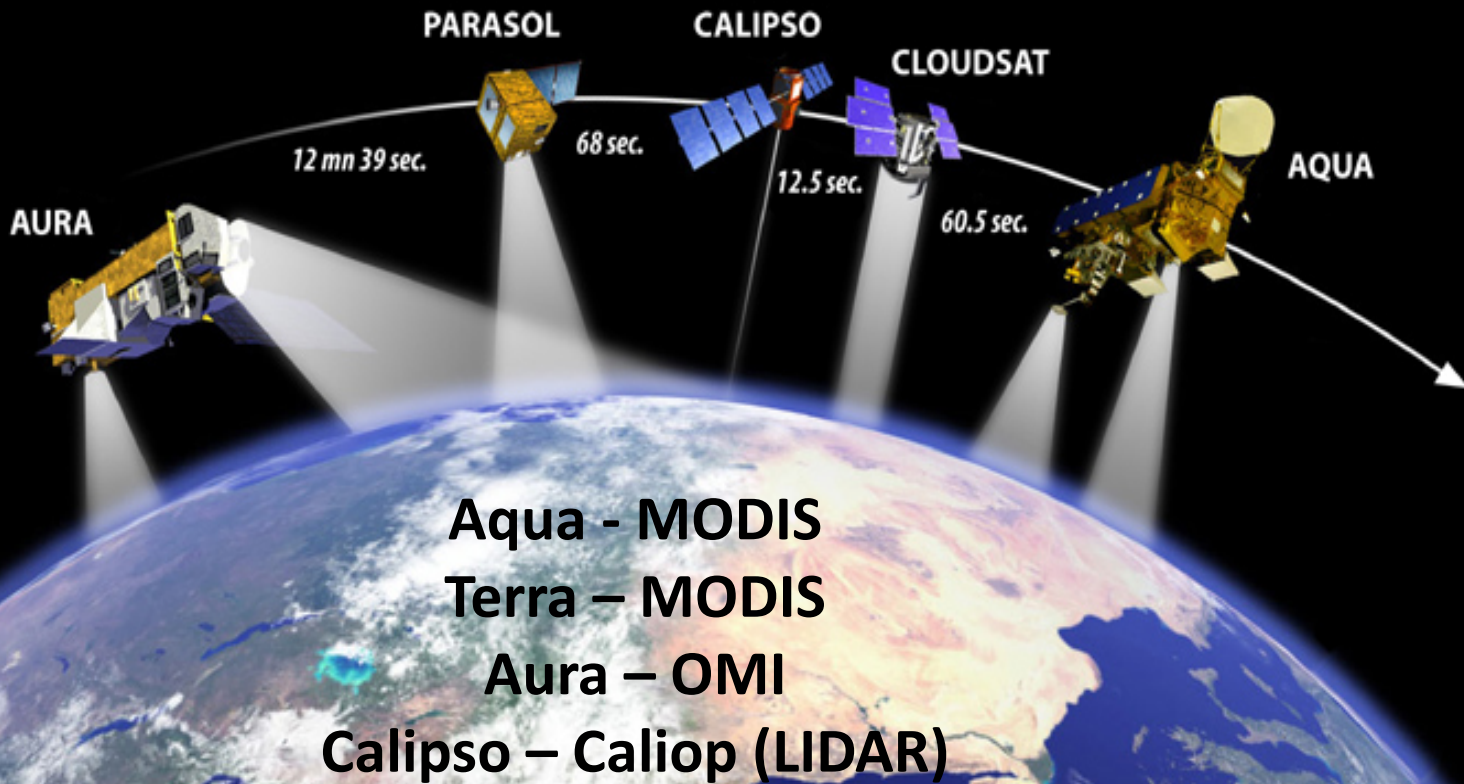
Models

Global Status of PM2.5 Monitoring Networks



Air Quality Satellite Platforms

Most satellites used for Air Quality (AQ) remote sensing (in this study) are Sun synchronous, polar orbiting satellites with 1-2 observations per day.



