

# **AeroSAT posters**



- AROLA
- BAI
- CHE
- FARLIE
- FAN
- GARAY
- GINOUX
- GUANG
- HSU
- HU



- : JIE
- KAZUMA
- KOLLACHI
- LACAGNINA
- LIPPONEN
- LI
- LIU
- LU
- MAHMOOD
- MATTOO



- ⋮ MEI
- NOROOZI
- SHAHZAD
- SHUKLA
- THOMAS
- TSAY
- VANDEN-BUSSCHE
- VANDEN-BUSSCHE
- XIE

**AROLA**

Huttunen, J., Kokkola, H., Mielonen, T., Mononen, M. E. J., Lipponen, A., Reunanen, J., Lindfors, A. V., Mikkonen, S., Lehtinen, K. E. J., Kouremeti, N., Bais, A., Niska, H., and [Arola, A.:](#)

## Retrieval of aerosol optical depth from surface solar radiation measurements using machine learning algorithms, non-linear regression and a radiative transfer-based look-up table,

Atmos. Chem. Phys., 16, 8181-8191, doi:10.5194/acp-16-8181-2016, 2016.



**SSR = Function of SZA, AOD, SSA, WVC**

SSR = Surface Solar Radiation

SZA = Solar Zenith Angle

AOD = Aerosol Optical Depth

SSA = Single Scattering Albedo

WVC = Water Vapor Content

Measured or known / input,  
Retrieved / output

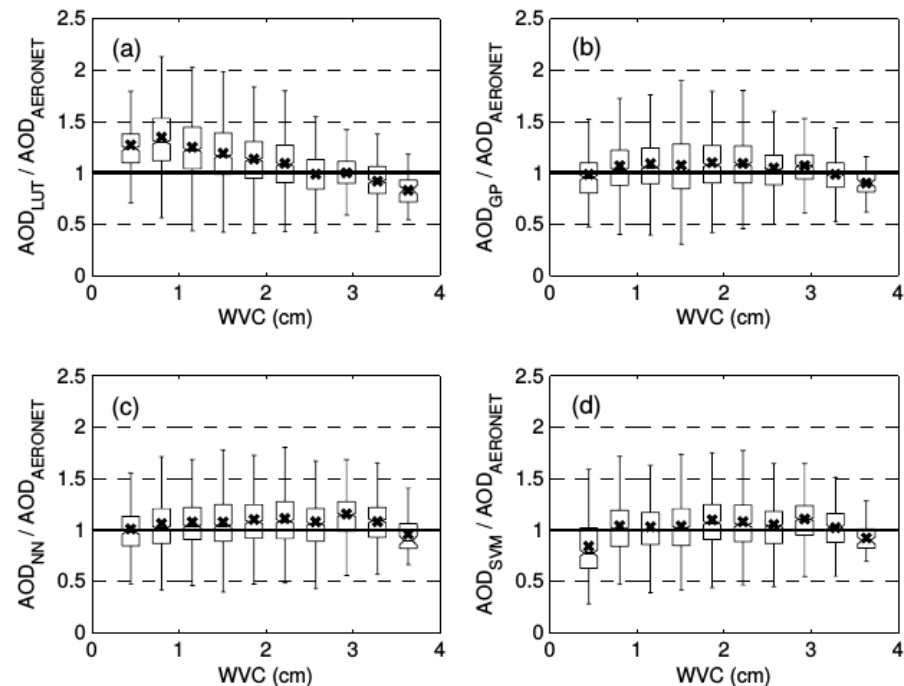
### Following methods:

- 1) a look-up table method based on radiative transfer modelling
- 2) a non-linear regression method
- 3) Gaussian Process (GP)
- 4) Neural Network (NN)
- 5) Random Forest (RF)
- 6) Support Vector Machine (SVM)

Compared with AOD from AERONET site in Thessaloniki, Greece.

Look-up-Table

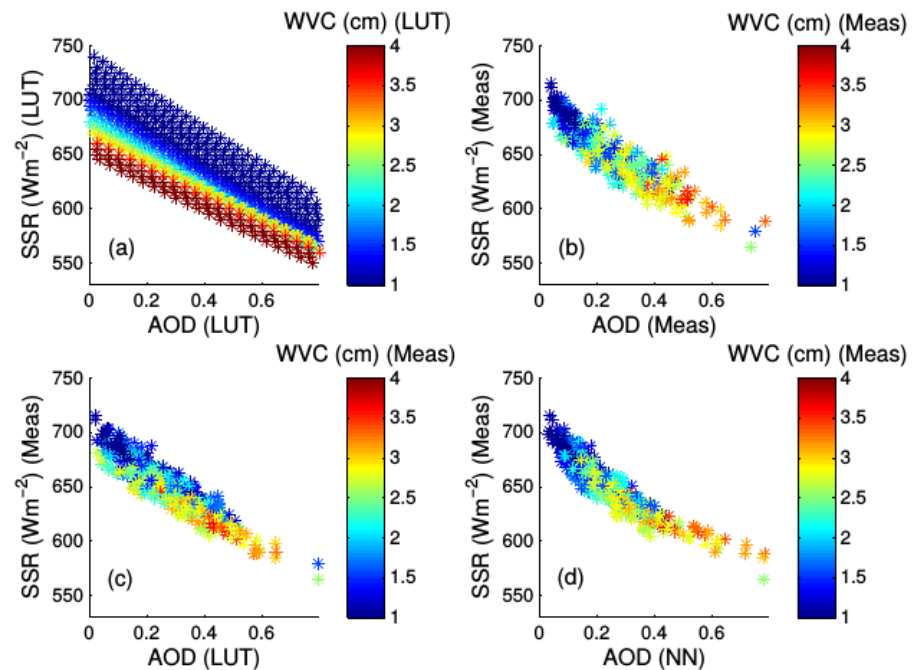
Gaussian Process



**Figure 4.** The same as Fig. 3, but the ratio of predicted to measured AOD is given as a function of the water vapour content (WVC).

Neural Network

Support Vector Machine



**Figure 5.** Solar surface radiation (SSR), aerosol optical depth (AOD) and water vapour content (WVC) for a fixed solar zenith angle ( $48.50\text{--}51.50^\circ$ ) for (a) look-up table (LUT) and (b) measurements (Meas). The predicted AODs for (c) LUT and (d) neural network (NN) are the same for SSR, WVC and SZA.

**BAI**

# Ground-level PM<sub>2.5</sub> concentrations derived from 3km resolution MODIS AOD over the Yangtze River Delta in China

- A mixed effects model (Lee *et al.* 2011) was employed to predict PM<sub>2.5</sub> concentrations without using meteorological data.
- An obvious increase in R<sup>2</sup> for the model was found compared to the linear regression (Fig.1.).

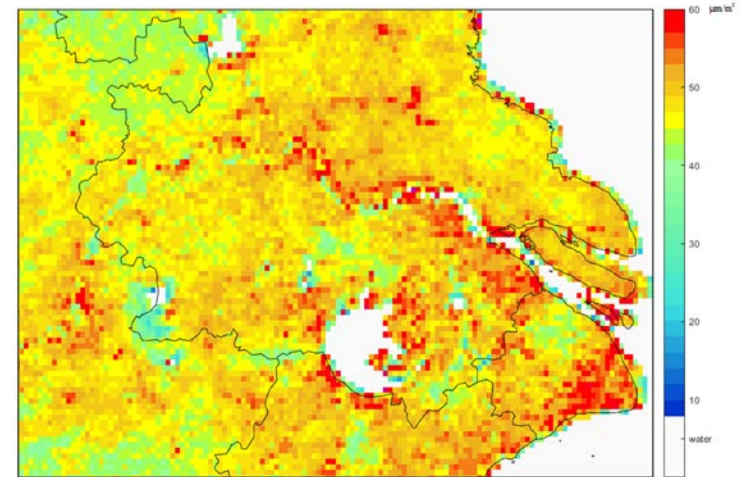


Fig.2. Mean PM<sub>2.5</sub> concentrations over the Yangtze River Delta.

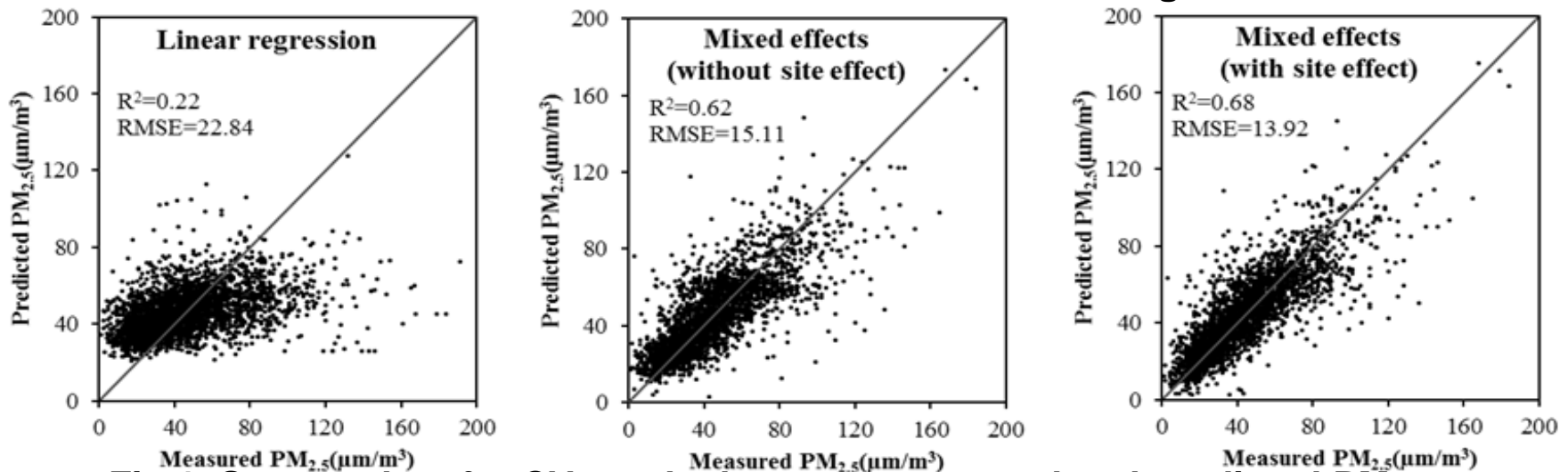


Fig.1. Scatter plots for CV results between measured and predicted PM<sub>2.5</sub>

concentrations from different models.

# Susceptibility of $R^2$ to the number of monitor sites involved

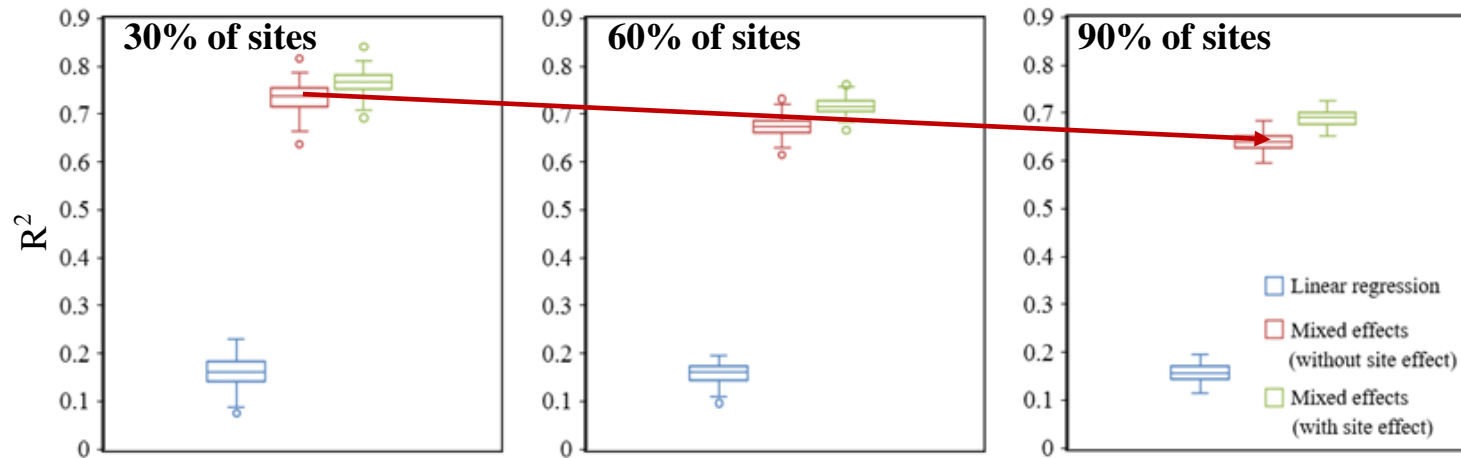


Fig.3. Box plots of  $R^2$  of different models for 3 sensitivity tests.

(Left test involved 30% of sites, center and right employed 60% and 90%, respectively. Stations were chosen randomly for given number of sites and repeated 100 times)

- The method has a good performance.
- **$R^2$  of this model reduced when the number of monitor sites increased, which was opposite to the previous study (Xie et al. 2015).**
- This was partly because the method insufficiently represented the spatial and temporal relationship between AOD and  $PM_{2.5}$  over the given region, especially in heterogeneous area.



