

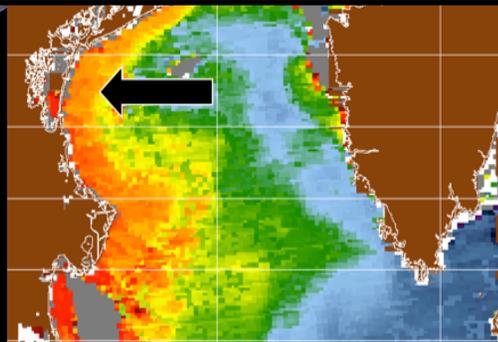
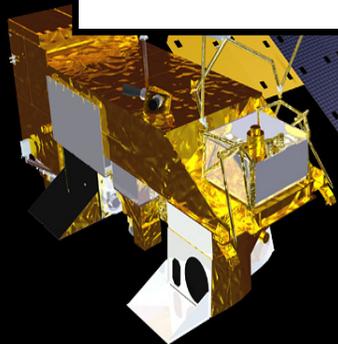


**NURC** - Partnering for Maritime Innovation



# Application of AERONET-OC Data in Satellite Product match-ups: Covariance and Uncertainty index

A. Alvarez, G. Pennucci, C. Trees and G. Fargion



AERONET-OC: 1<sup>st</sup> Workshop Overview and Future Developments, Feb 23-24, 2011

Greenbelt, Maryland (USA)

NATO UNCLASSIFIED



## Goals:



Our main goals are to 1) **merge remote sensing data with *in situ* measurements to increase the quality of satellite-derived products,** and 2) **estimate satellite derived product uncertainties based on the number of in situ measurements**

- Develop techniques to evaluate the covariance in SST and ocean color images;
- Develop a methodology to merge satellite images with *in situ* data to improve the satellite product;
- Develop software to ingest single or time series images to compute covariance, to merge satellite and *in situ* data and to determine which is the best placement for *in situ* measurements
- Estimate product retrieval uncertainties to estimate optimal *in situ* sampling strategies (one basis for a defined Uncertainty Index-UI).



## Summary:



- ✓ The general procedure;
- ✓ Covariance evaluation (spatial-temporal and spatial domains);
- ✓ Statistical model definition (spatial domain);
- ✓ Merging technique;
- ✓ Data ingestion criteria and cloud cover analysis;
- ✓ Example: covariance of MODIS time-series;
- ✓ Demo of the GUI (Version 2);
- ✓ Conclusion and work in progress.



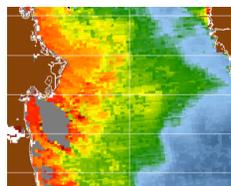
# Procedure Flow Diagram:



## SPATIAL DOMAIN

## SPATIAL-TEMPORAL DOMAIN

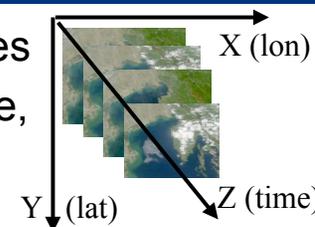
Ingestion of a satellite image  
(geotiff, tiff, hadf4, hadf5,  
grid or MATLAB format)



MODIS

STEP 1

Ingestion of a time-series  
(3D-cube with Longitude,  
Latitude and time info)



Selection of the desired  
geographic area

STEP 2

(with **low cloud  
percentage**)

Spatial Covariance Evaluation

STEP 3

Spatial-temporal Covariance Evaluation



Merging with *in situ* data  
(such as AERONET,  
CTD, HyperPRO, and gliders), if they are available (~same time

STEP 4





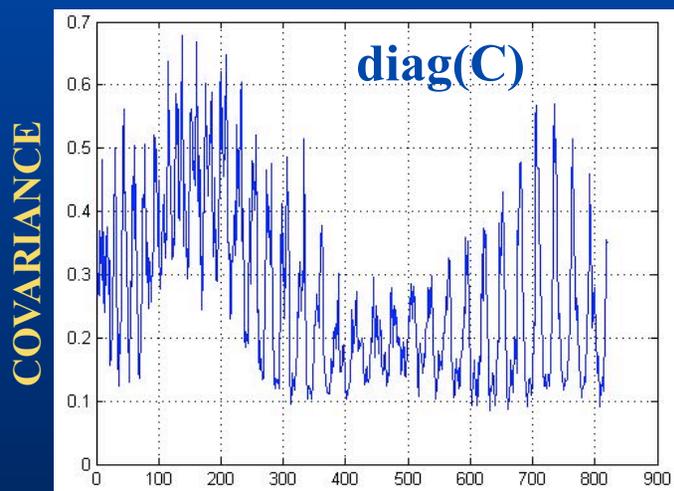
# Covariance Evaluation

## SPATIAL-TEMPORAL DOMAIN

If a time-series is available ( $\{\psi_i(\vec{x})\}_{i=1}^W$ )



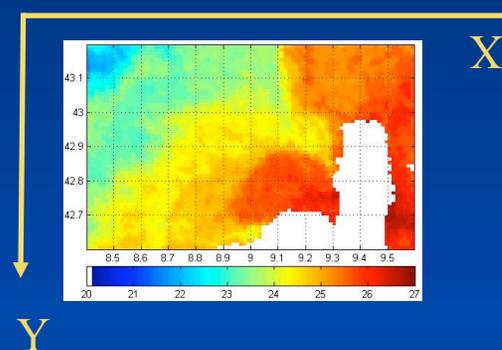
$$T_{mn} \longrightarrow C_{nn} = T^T \cdot T$$



TIME SERIES

## SPATIAL DOMAIN

Just an image is available:  $\psi(\vec{x})$



How to make statistic without temporal variations... ?