Research around the world

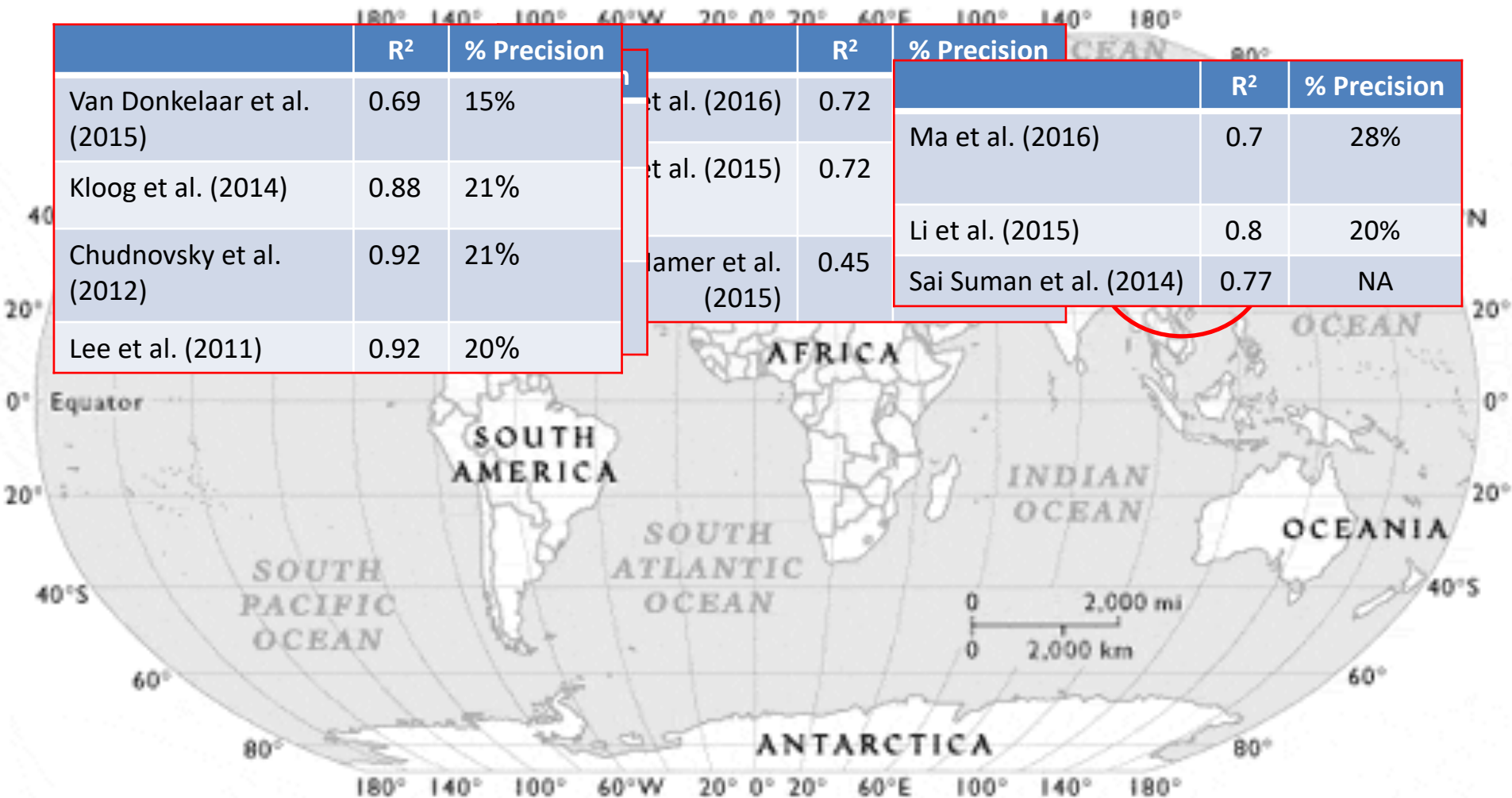
Author	Sensor	Date	Region	Number of Ground Monitors	PM _{2.5} /PM ₁₀	Linear Regression	R
Wang ¹⁵⁴	MODIS (Terra)	2002	Alabama	7	PM _{2.5} (24 hr) ^a	77.0τ - 0.23	0.67
	MODIS (Aqua)	2002	Alabama	7	PM _{2.5} (24 hr) ^a	68.6τ + 1.93	0.76
	Average	2002	Alabama	7	PM _{2.5} (24 hr) ^a	72.3τ + 0.85	0.98
Chu ¹⁵³	MODIS	August–October 2000	Italy	1	PM ₁₀	54.7τ + 8.0	0.82
Engel-Cox ¹⁶¹	MODIS	April–September 2002	United States	1338	PM _{2.5}	22.6τ + 6.4	0.4
					PM _{2.5} (24 hr)	18.7τ + 7.5	0.43
Liu ²⁰⁸	MISR	2003	St. Louis	22	PM _{2.5}	NA	0.8
Engel-Cox ¹⁶³	MODIS	July 1 to August 30, 2004	Baltimore	4	PM _{2.5}	31.1τ + 5.2	0.65
					PM _{2.5} (<PBL)	48.5τ + 6.2	0.65
					PM _{2.5} (24 hr)	25.3τ + 11.1	0.57
					PM _{2.5} (24 hr < PBL)	64.8τ + 1.76	0.76
Liu ¹⁶⁹	MISR	2001	Eastern United States	346	PM _{2.5}		–
Al-Saadi ¹⁶⁴	MODIS	Review	United States		PM _{2.5}	62.0τ	NA
Gupta ¹⁷¹	MODIS	2002 and July–November 2003	Global cities	26	PM ₁₀ ^a	141.0 τ	0.96
Koelemeijer ¹⁵²	MODIS	2003	Europe	88 (PM _{2.5})	PM _{2.5} ^a	NA	0.63
					PM ₁₀ ^a	214.0τ - 42.3	0.58
Kacenelenbogen ¹¹⁸	POLDER	April–October 2003	France	28	PM _{2.5}	26.6τ + 13.2	0.7
Gupta ¹⁷³	MODIS	February 2000 to December 2005	Southeastern United States	38	PM _{2.5}	29.4τ + 8.8	0.62
					PM _{2.5} (24 hr)	27.5τ + 15.8	0.52
Hutchison ¹⁵⁸	MODIS	August–November 2003 and 2004	Texas	28	PM _{2.5} (August) ^a	68.8τ - 39.9	0.47
					PM _{2.5} (September) ^a	59.7τ - 17.2	0.98
Paciorek ¹⁷⁷	GOES-12	2004	United States	Not given	PM _{2.5} (24 hr)	NA	0.5
					PM _{2.5} (yearly)	NA	0.75
An ¹⁷⁹	MODIS	April 3–7, 2005	Beijing	6	PM ₁₀ ^a	21.7τ + 6.1	0.92
					PM _{2.5} ^a	31.1τ + 5.1	0.92
Schaap ¹⁸⁰	MODIS	August 2006 to May 2007	Cabauw, Netherlands	1	PM _{2.5}	120τ + 5.1	0.72