**CHE**



中国科学院遥感与数字地球研究所  
Institute of Remote Sensing and Digital Earth ,CAS

# VALIDATION AND COMPARISON OF **AATSR AOD L2** PRODUCTS **OVER CHINA**

***Yahui CHE***

*Yong XUE, Jie GUANG, Jianping GUO, Ying LI, Cheng FAN, Yanqing XIE*  
*RADI/CAS, Beijing*

*Web: [www.tgp.ac.cn](http://www.tgp.ac.cn)*

*Email: [yx9@hotmail.com](mailto:yx9@hotmail.com)*



## Summary:

The main purpose of this work is to **validate different** performances of **AATSR AOD retrieval algorithms** over China from Aerosol\_cci project, including the Swansea algorithm (SU), the ATSR-2/ATSR dual view aerosol retrieval algorithm (ADV), and the Oxford-RAL Retrieval of Aerosol and Cloud algorithm (ORAC), using ground-based data from AERONET and CARSNET (China Aerosol Remote Sensing Network) in 2007, 2008 and 2010.

Five parts will be represented by this poster:

- Part 1 is about introduction of AERONET and CARSNET
- Part 2 is introduction of ADV, ORAC and SU algorithms.
- Part 3 validation results, including seasonal validations
- Part 4 is an analysis of results
- Part 5 is a summary.

**FARLIE**

**FAN**



twas



# 15th CAS-TWAS-WMO Forum 15th AeroCom and 4th AeroSAT Workshops

19-24 September, 2016 | Beijing, China

## Retrieval of Aerosol Optical Depth and Land Surface Reflectance from FY3/MERSI

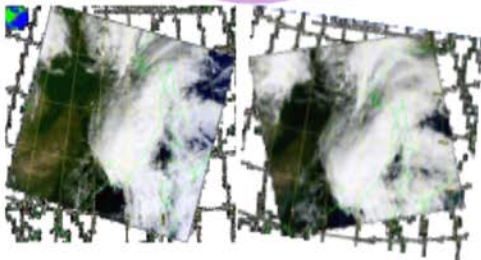
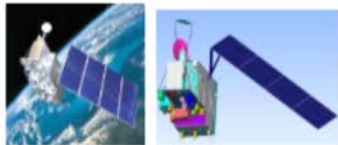
*Cheng FAN*

*Jie GUANG, Yong XUE, Aojie DI, Lu SHE, Yahui CHE  
RADI/CAS, Beijing*

[www.tgp.ac.cn](http://www.tgp.ac.cn)

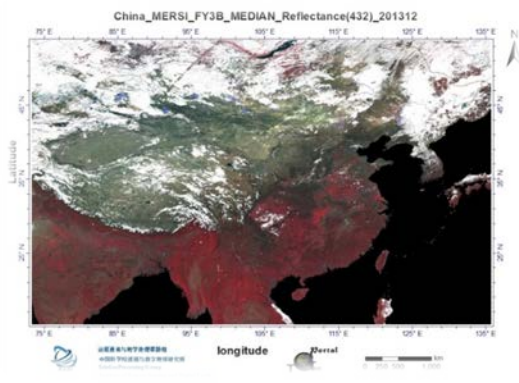
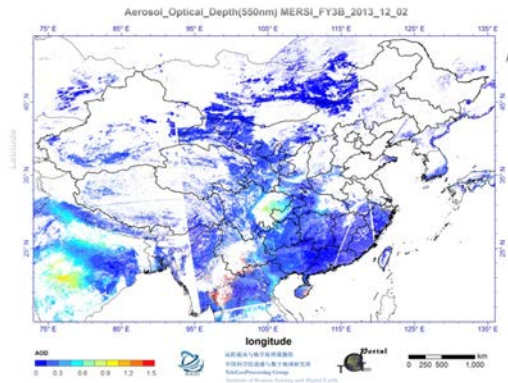
*Email: [chengfjane@163.com](mailto:chengfjane@163.com) [yx9@hotmail.com](mailto:yx9@hotmail.com)*

FY3A (10:05 AM) FY3B (13:00 PM)  
 (May 2008~ ) (Nov 2010~)

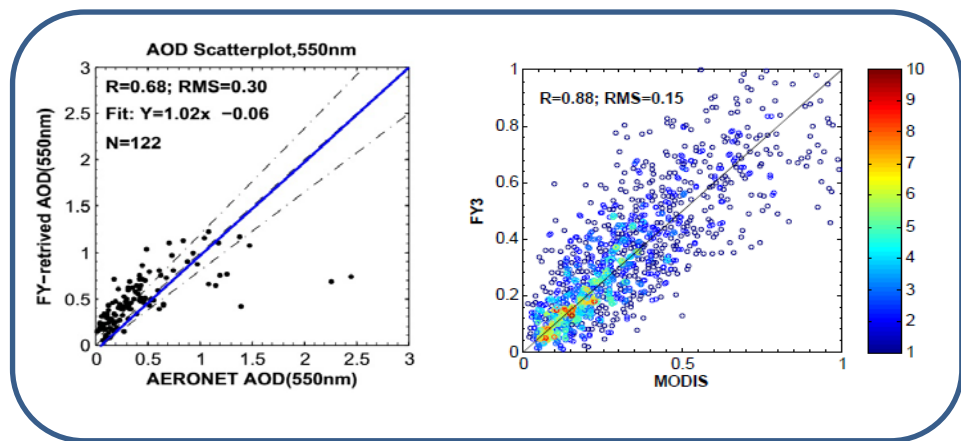


FY3/MERSI image  
 Left: FY3A, Right: FY3B

# Results



# Validations



**GARAY**

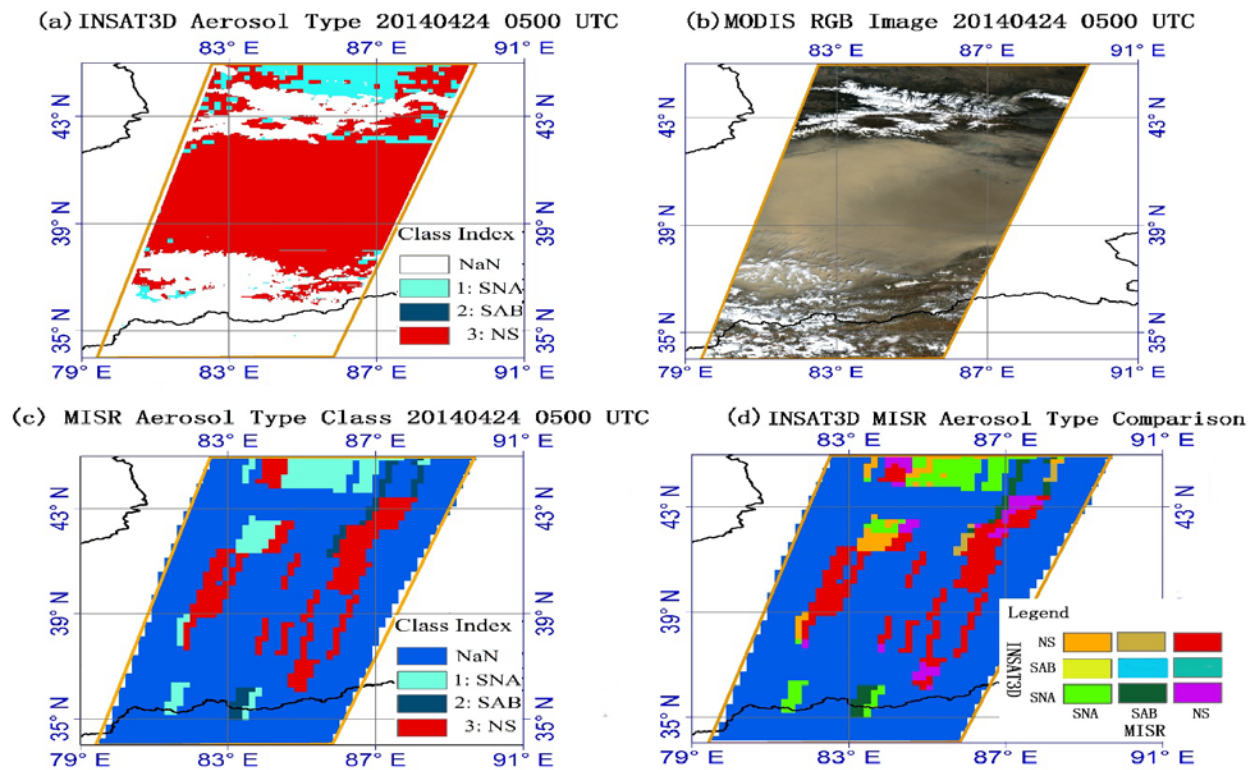


**GINOUX**

**GUANG**

# Aerosol Optical Depth Retrieval in Xinjiang Region Using Indian National Satellite (INSAT 3D) Data

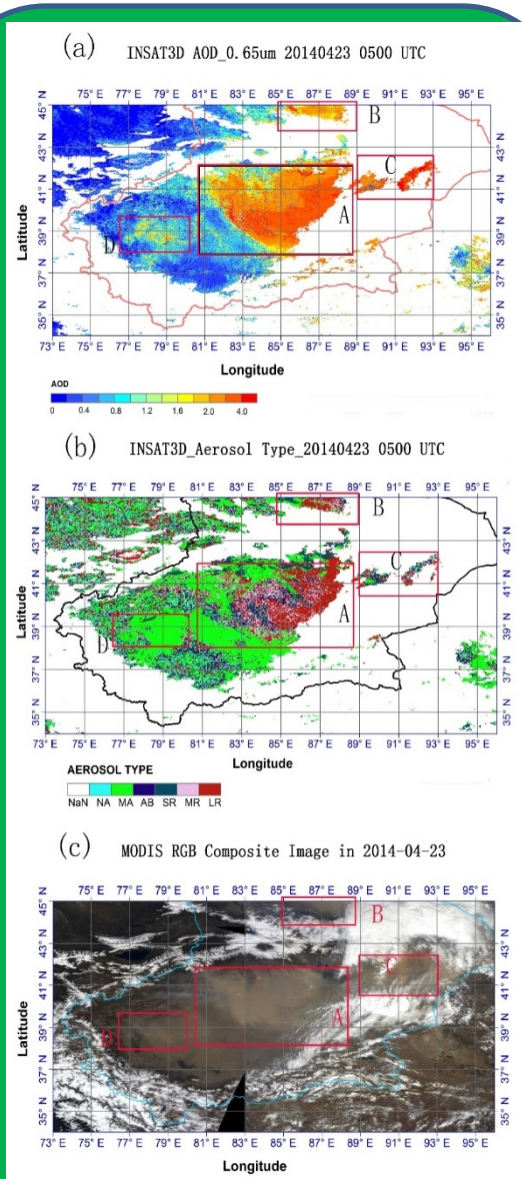
- This paper proposed an improved algorithm to retrieve **AOD** over bright surface (desert area) using a **Geostationary satellite**-Indian National Satellite (INSAT 3D) data.



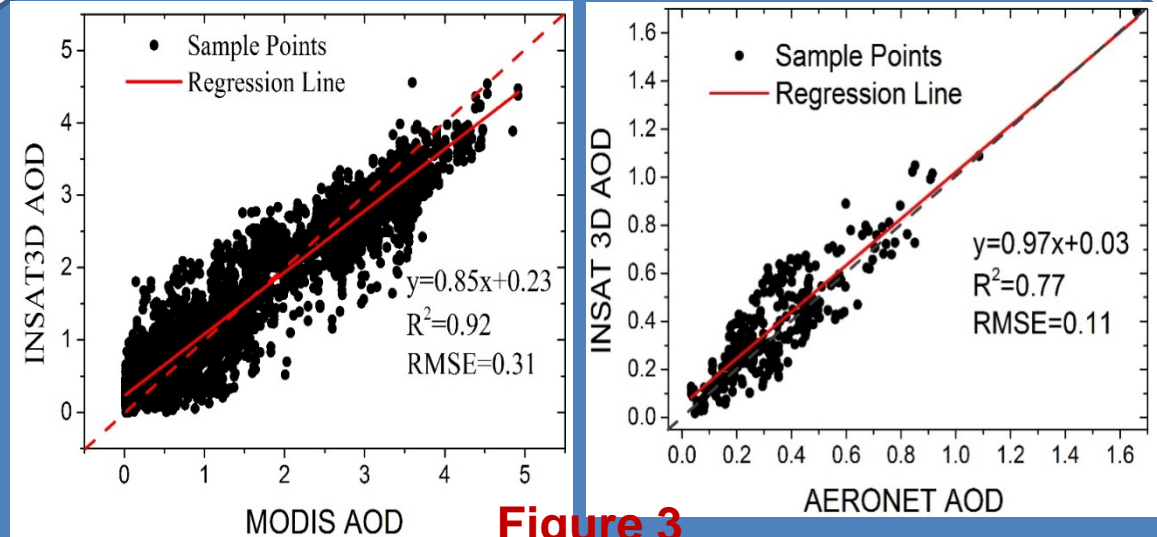
More than 80 percent of valid pixels get the same types!

