

# GEOS Neural Network Retrieval for AOD Data Assimilation

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# **Comparison of MODIS DT & DB AOD Retrievals**



- Biases between datasets can propagate in the model forecast and lead to artificial time variability.
- The AOD data assimilation problem requires a homogenized AOD observing system across different platforms



# **Empirical Retrievals**



- f is a continuous function that maps S to G
- Represent *f* with a <u>mathematical function</u> that contains a set of empirical parameters, A
- A are determined from a training dataset of pairs of *G* and *S* <u>observations</u>.
- Training empirically captures physical relationships and tunes away calibration issues.





- F is a <u>physical model</u> derived from first principles (e.g. radiative transfer model)
- **F** is not easily inverted
- The objective of the retrieval algorithm is to search for a G\* that minimizes ||S – F(G)||
- Product quality affected by calibration issues.





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# **Observations**

### Satellite Sensor Observation:

### MODIS MOD04 /MYD04 Level 2 Reflectance

- Cloud masked, quality controlled, 10 km data
- Deep Blue Land
  - □ 3 channels over bright surfaces
  - 412 nm, 470 nm, and 670 nm

Dark Target Land

- 9 channels over dark surfaces
- □ 412-2100 nm

Geophysical Parameter of Interest:

### 440, 470, 550, 670, 870 nm AOD

- Aerosol Robotic Network (AERONET) observations of AOD
  - □ Global network of sunphotometers
  - 15 minute sampling
  - Low uncertainty (±0.01)



- Dark Target Ocean
  - □ 7 channels over ocean
  - □ 470-2100 nm





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### **MODIS-AERONET Data Pairs**



### □ 20 years of data (2000-2021)

### Additional Data Screening

- Outlier removal
- □ Cloud Fraction < 0.7
- Used MERRA-2 to "balance" the dataset by aerosol type
  - Dust
  - Smoke (Black Carbon + Organic Carbon)
  - Sea Salt
  - Sulfate





# **GEOS NNR for AOD**

MODIS OBSERVATION

**AUXILIARY DATA** 











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870 nm AOD

### **Testing at Specific Sites**

470 nm AOD

440 nm AOD

1.2 0.7 NNR • DT-0 DT-0 DT-0 • DT-O ٠ 0.6 NNR NNR NNR NNR 1.0 -0.8 1.0 0.6 0.5 0.8 0.5 0.8 0.6 0.4 AOD AOD 0.4 OV AOD Retrieved , 0 ਨੂ 0.6 σ σ ed Retrieve c.0 Retriev Retriev Retrie 0.4 0 Orig RMSE=0.09 Orig RMSE=0.09 0.2 Orig RMSE=0.09 Orig RMSE=0.08 NNR RMSE=0.05 NNR RMSE=0.05 NNR RMSE=0.04 NNR RMSE=0.05 NNR RMSE=0.03 0.2 0.2 0.2 0.1 0.0 0.0 0.0 0.0 0.0 0.6 0.8 1.0 0.2 0.4 0.5 0.3 0.0 0.2 0.4 1.2 0.0 0.2 0.4 0.6 0.8 1.0 0.0 0.2 0.4 0.6 0.8 0.0 0.1 0.3 0.6 0.7 0.0 0.1 0.2 0.4 0.5 0.6 AERONET AOD AERONET AOD AERONET AOD AERONET AOD AERONET AOD Terra DT-LAND Beijing-CAMS 440 nm AOD 470 nm AOD 550 nm AOD 660 nm AOD 870 nm AOD 3.0 NNR • DT-L • DT-L • DT-L NNR ٠ • 1.2 NNR NNR NNR 1.75 2.5 2.5 2.0 1.0 1.50 2.0 2.0 00 <sup>1.25</sup> Retrieved AOD 1.2 40D 40D AOD 00 I.5 Retrieved , 9.0 9 1.00 8 1.5 0 Retriev 0.72 Retrie 1.0 Retri 1 ( 1.0 0.4 Orig RMSE=0.20 NNR RMSE=0.14 Orig RMSE=0.23 Orig RMSE=0.19 0.50 NNR RMSE=0.18 NNR RMSE=0.17 NNR RMSE=0.12 NNR RMSE=0.09 0.5 0.5 0 0.25 0.0 0 0.0 0.00 0.0 0.5 1.5 2.0 2.5 3.0 0.5 1.0 1.5 2.0 2.5 0.5 1.0 1.5 2.0 0.00 0.25 0.50 0.75 1.00 1.25 1.50 1.75 0.2 0.4 0.6 0.8 1.0 0.0 1.0 0.0 0.0 1.2 0.0 AERONET AOD AERONET AOD AERONET AOD AERONET AOD AERONET AOD

Terra DT-OCEAN Shirahama

550 nm AOD

660 nm AOD



# **AOD Spectral Dependence**



# **AOD Spectral Dependence**



**SODDARD** 

# **AOD Spectral Dependence**





# **AOD Spectral Dependence**









NNR ALL AOD 20160801



MOD04 ALL AOD 20191201







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# **Transition to DB & DT Applied to VIIRS**





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# **Summary and Outlook**

- □ The NNR provides a way to homogenize the AOD observing system for data assimilation
- A NNR for VIIRS, GOES-16 and AHI-8 will be developed using the same methodology as the MODIS-NNR
- JEDI-based aerosol analysis system supports new multi-wavelength NNR AOD
  - may expand wavelength range to 340–1020



Target Aerosol Observing System in GEOS

# **Summary and Outlook**

- Innovate the training methodology to give probabilistic predictions
  - e.g. ensemble learning, gaussian process learning, deep evidential regression
  - Benefits: provides quantitative uncertainties of the predictions that can be used in the aerosol assimilation