(Hoff & Christopher. AWMA. 2009)

Sample of Current Research Around the World

	R ²	% Precision		R ²	% Precision
Van Donkelaar et al. (2015)	0.69	15%	et al. (2016)	0.72	
Kloog et al. (2014)	0.88	21%	et al. (2015)	0.72	
Chudnovsky et al. (2012)	0.92	21%	amer et al. (2015)	0.45	
Lee et al. (2011)	0.92	20%			

	R ²	% Precision
Ma et al. (2016)	0.7	28%
Li et al. (2015)	0.8	20%
Sai Suman et al. (2014)	0.77	NA



AOD contribution to MEM

- (1) $PM_{ij} \sim \alpha + (\alpha_j) + \varepsilon_{ij}$ (Base Model)
- (2) $PM_{ij} \sim \alpha + (\alpha_j + \beta AOD_{ij}) + \varepsilon_{ij}$
- (3) $PM_{ij} \sim \alpha + \beta AOD_{ij} + (\alpha_j) + \varepsilon_{ij}$



Training subset

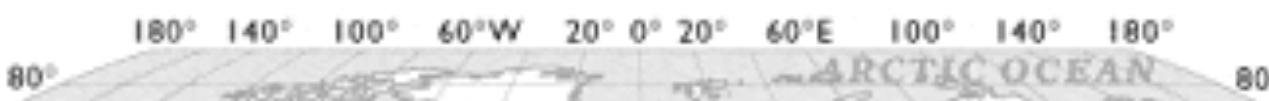
$$\text{AOD Contribution} = \frac{(\text{Deviance}_{(2) \text{ or } (3)} - \text{Deviance}_{\text{base model}}) * 100}{\text{Deviance}_{\text{base model}}}$$



Testing subset

Model fit: R^2 , RMSE, % Precision

AOD contribution



Published			Current Study		
	R ²	% Precision	R ² MEM Base model (Linear reg.)	% Precision (mean PM [μg/m ³])	AOD Contr.
North East USA	0.88-0.92	20%-21%	0.71 (0.25)	38% (10.4)	2%
SJV, CA, USA	0.66-0.77	43%-48%	0.41 (0.13)	70% (19.4)	1.5%
Israel	0.45-0.72	37%-52%	0.47 (0.07)	46% (16.9)	0.35%
Italy	0.72	30%	0.42 (0.01)	52% (22.4)	0.64%
France	0.64*	NA	0.52 (0.1)	44% (108.8)	1.1%
India	0.77	NA	0.44 (0.38)	50% (20.4)	2.1%

“...none of these models has been tested in developing countries...It would be appropriate to start building a simpler model...cross validate...”

(Liu Y. 2013)