Figure 1. comparison between INSAT-3D retrieved aerosol type and MISR Aerosol type (SNA: Spherical Non-absorbing, SAB: Spherical Absorbing, NS: Non-Spherical)

# Aerosol Optical Depth Retrieval in Xinjiang Region Using Indian National Satellite (INSAT 3D) Data



**Figure 2**



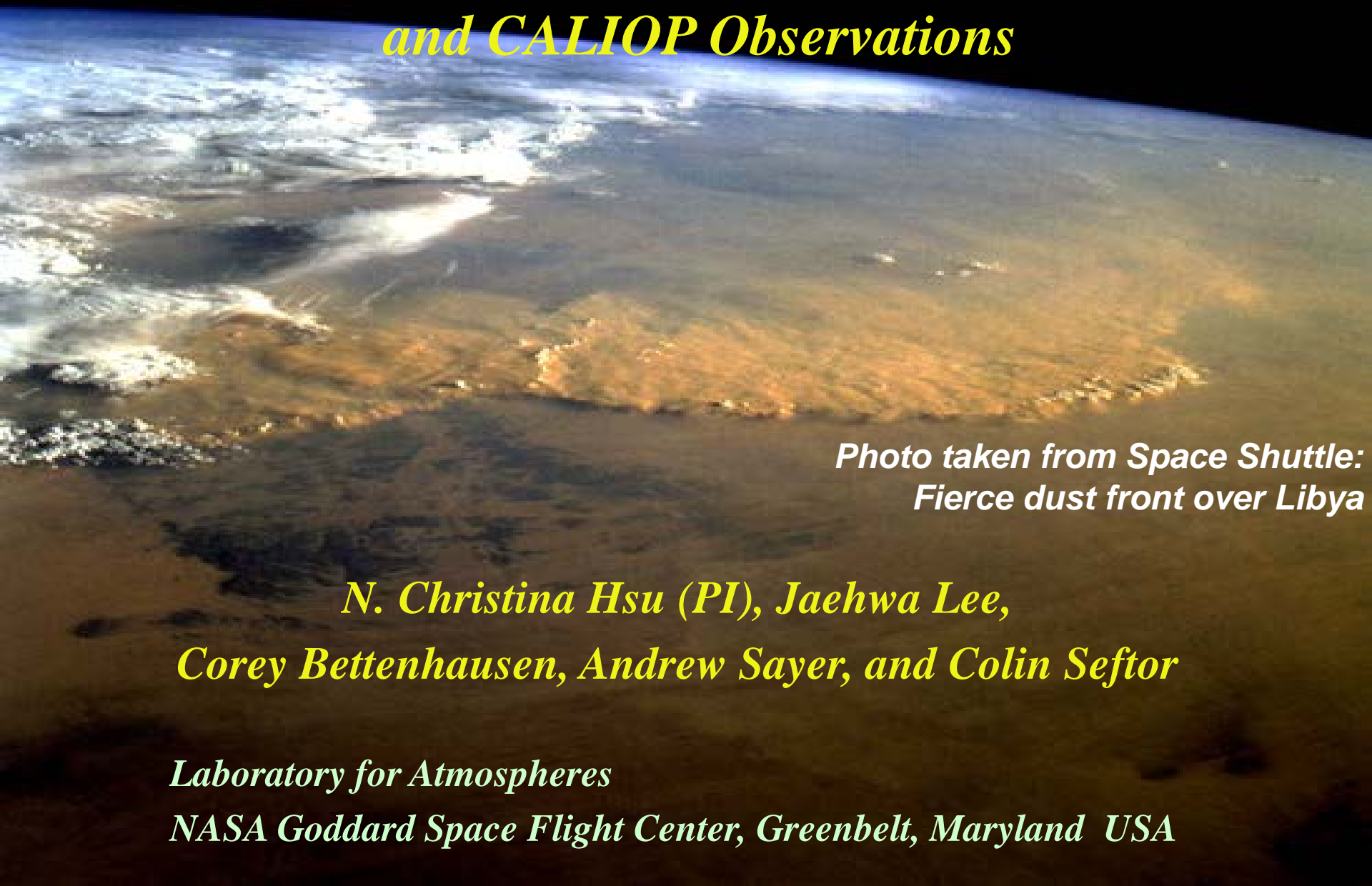
**Figure 3**

- **Figure 2. Comparison among the retrieved AOD (a), aerosol types (b), and MODIS RGB image (c) at the corresponding time.**
- **Figure 3. Validation of retrieved AOD from INSAT3D with MODIS AOD product and AERONET in-situ AOD.**

Correspondence: [xueyong@radi.ac.cn](mailto:xueyong@radi.ac.cn),  
[guangjie@radi.ac.cn](mailto:guangjie@radi.ac.cn)  
[www.tgp.ac.cn](http://www.tgp.ac.cn)

**HSU**

# *Retrieving the Height of Smoke and Dust Aerosols by Synergistic Use of VIIRS, OMPS, and CALIOP Observations*



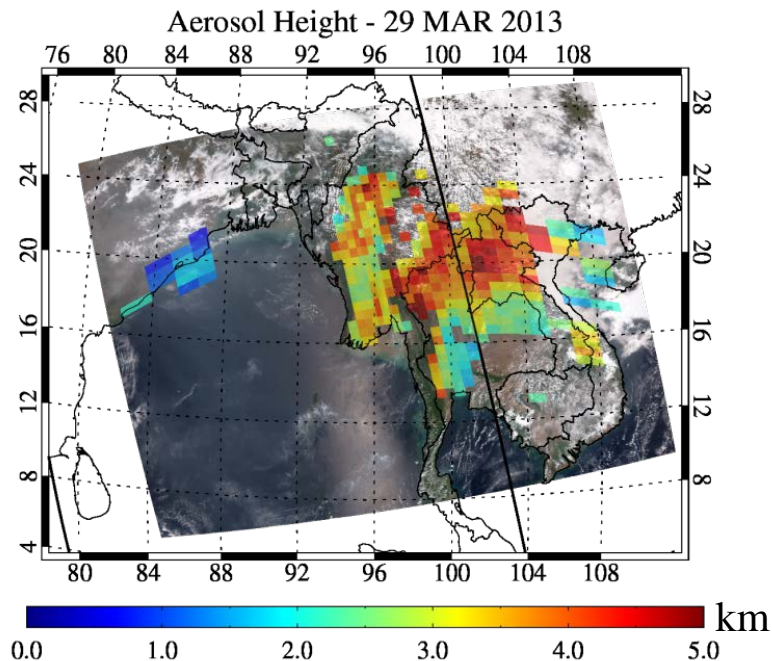
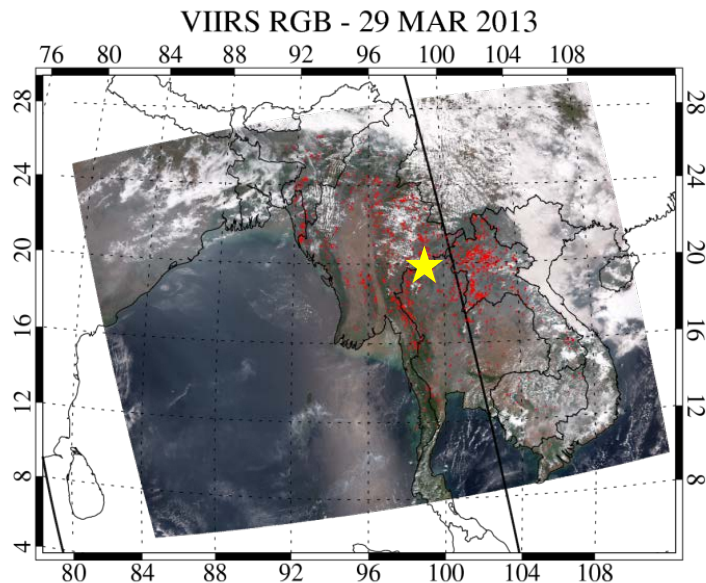
*Photo taken from Space Shuttle:  
Fierce dust front over Libya*

*N. Christina Hsu (PI), Jaehwa Lee,  
Corey Bettenhausen, Andrew Sayer, and Colin Seftor*

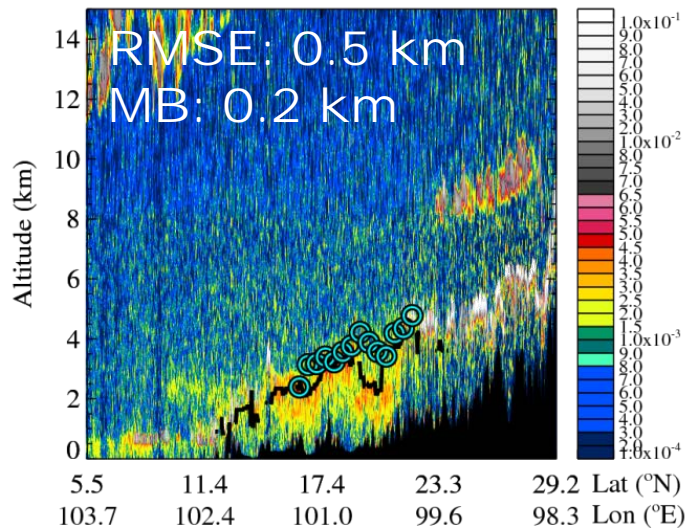
*Laboratory for Atmospheres*

*NASA Goddard Space Flight Center, Greenbelt, Maryland USA*

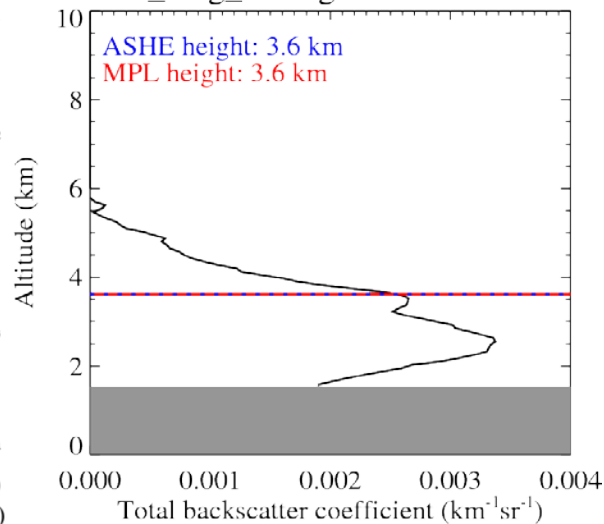
# Product Validation Using 7SEAS field campaign Data



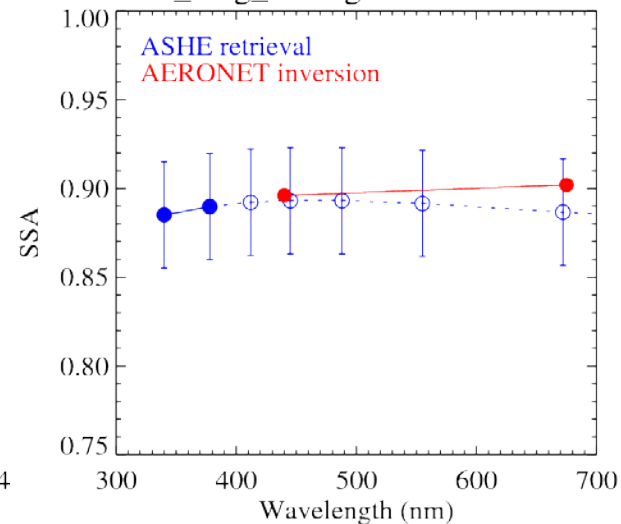
CALIOP Total Attenuated Backscatter



Doi\_Ang\_Khang - 29 MAR 2013



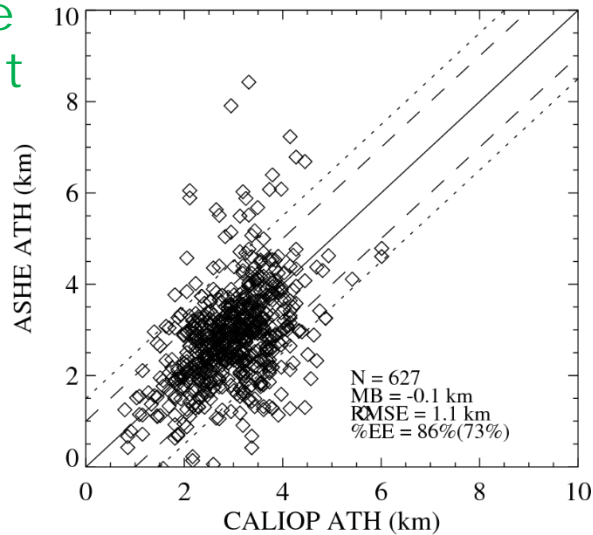
Doi\_Ang\_Khang - 29 MAR 2013



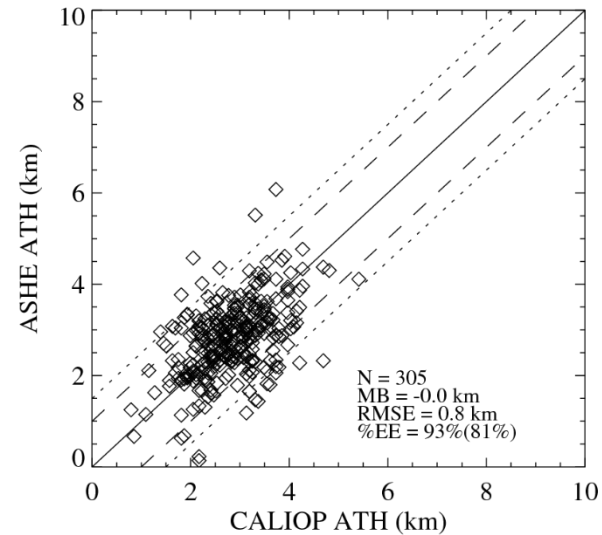
# Retrieval Performance Evaluation Using Multi-year CALIOP and AERONET Data