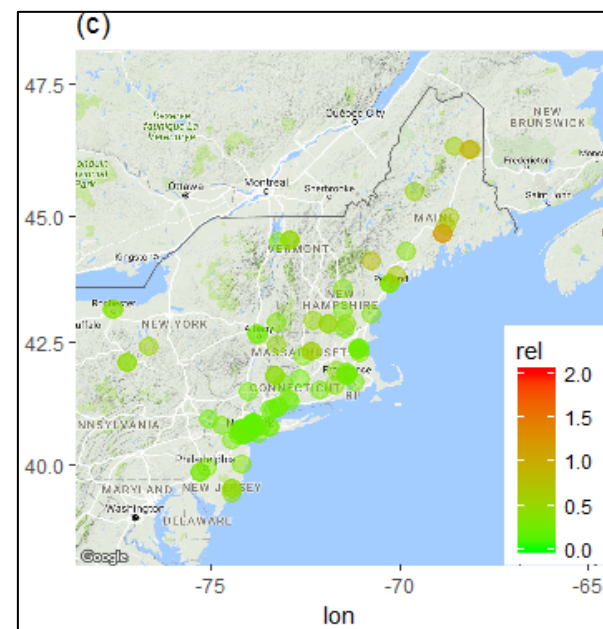
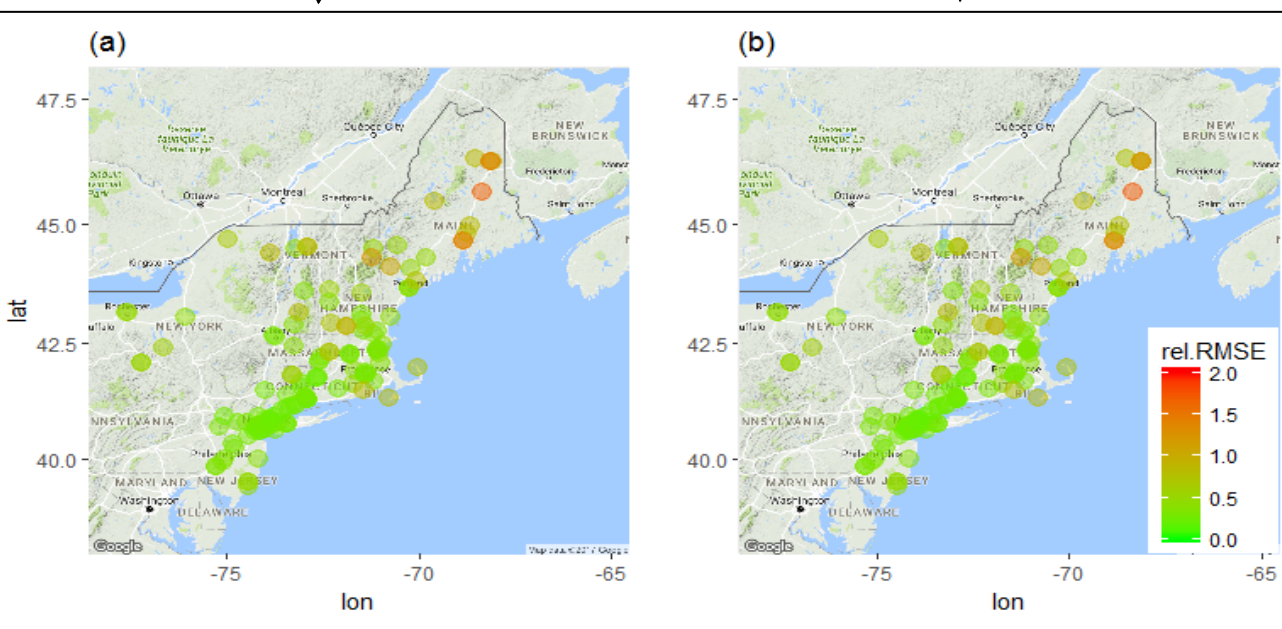
*In preparation

AOD Spatial Contribution

Date only



Date , AOD



Date +AOD+Met

Specific spatiotemporal conditions count

- Van Donkelaar 2011
- Chatfield 2017* and Chu 2015
- Kim 2016

Biomass burning event
Discover-AQ results
Localized empirical model

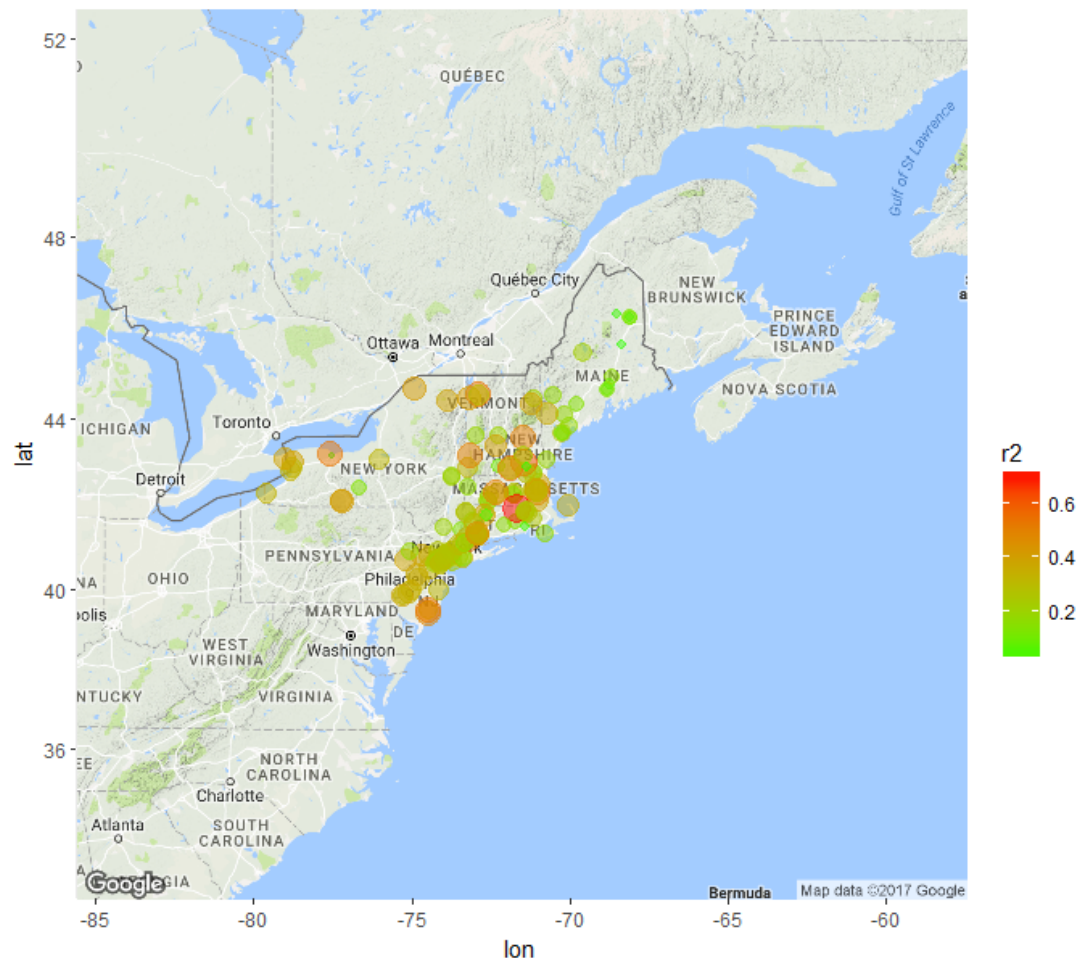
What are the Environmental-Physical conditions that facilitate a certain accuracy level of AOD-PM association ?