Plume  
Height

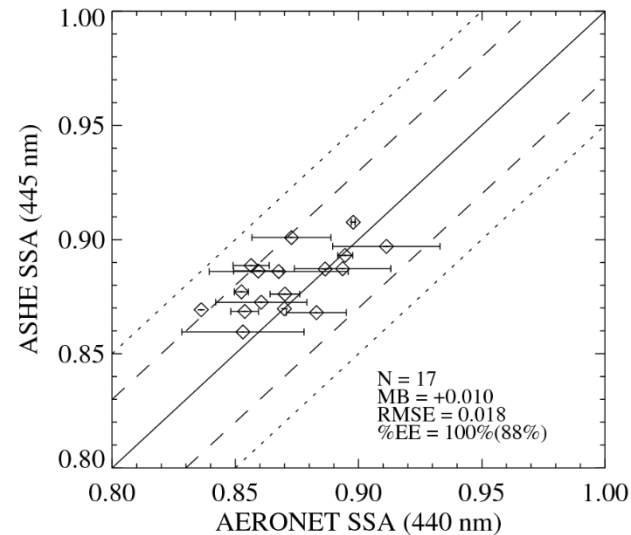
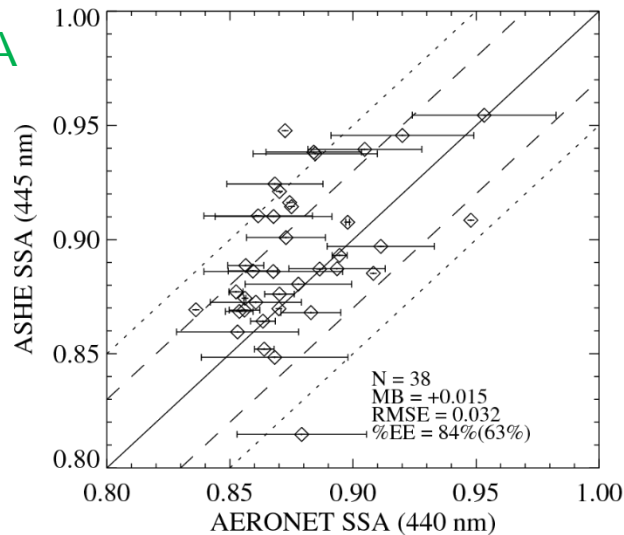
Unfiltered



Quality Assured



SSA





**HU**



# **Climatology (2002-2014) of aerosol products derived from MODIS, MISR and OMI sensors over the Yangtze River Delta region**

**Kang Hu, K. R. Kumar, Na Kang**

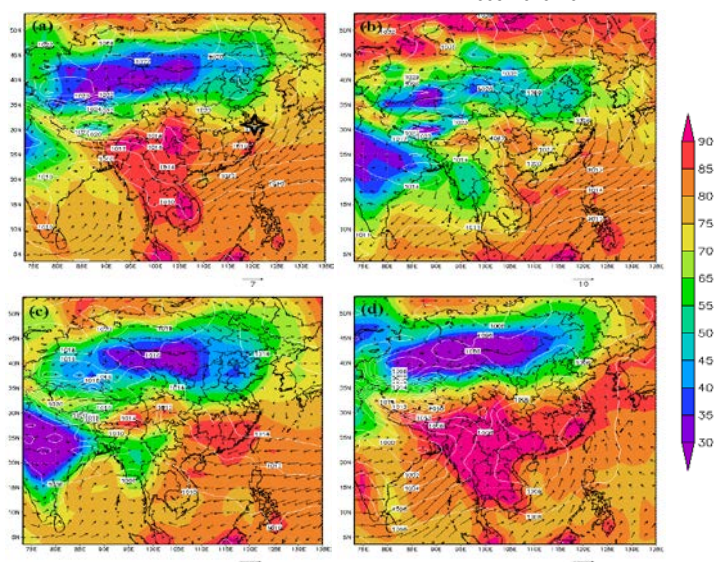
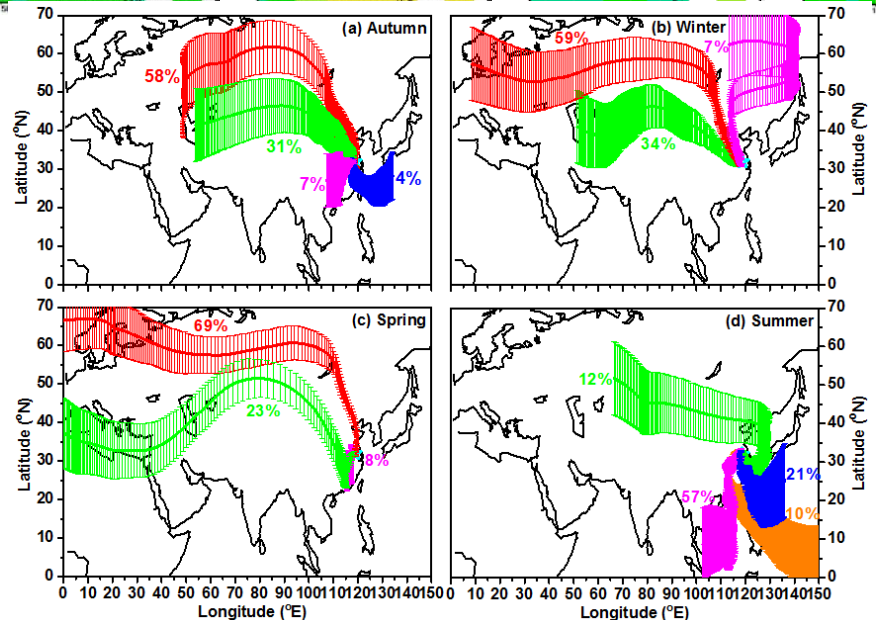
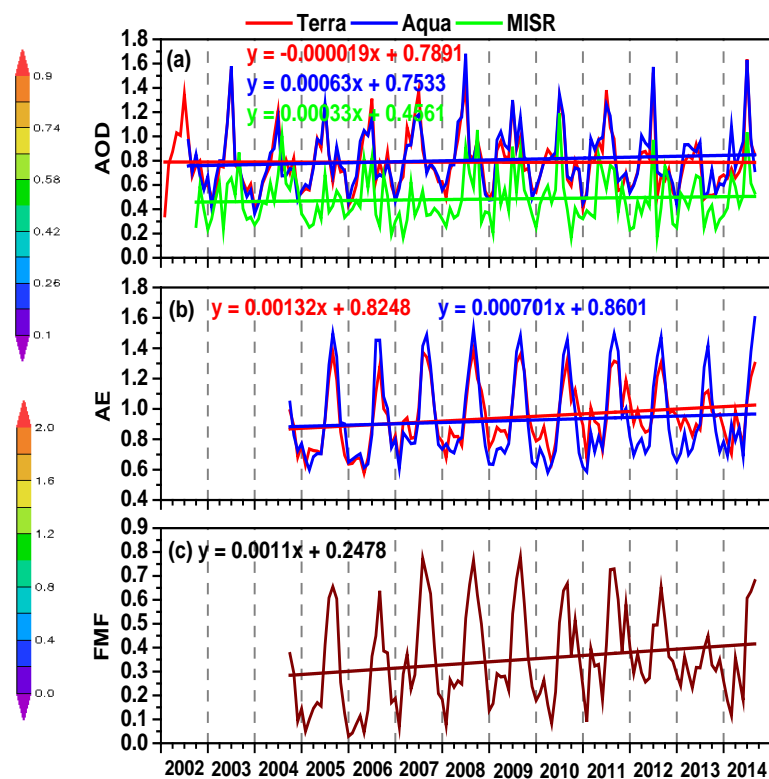
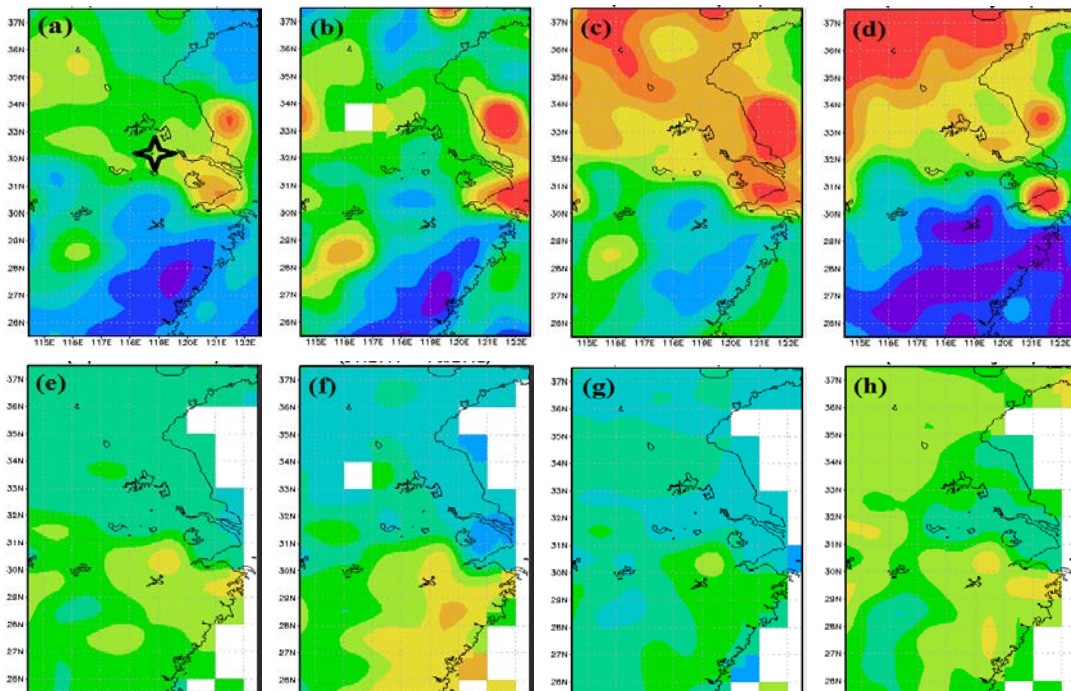
**Key Laboratory for Aerosol-Cloud-Precipitation of CMA**

**Nanjing University of Information Science & Tech.**

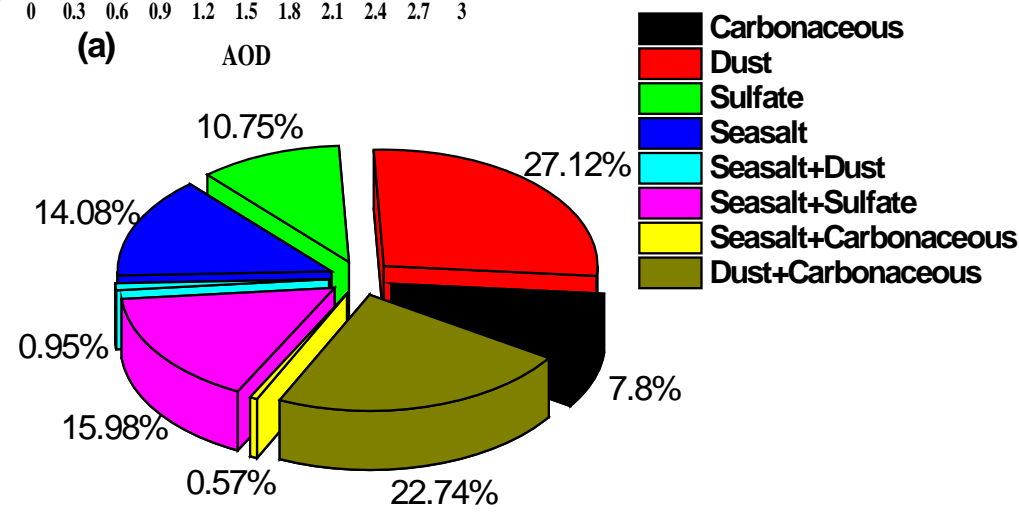
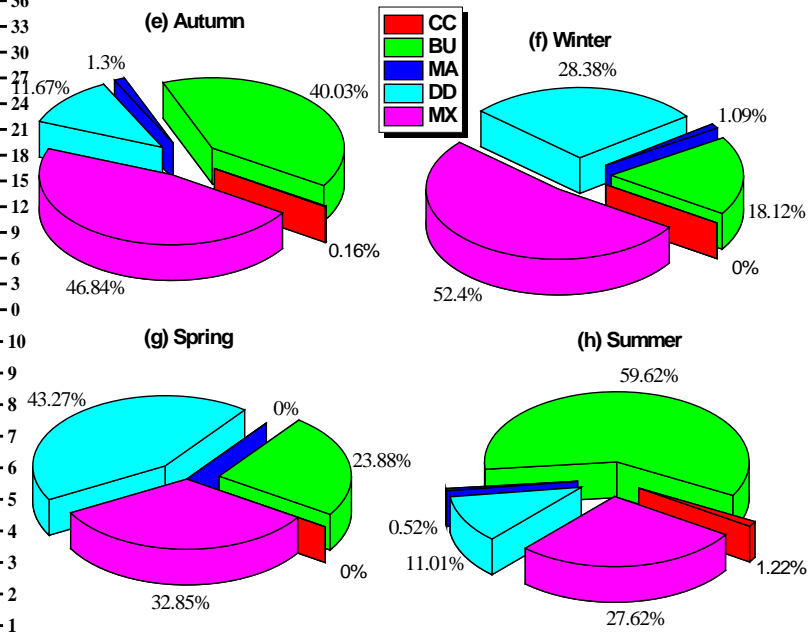
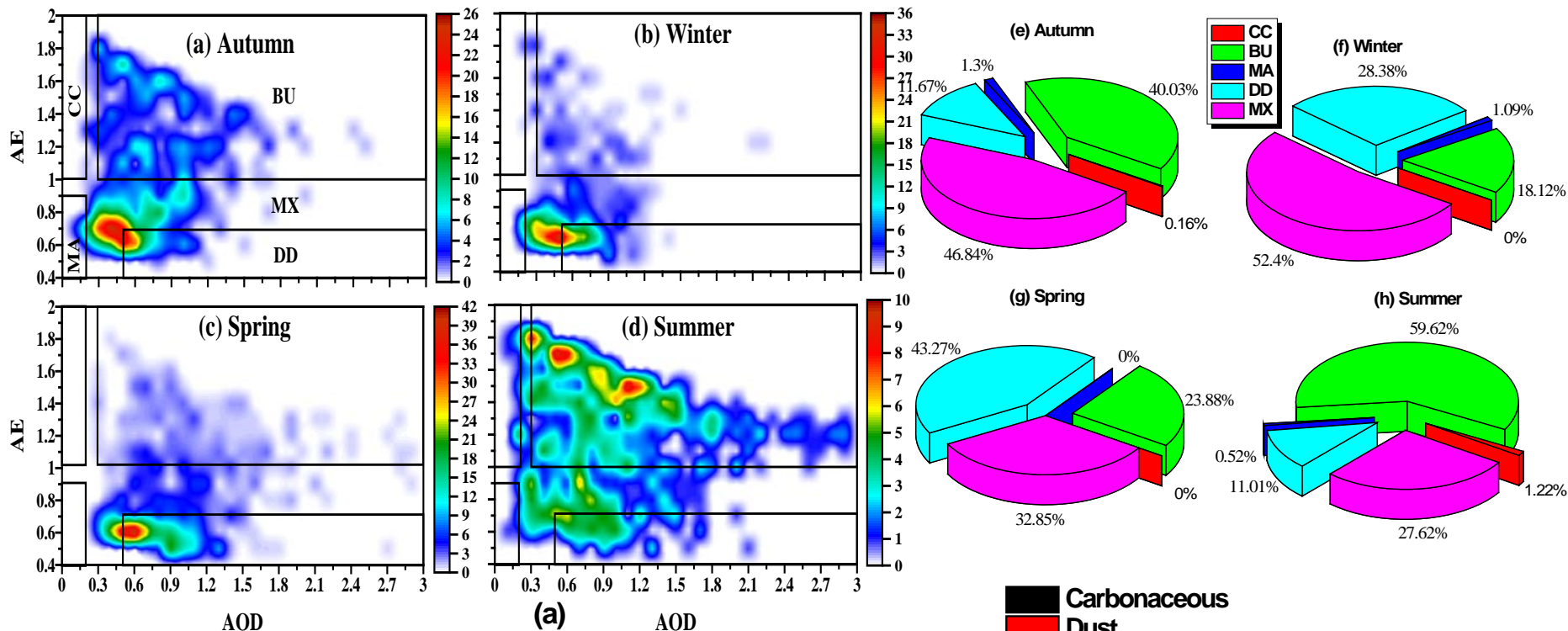
**Nanjing, CHINA**

**Email: 18252085328@126..com**

# Spatial and temporal distribution of AOD, AE over East China



# Aerosol type classification : MODIS – OMI algorithm



THANK YOU

**JIE**

**KAZUMA**



# Temporal and spatial variability of Aerosol optical properties retrieval from sky radiometer observation in Japan sites



**Kazuma Aoki**

**University of Toyama, Japan**



15<sup>th</sup> AeroCom & 4<sup>th</sup> AeroSAT Workshop in Beijing, China: 2016.9.22, Kazuma Aoki

きつときとな大学やちゃ〜

富山大学



### 3. Example of Ground-based observation at several Japan sites.

We started the long-term monitoring of aerosol optical properties by using a sky radiometer since 1990's. The sky radiometer is an automatic instrument that takes observations only in daytime under the clear sky condition without cloud. Observation of direct and diffuse solar intensity of interval was made every ten or five minutes by once (direct measurement every one minute). Ship-borne type, GPS provides the position with longitude and latitude and heading direction of the vessel, and azimuth and elevation angle of sun. Horizon sensor provides rolling and pitching angles. We used seven wavelengths (0.315, 0.4, 0.5, 0.675, 0.87, 0.94, 1.02  $\mu\text{m}$ ). The two wavelengths (0.315 and 0.94  $\mu\text{m}$ ) can be used to estimate total ozone amount and precipitable water. These were used to analysis direct solar irradiance and diffuse solar radiance at fifth wavelength (0.4, 0.5, 0.675, 0.87, 1.02  $\mu\text{m}$ ) by aerosol channel. The aerosol optical characteristics were computed using the SKYRAD.pack version 4.2 developed by Nakajima et al. (1996).

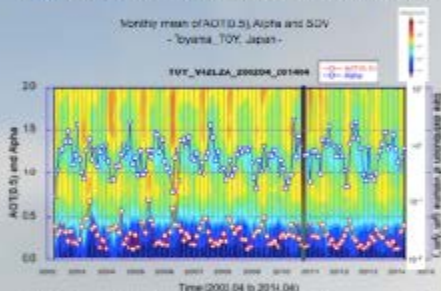


Fig. 3 Aerosol optical thickness at 0.5  $\mu\text{m}$ , Angström exponent and volume size distribution averaged monthly for the period from April 2002 to April 2014 in Toyama

### 4. Maritime Aerosol

Maritime Aerosol optical properties measurements were made in MR10-02 to MR14-06 cruise onboard the R/V Mirai, JAMSTEC. This cruise area included a Japan to Tropical ocean. Figure 6 shows the relationship of aerosol optical thickness at 0.5  $\mu\text{m}$  (Angström exponent) to Latitude (N) between 120E to 180E. Near the coast area, AOT and Alpha is high value and variability. Especially, near the Japan coast.

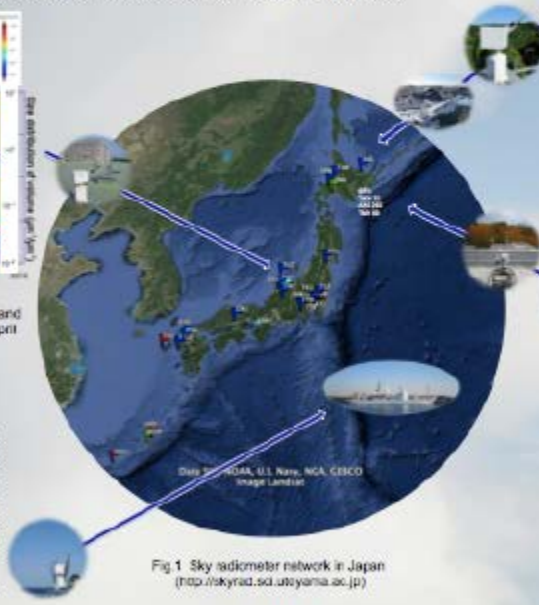


Fig. 1 Sky radiometer network in Japan (<http://skyrad.sdl.uoyama.ac.jp>)

### 5. Summary

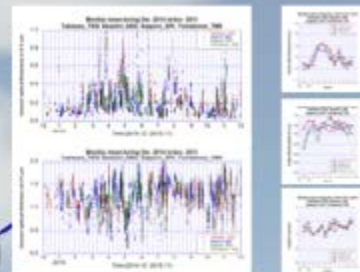


Fig.4 Aerosol optical thickness at 0.5  $\mu\text{m}$ , Angström exponent, and Single scattering albedo during December 2014 to November 2015 in Tokkawa, Abashiri, Sapporo and Tomakomai/Hokkaido area.

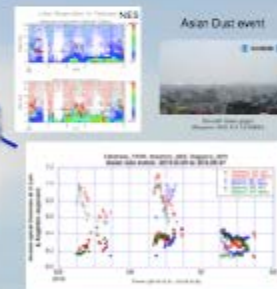


Fig.5 Aerosol optical thickness at 0.5  $\mu\text{m}$  and Angström exponent at 5 to 7 May 2015 in Tokkawa, Abashiri, and Sapporo/Hokkaido at Asian dust event cases.





**KOLLACHI**



*The relationship between characteristics of physiographic and dust storm phenomena in Khuzestan, Iran*

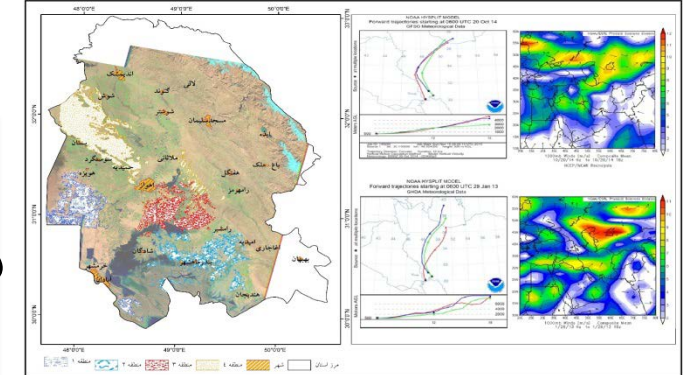
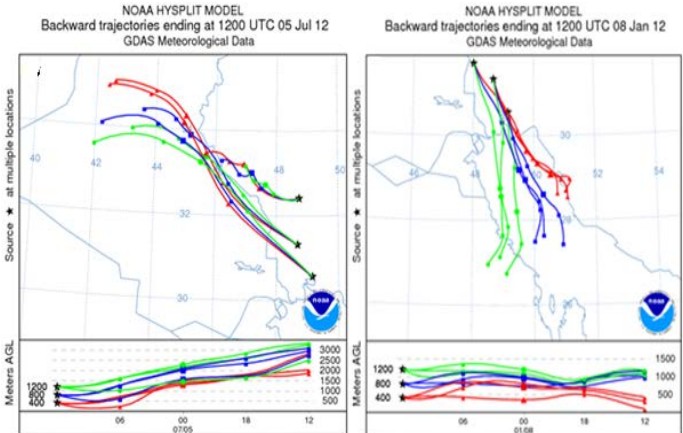
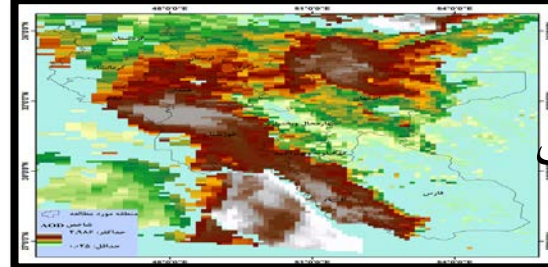
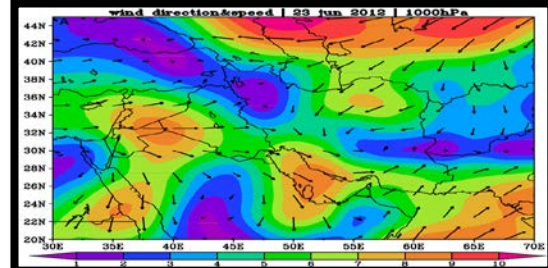
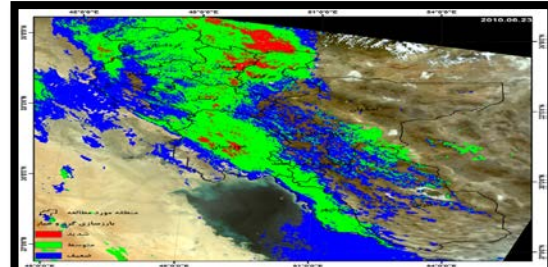
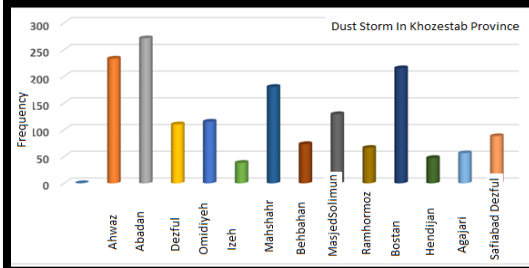
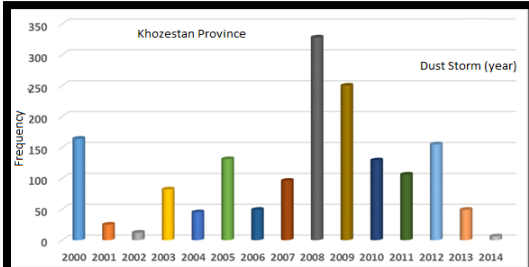
*Abdolnabi Abdeh Kolahchi and Ali Akbar Noroozi*  
[kolahchi@scwmri.ac.ir](mailto:kolahchi@scwmri.ac.ir), [noroozi\\_aa@scwmri.ac.ir](mailto:noroozi_aa@scwmri.ac.ir)

The **aims** of this study: **To investigate** Role of land cover changes in **Internal dust storm generation** in the Khuzestan province, Iran.

- **Data**
  - **Remote Sensing (Landsat 7 and 8)**
    - Various Supervised Classification method (MLC, SAM and SID)
  - **ASTER (DEM)**
    - Elevation, Slope, Aspect
  - **The MODIS (level 2) images and the Brightness Temperature Difference (BTD) Index**
    - For Detection of Dust Phenomenon
  - **Climate Data**
    - (Wind Speed, Wind Direction)
  - **Physiographic**
    - (Slope, Aspect, Soil erosion and Geological sensitive formations)
  - **Modeling Techniques**
    - Tracking the movement of dust particles using wind speed and wind direction

# Conclusions:

The **tracking dust storm** of the regions shows that most of dust comes from **North West direction toward South East** as well as **West direction toward East** during **warm seasons** and from **South direction to North** in **cold seasons**. The **farmland and irrigated farm decrease** and **bare land increase** during 2000-2015. These features, along with physiographic features such as **erosion on formations, salty and swampy soil, low soil moisture, land use changes** caused potential for dust to be targeted west and south west province such as Khuzestan province.