Linear regression – $PM \sim AOD$

NE USA, 2000-2013.



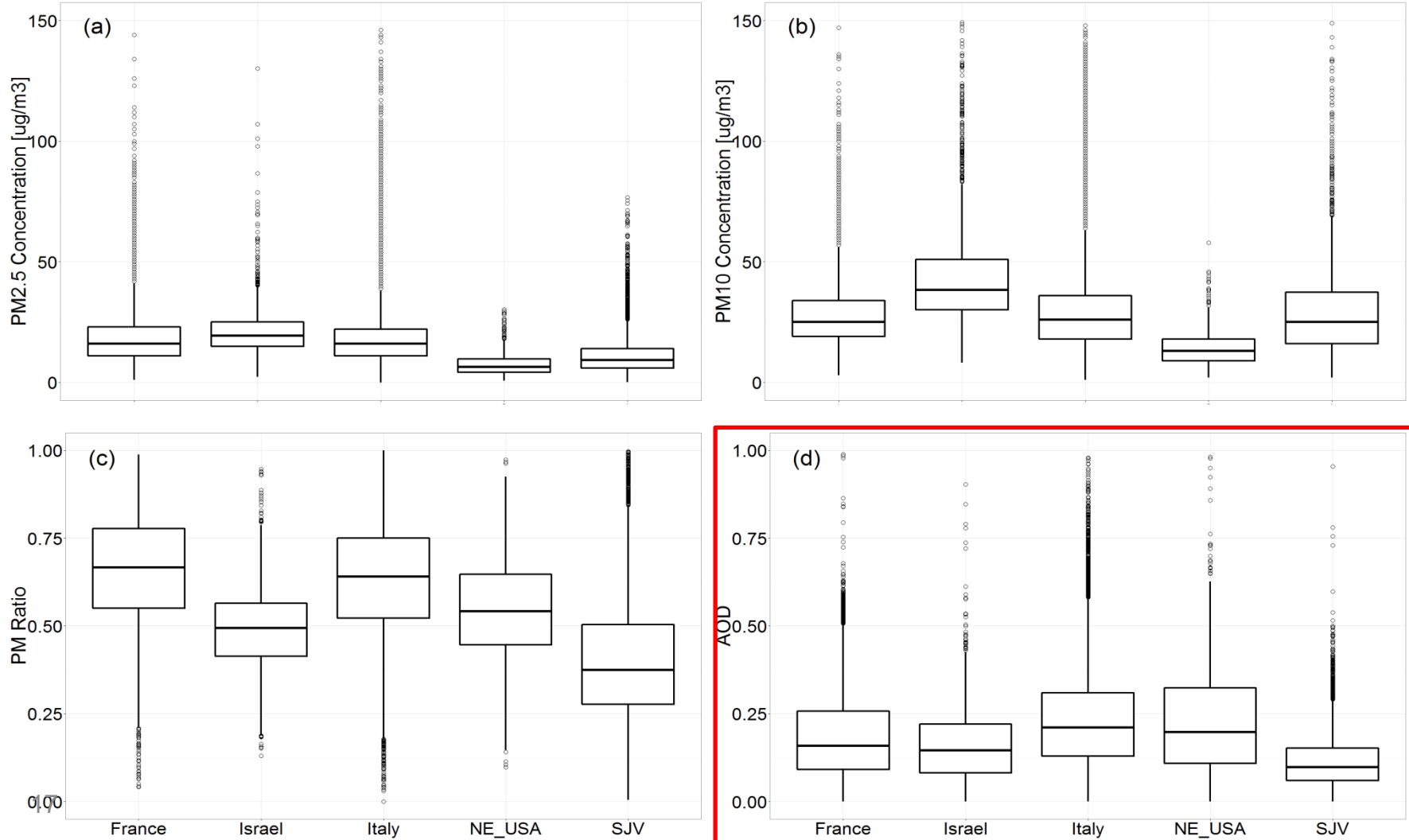
Only Stations with $R^2 > 0.5$



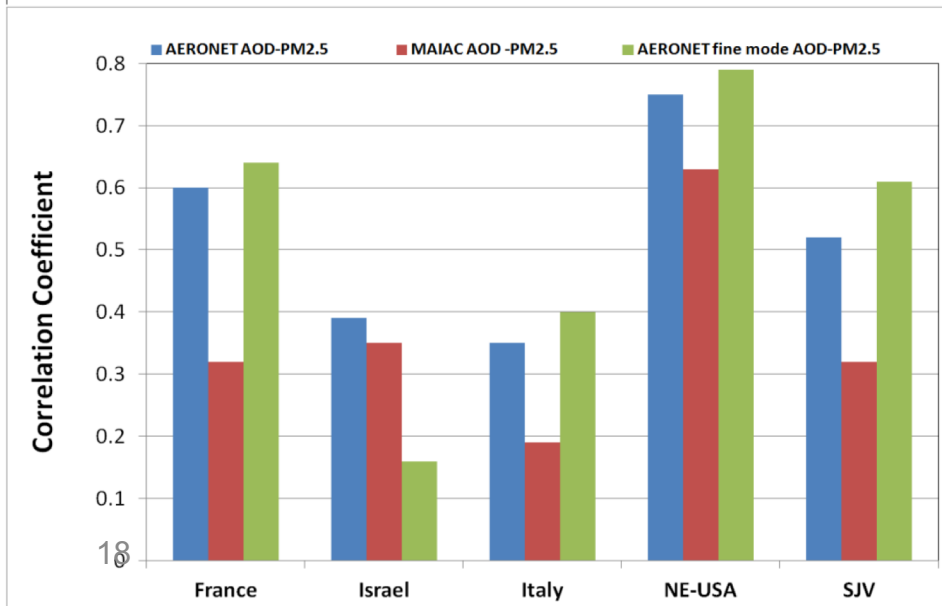
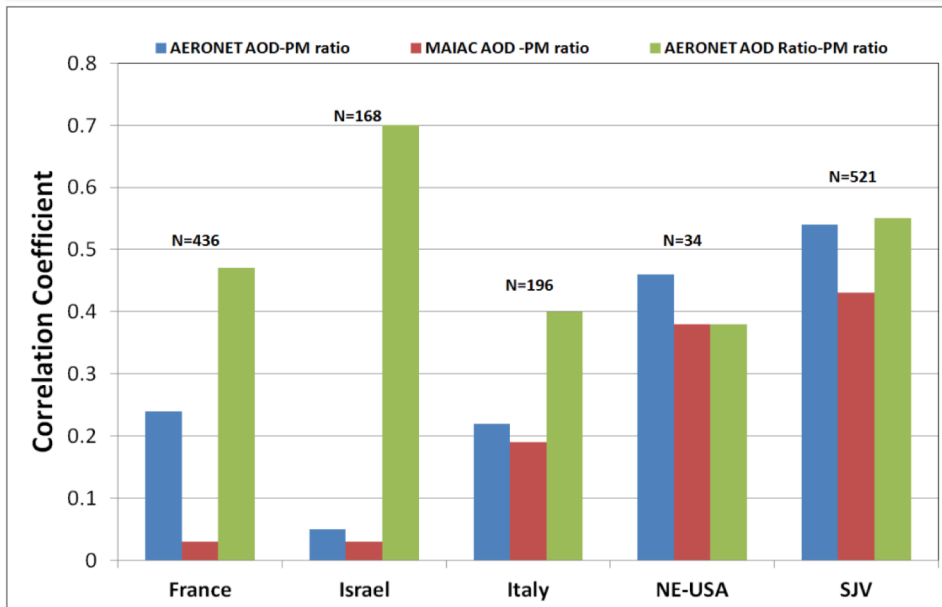
All stations

AOD sensitivity to PM Ratio

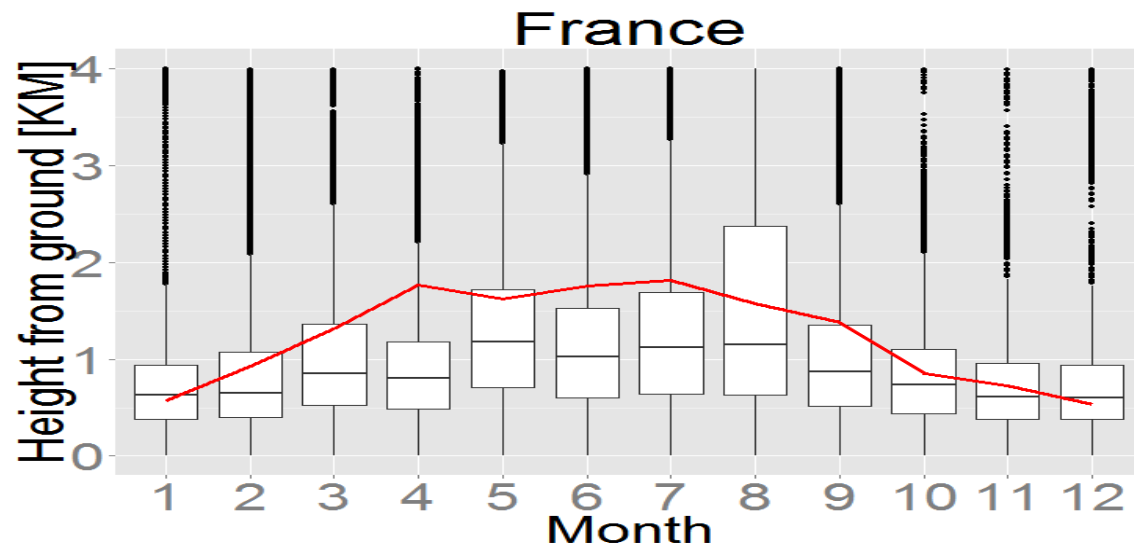
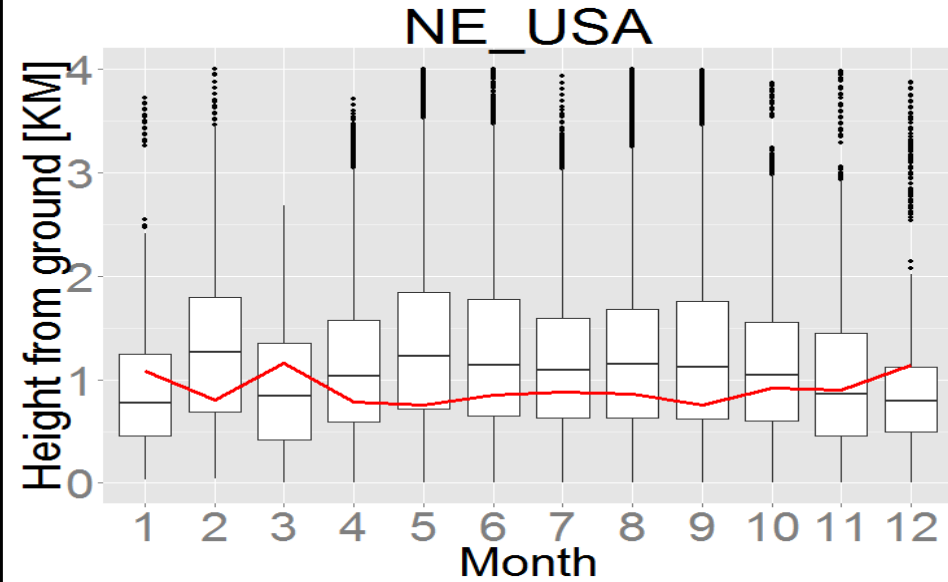
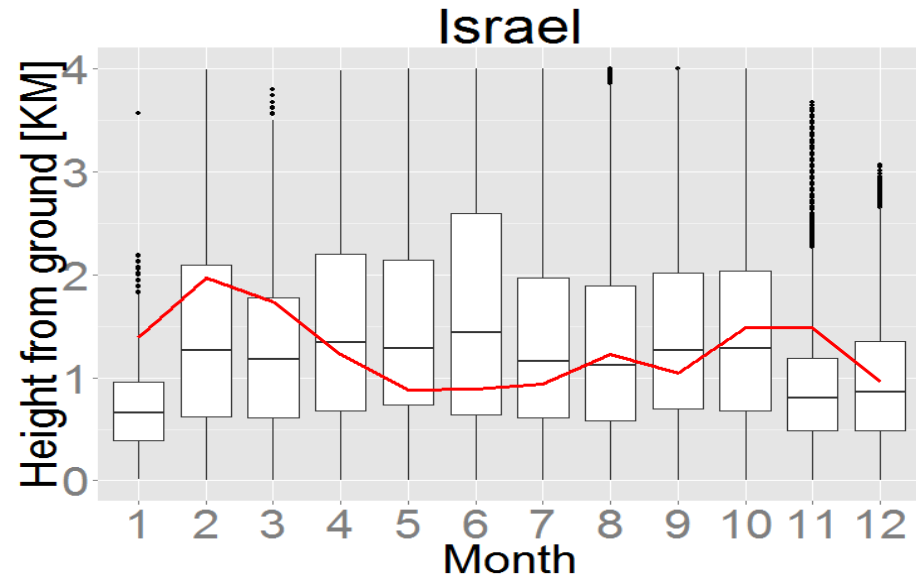
$$\frac{PM_{2.5}}{PM_{10}} = \frac{PM_{2.5}}{(PM_{2.5} + PM_{2.5-10})}$$