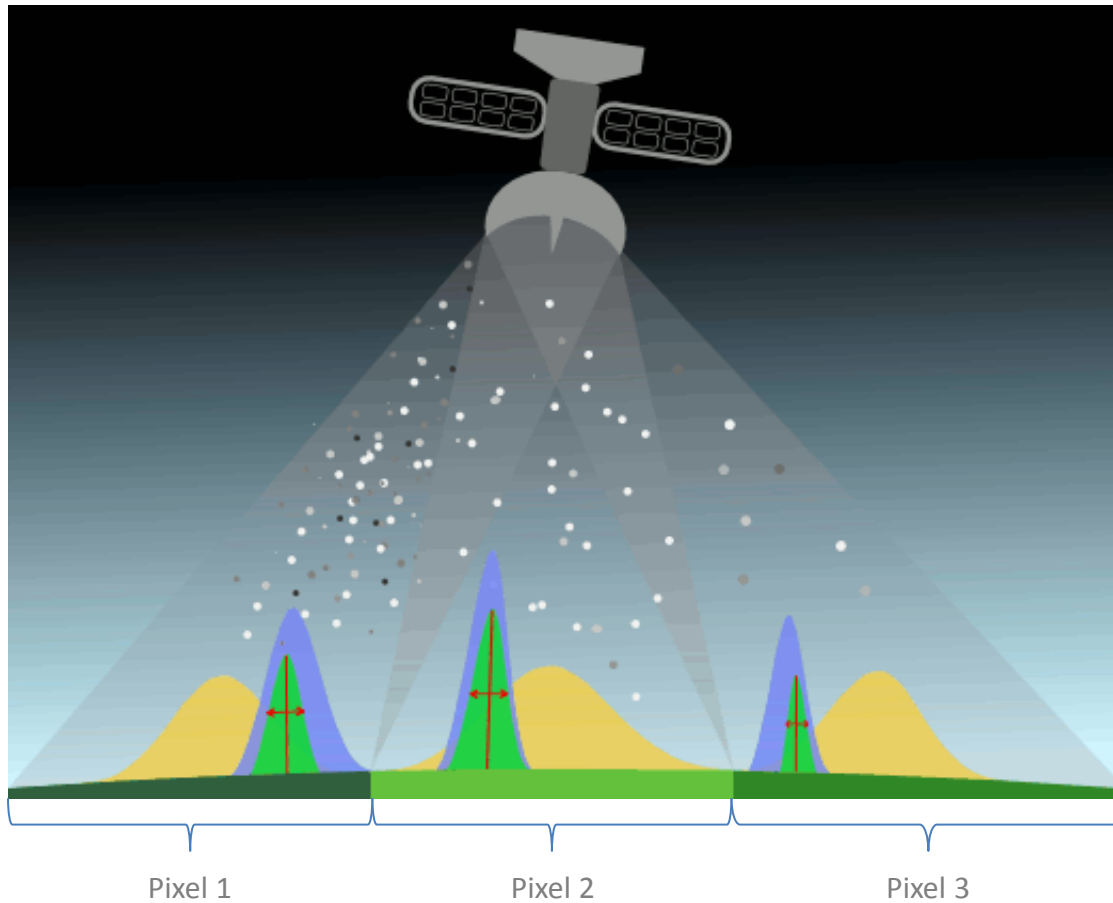


**LACAGNINA**

**LIPPONEN**

# Pixel level uncertainty estimates for AOD using Bayesian Dark Target algorithm

Antti Lipponen



## Probability distributions:

### Prior distribution

- Spatial correlations
- Seasonality
- Non-negativity of AOD

### Likelihood distribution

- Connects observations to unknowns
- Observation model
- Uncertainty models

### Posterior distribution

- Solution to our retrieval
- Pixel level uncertainty estimates

LI

**LIU**





# NOAA VIIRS Dark Target-Bright Surface Aerosol Optical Depth Algorithm

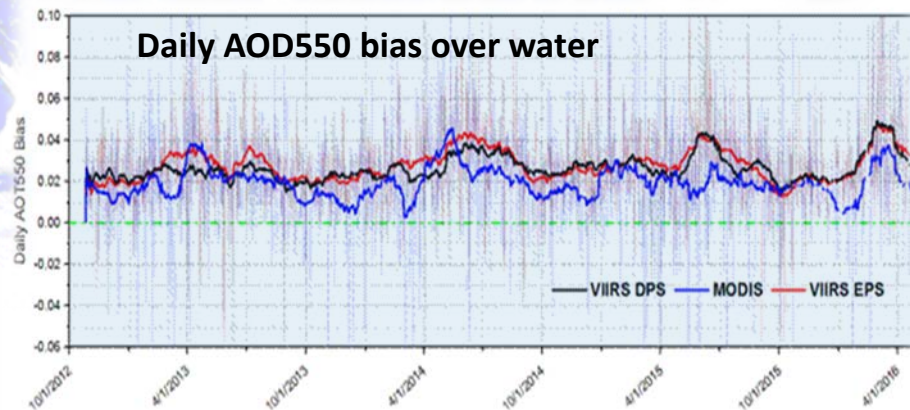
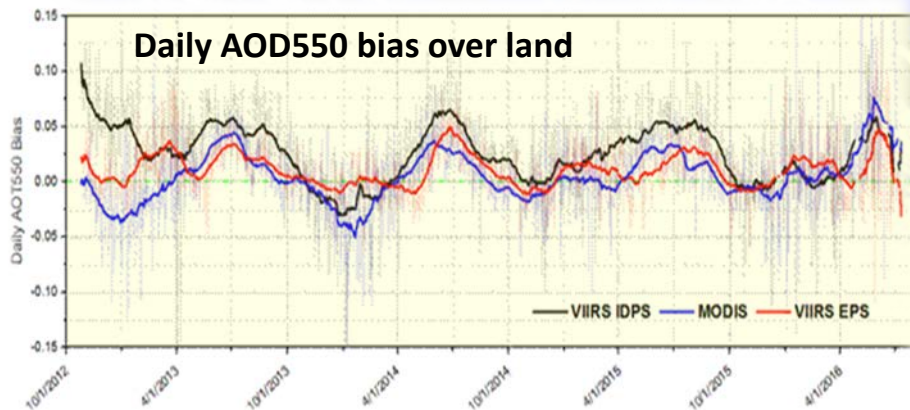


Hongqing LIU<sup>1,2</sup> ([Hongqing.Liu@noaa.gov](mailto:Hongqing.Liu@noaa.gov)), Hai ZHANG<sup>1,2</sup>, Istvan LASZLO<sup>2,3</sup>, Shobha KONDRAGUNTA<sup>2</sup>, Lorraine REMER<sup>4</sup>, Pubu CIREN<sup>1,2</sup>, Jingfeng HUANG<sup>2,5</sup>, and Stephen SUPERCZYNSKI<sup>2,6</sup>

1. I.M. Systems Group, Inc., Rockville, USA
2. National Oceanic and Atmospheric Administration, College Park, USA
3. Department of Atmospheric and Oceanic Science, University of Maryland, College Park, USA
4. Joint Center for Earth systems Technology, University of Maryland Baltimore County, Baltimore, USA
5. Earth System Science Interdisciplinary Center, University of Maryland, College Park, USA
6. Systems Research Group, College Park, USA

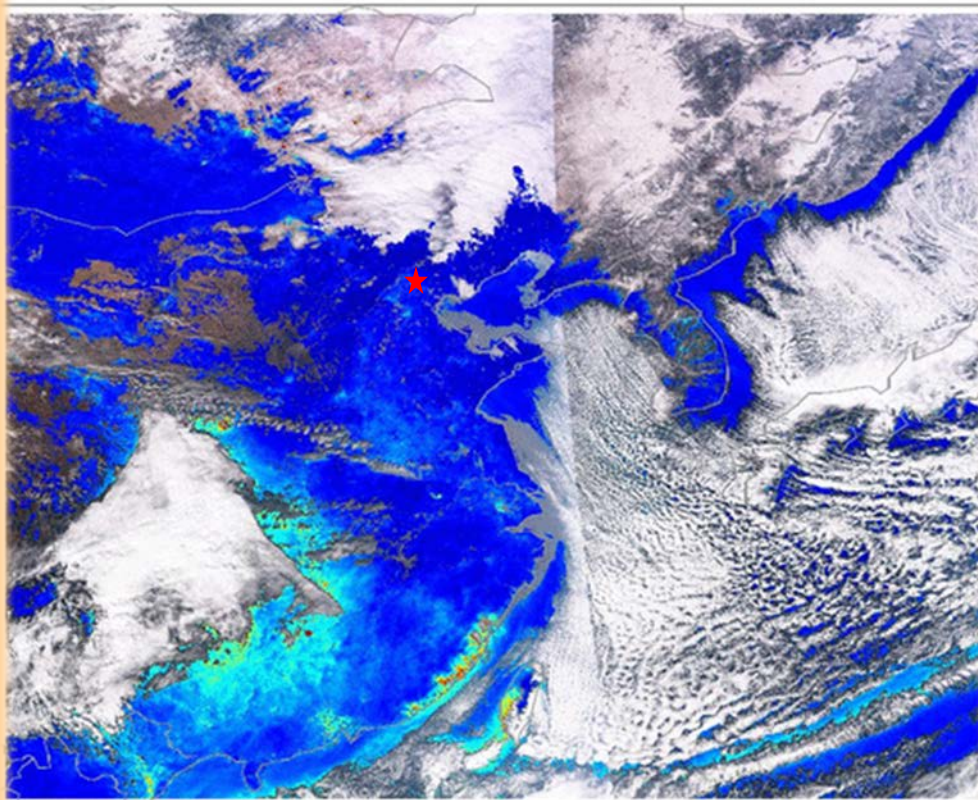
**A new algorithm (EPS) was developed at NOAA/NESDIS to retrieve aerosol from multispectral, single-look, unpolarized Visible Infrared Radiometer Suite (VIIRS) reflectances.**

- **Retrieves over both dark and bright snow-free surface**
- **High spatial resolution (0.75km)**
- **Improves over current operational VIIRS aerosol product (IDPS)**
- **Validation shows performance is comparable to MODIS**

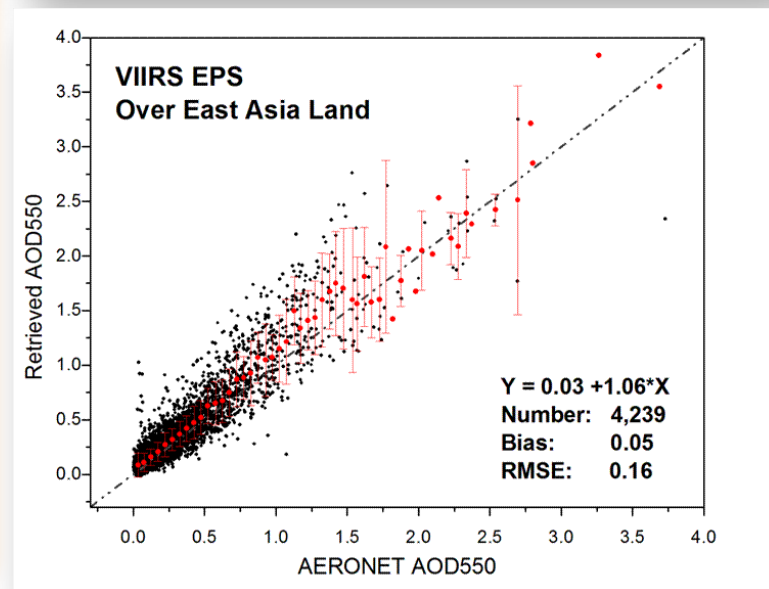
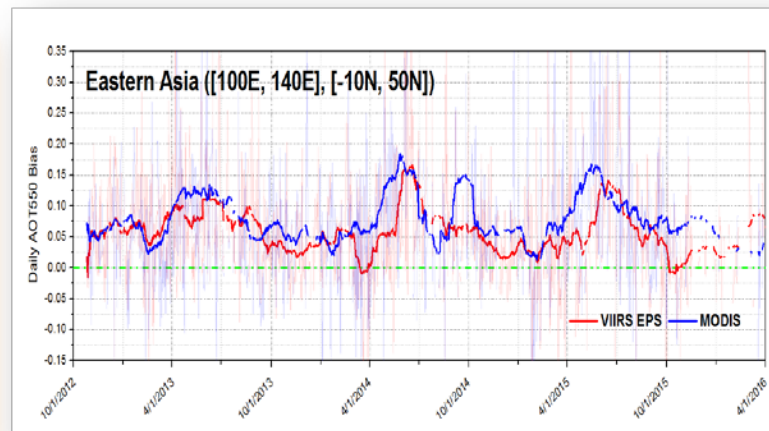


# Focus on Asia

High Quality Aerosol Optical Depth at 550nm (2015001)



-0.1 0.1 0.3 0.5 0.7 0.9 1.1 1.3 1.5



**LU**

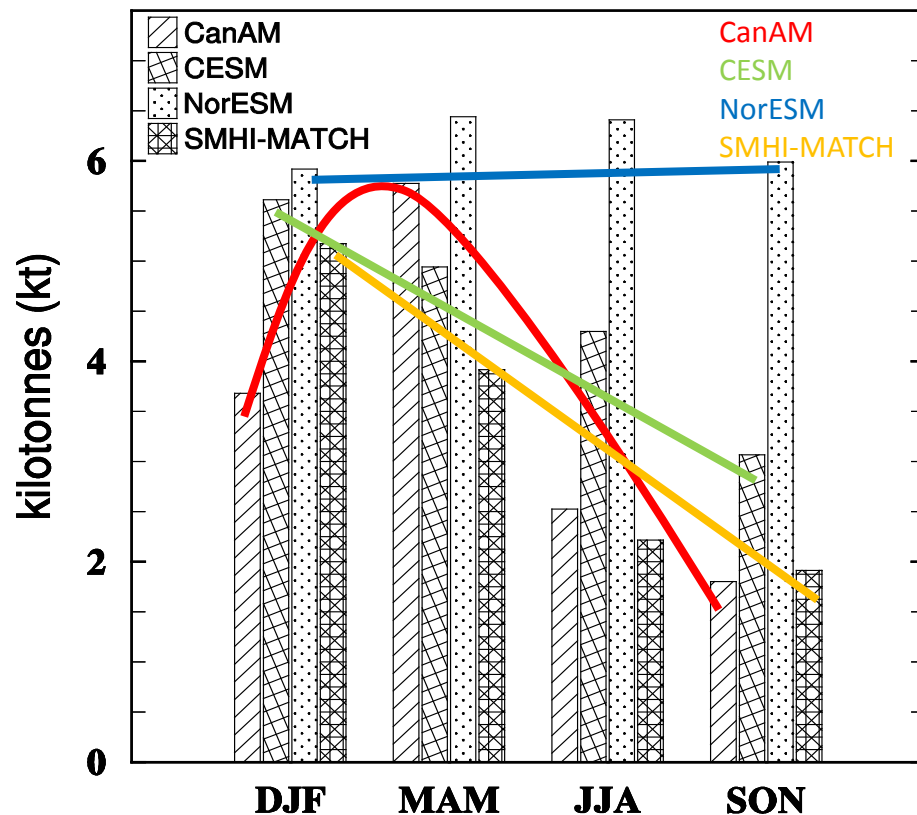
**MAHMOOD**

# Seasonality of global and Arctic black carbon processes in AMAP models

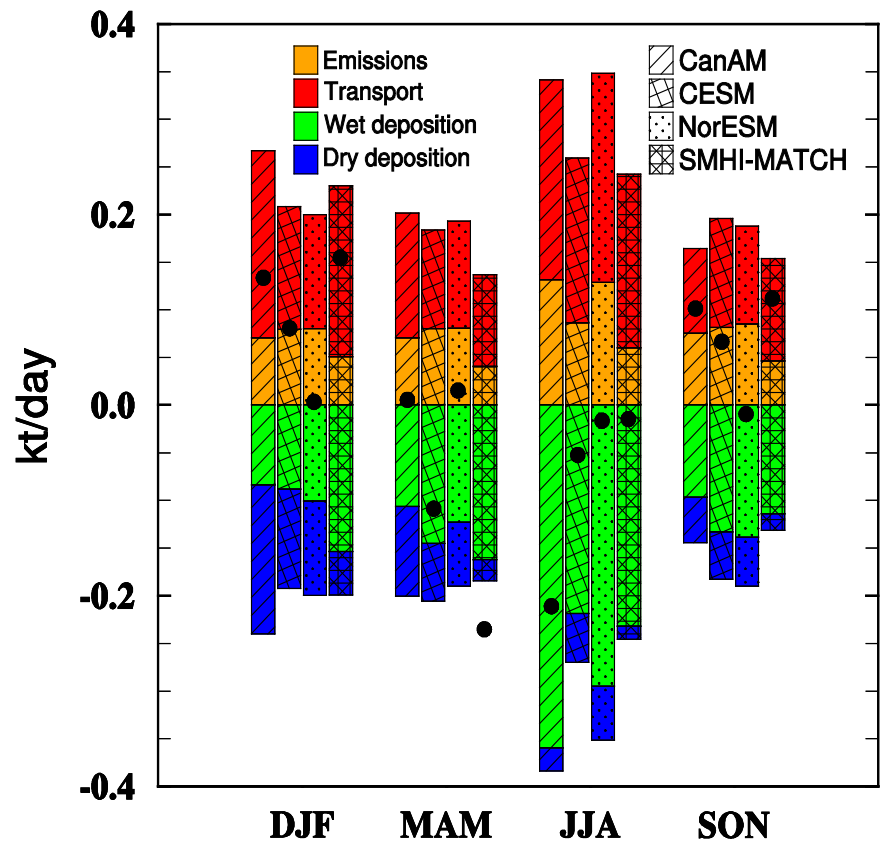
R. Mahmood, K. von Salzen, M.G. Flanner, M. Sand, J. Langner, H. Wang, L. Huang

Postdoc (SEOS, University of Victoria, Victoria BC, Canada), Funding: NETCARE project

### BC Burden in Arctic



### BC budgets in Arctic



**MATTOO**

**MEI**

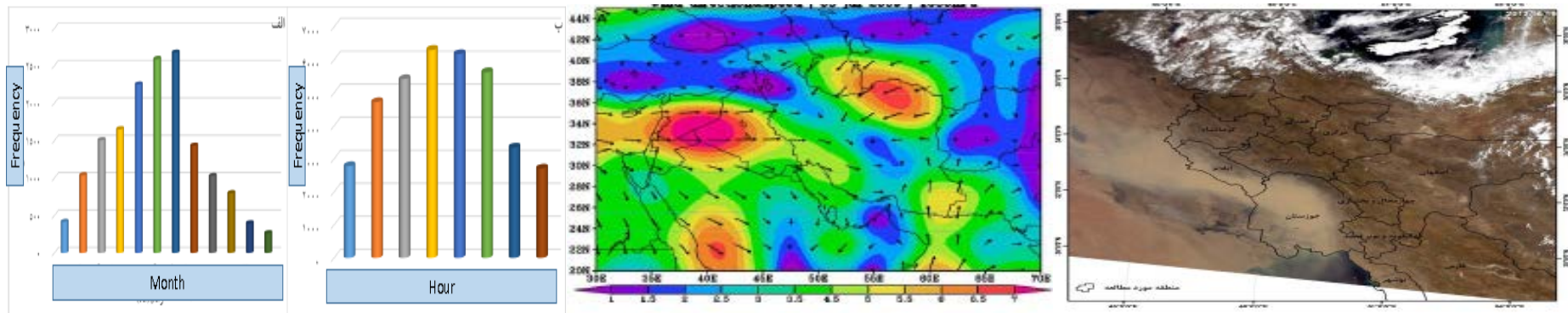
**NOROOZI**



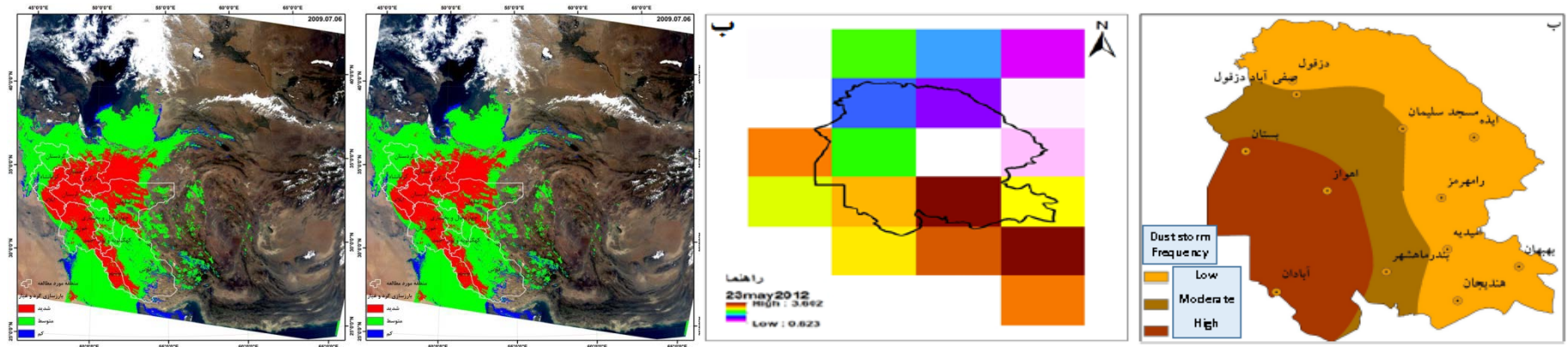
# Identify areas with dust storm potential of physiographic and climatic characteristics

- **Ali Akbar Noroozi & Elham Haghnejad** [noroozi.aa@gmail.com](mailto:noroozi.aa@gmail.com) ; [noroozi\\_aa@scwmri.ac.ir](mailto:noroozi_aa@scwmri.ac.ir)

A combination of statistical methods, remote sensing and modeling were used. The data used included: daily synoptic stations, Landsat and MODIS daily satellite images, U and V data from the NOAA Web site and Landuse changes over the period 2000-2015.



Detection of dust storm on satellite imagery and their concentration using images obtained from AOD and BTM indices, represent the largest concentration of dust in the West and South West regions of Khuzestan province, where more than 80 percent of the dust in these areas were consistent with the results of statistical data.



**SHAHZAD**

**SHUKLA**

**THOMAS**

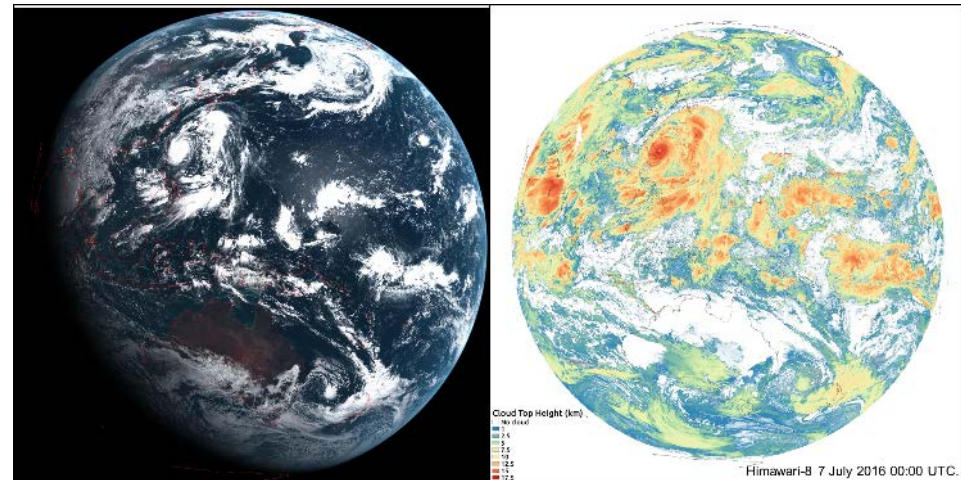
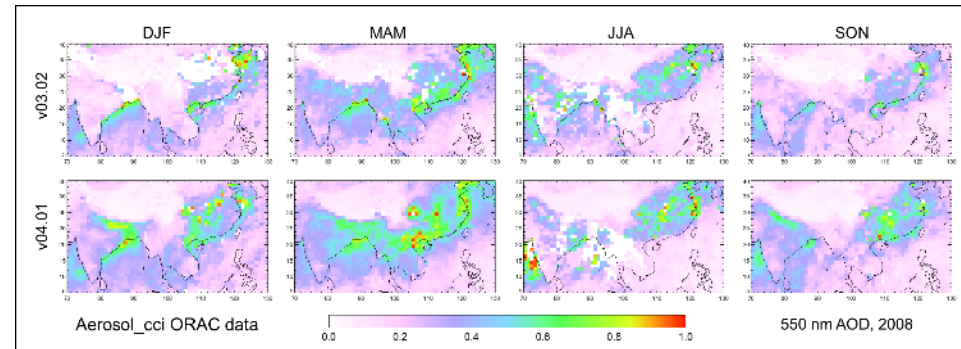
# ORAC

## Optimal Retrieval of Aerosol and Cloud

Gareth Thomas

RAL Space, Rutherford Appleton Lab, UK

- Introduction to, and update on the ORAC algorithm
- A one-stop-shop for the retrieval of aerosol and cloud from visible-IR satellite imagers