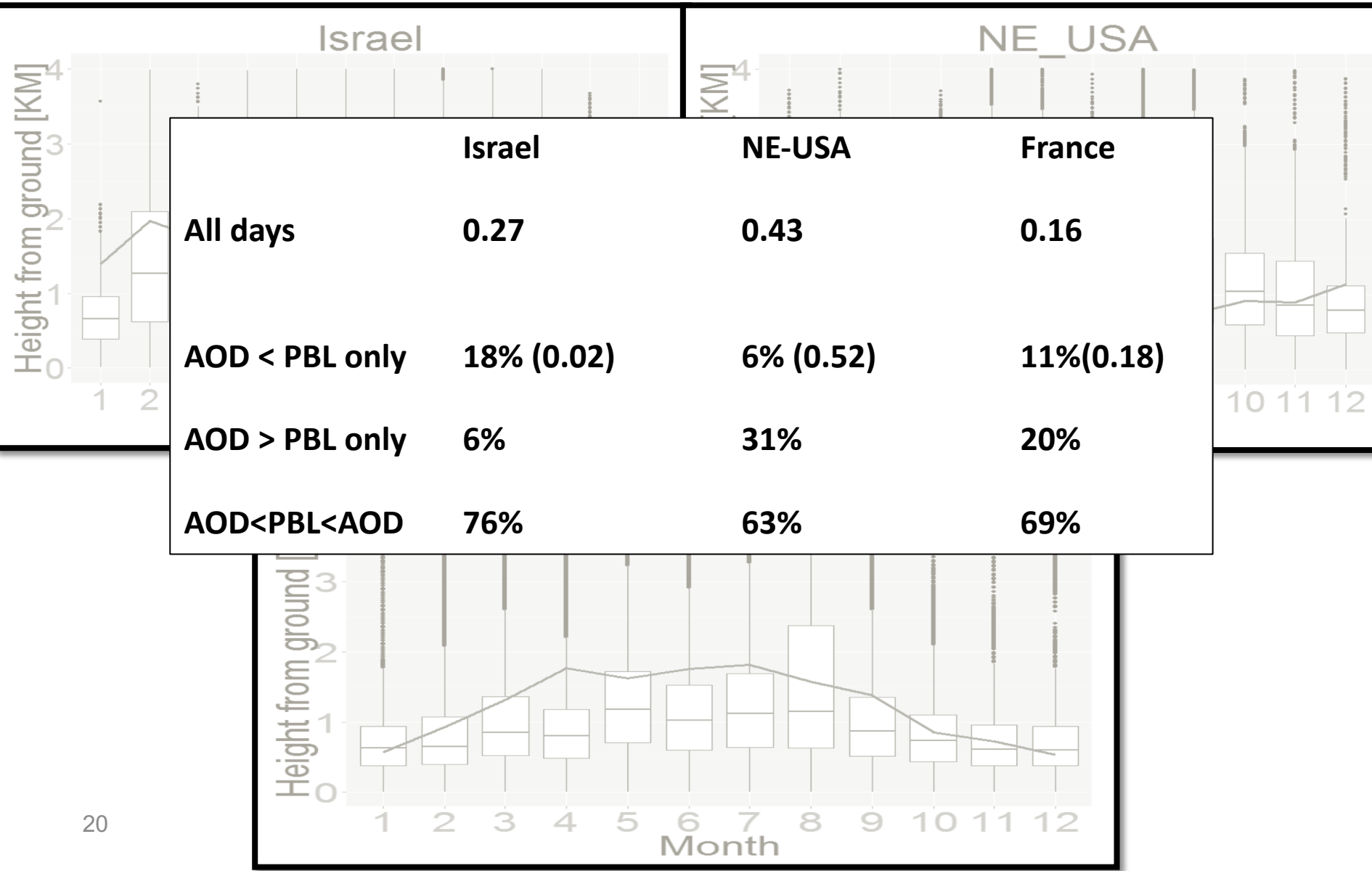
AOD sensitivity to PM ratio



Aerosol Vertical Profile



AOD-PM Fit by Vertical Subset Location



Conclusions

- ✓ The possible advantages of using satellite data are known.
- ✓ This is the first study, to our knowledge, that quantitatively presents the AOD spatiotemporal contribution in different locations which has been found to be $< 2\%$.
- ✓ The underlying assumptions and conditions when using AOD to estimate PM concentrations should be further examined on a local spatial and temporal basis.
- ✓ Accurate understanding of these conditions will enable better utilization of satellite data to improve our knowledge on air quality, to develop localized AOD/models and expand exposure matrices for Env. health studies.
- ✓ **Next step:** Classification of daily conditions on a worldwide database (SOM)

Backup Slides

Table 3. Correlations between collocated AOD and PM_{2.5} concentrations in ground stations near the CALIPSO path.

		Israel		NE_USA		France	
		R	N(no. of stations)	R	N(no. of stations)	R	N(no. of stations)
Available days with obs. From MAIAC, and PM	Multianua 1	0.27	58,131 (41)	0.43	49,202 (144)	0.16	10231 (195)
	Winter	0.29	9,214	0.53	13,302	0.26	2,401
	Spring	0.37	9,529	0.24	8,332	0.32	2,414
	Fall	0.17	15,198	0.15	11,856	0.05	2,236
	Summer	0.18	24,190	0.46	15,712	0.25	3,180
Days which CALIPSO detected aerosols within PBL	Multianua 1	-0.02	788 (5)	0.52	306 (95)	0.18	1,707 (34)
	Winter	-0.11	120	0.46	16	0.34	383
	Spring	-0.25	117	0.52	104	0.38	428
	Fall	0.1	175	-0.01	177	0.04	387
	Summer	-0.002	376	0.94	9	0.14	509