# TSAY

- given in AeroCom

**VANDENBUSSCHE**



aerosol  
cci



BIRA-IASB



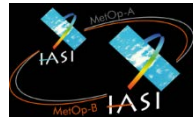
LMD



ULB

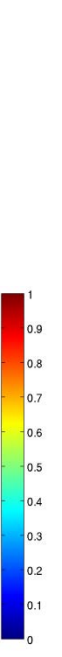
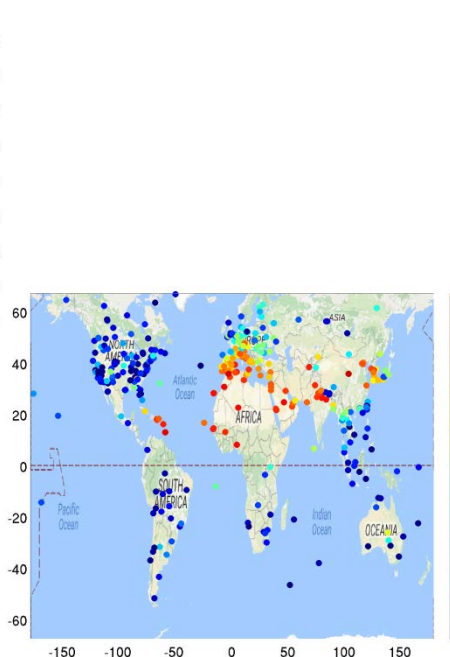
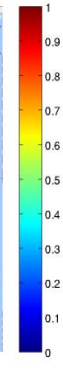
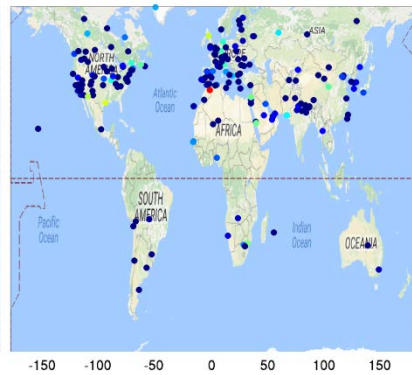
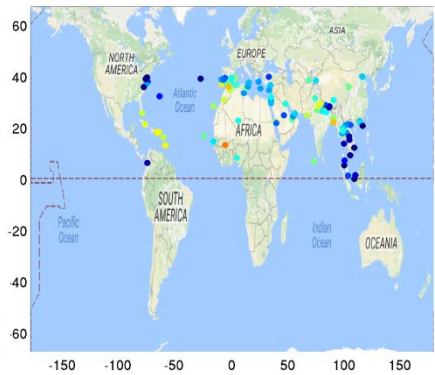
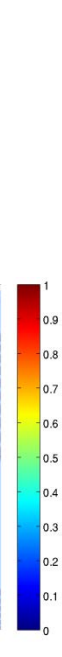
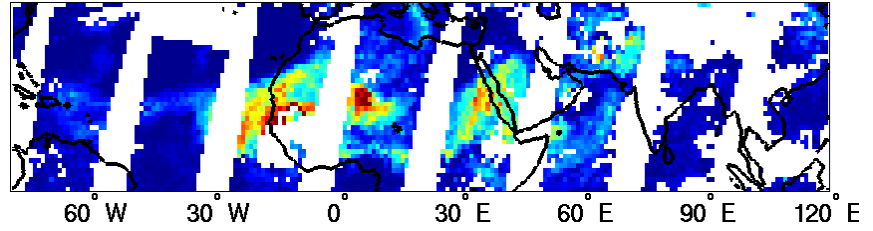
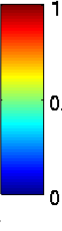
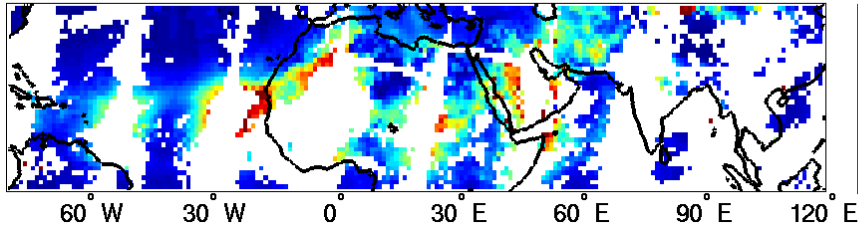
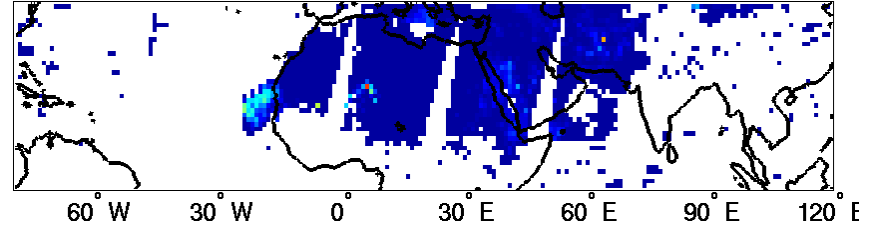
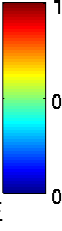
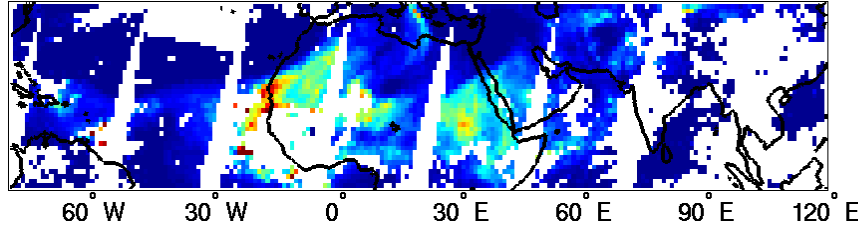


DLR



IASI  
IASI  
IASI

# Poster P2-24: IASI dust Comparison of 4 algorithms



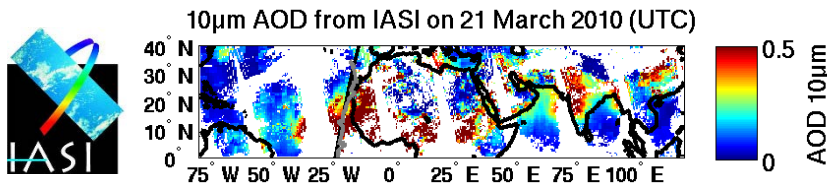
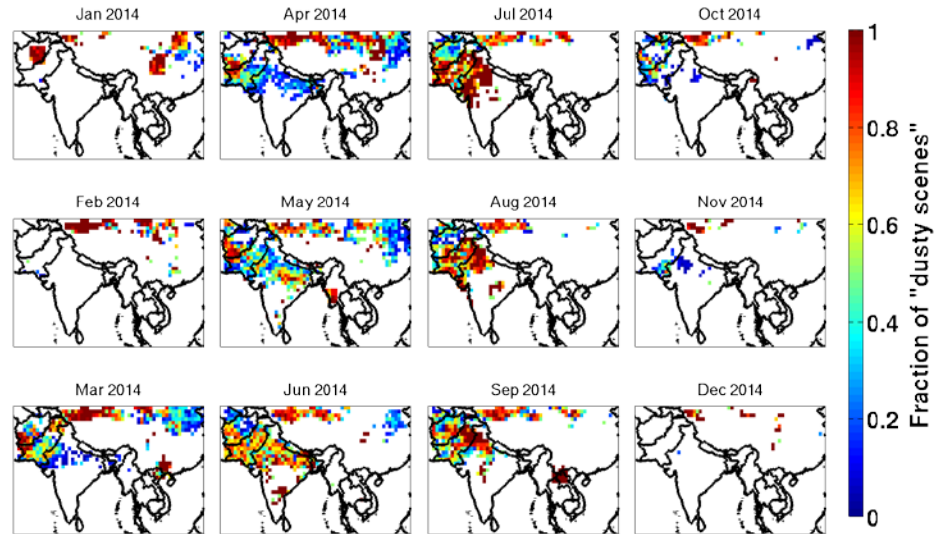
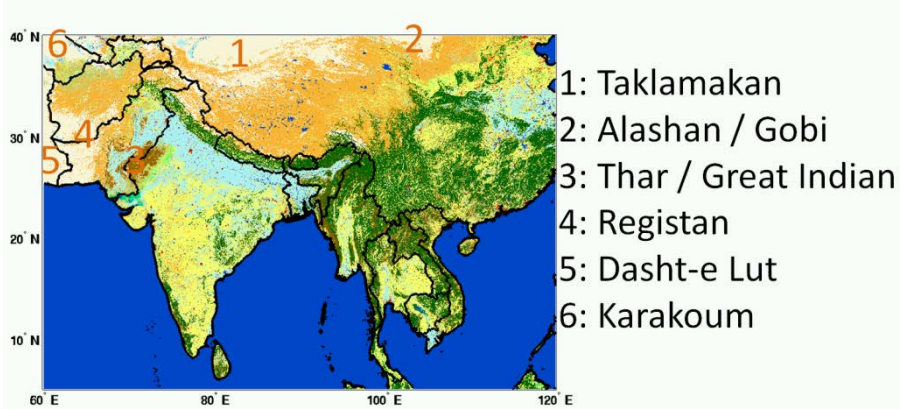
You want to know which is which?  
Come see the poster!



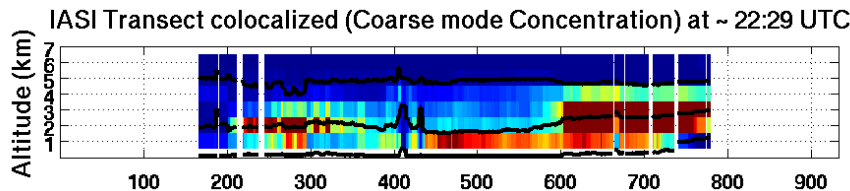
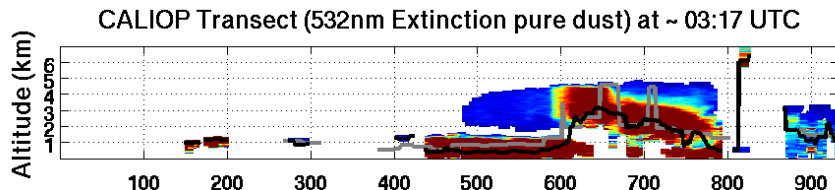
**VANDENBUSSCHE**

# Poster P2-25: Study of Asian dust sources

Seasonality, long term changes, ...



About 10 years of 3D dust distribution over the « dust belt »



Quality checks  
 Ensuring surface sensitivity

Reporting occurrence of dusty scenes

**XIE**

# Image fusion of MODIS AOD products based on maximum likelihood estimate method

## Method

In this study, we use the maximum likelihood method to determine the weights of various images involved in fusion to produce an Brazil's AOD data set from 2007 to 2010 based on two AOD products: MOD DB and MOD DT according to the pixel error

- First we compare the various products to AERONET to determine the error size of the pixels of the various products.
- Then, we study and determine the relationship among the observation error, the value of the remote sensing observations and the value of surface reflectance, so as to determine the error size of the observation value in the absence of the ground sensing station.
- Finally, we determine the weight of the fusion according to the root mean square error of the different products (Xu et al., 2015)

# Study results and conclusion

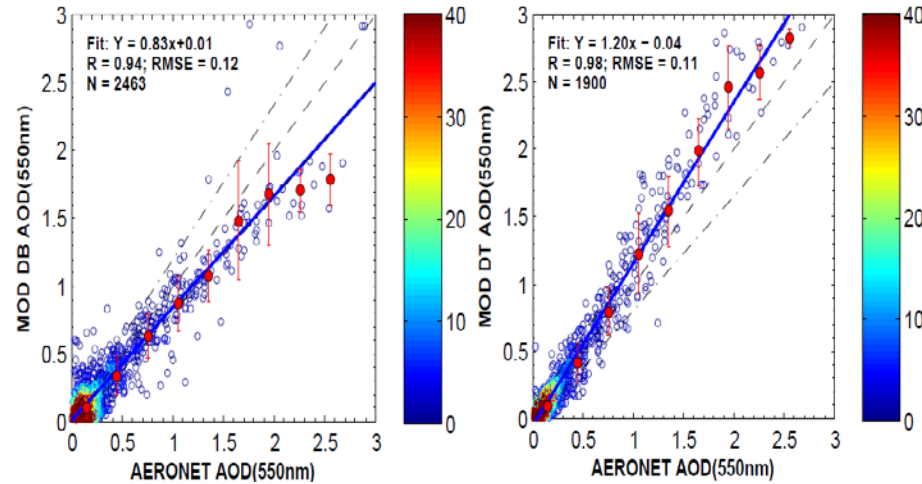


Fig. 1. MOD DB vs AERONET data from 2007 to 2010.

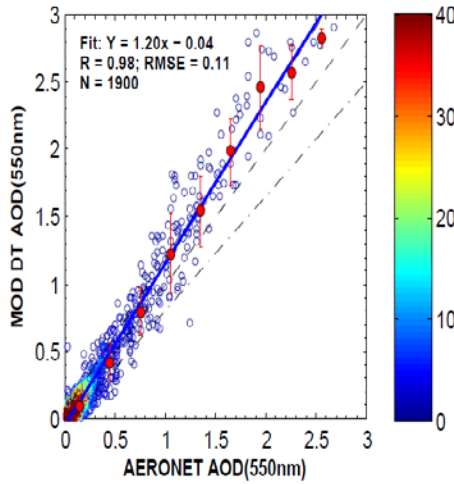


Fig. 2. MOD DT vs AERONET data from 2007 to 2010.

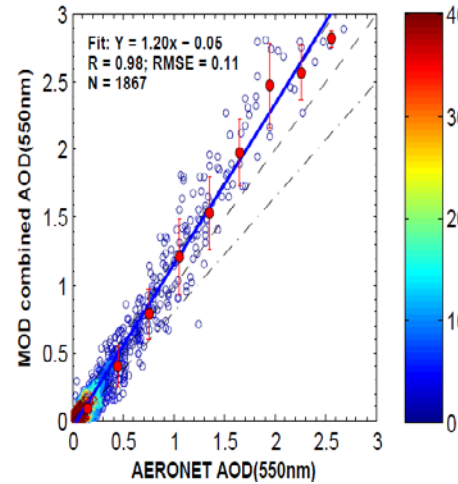


Fig. 3. MOD combined vs AERONET data from 2007 to 2010.

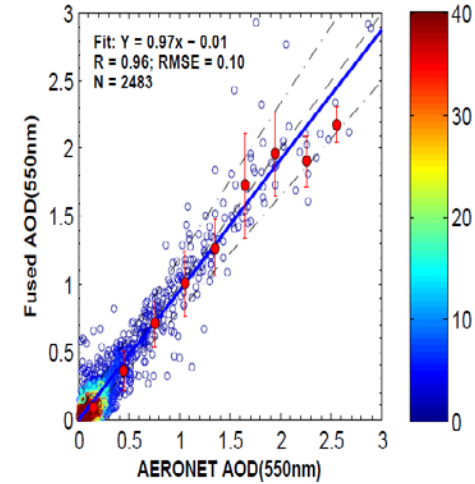


Fig. 4. MOD fused vs AERONET data from 2007 to 2010.

After comparing the original data, combined data and fused data with the AERONET observations, we find that the **RMSE of the fused image is smaller than all other data**, and **its correlation coefficient is better than MOD DB AOD**, so the data quality is improved. Meanwhile, the proportion of the fused AOD image with valid value is larger than any of the original products and MOD combined AOD. Thus, the fusion increases the spatial coverage.