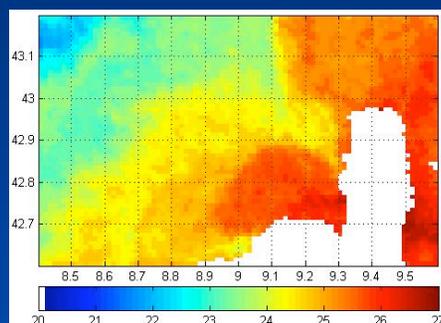


To properly evaluate the covariance (without the time domain) we have to define a statistical model.

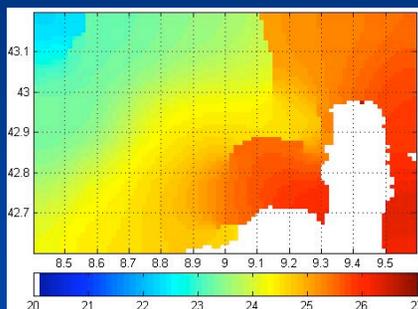


# Statistical model definition

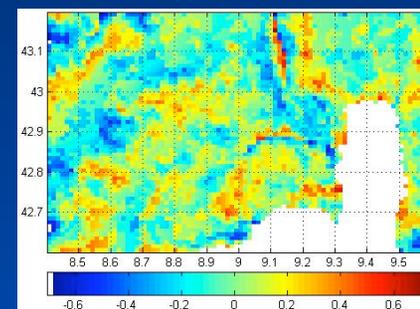
The image can be decomposed into: spatially varying mean + random Gaussian residual using an anisotropic diffusion operator that **preserves front structures**.



=



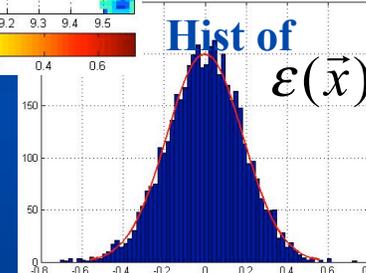
+



**Oceanographic field  $\psi(\vec{x})$**   
**(SST original image)**

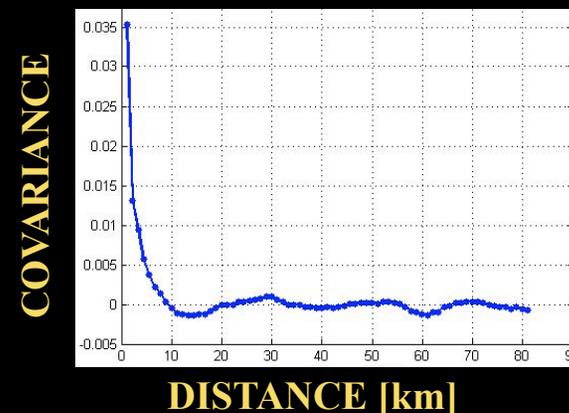
**Mean field,  $M\psi(\vec{x})$**

**Residual field  $\varepsilon(\vec{x})$**



We assume that covariance (C) is a function of the relative distance between 2 points.

C is the covariogram of the residual field :





# The merging technique:

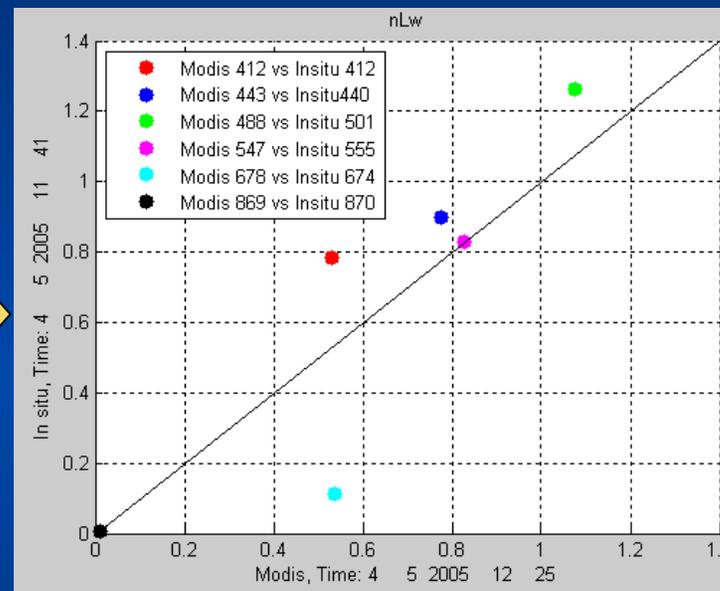
Assuming that *in situ* data are available:



AERONET data

**User responsible for the QC & QA of in situ data**

**EXCLUSION CRITERIA:**  
Max Solar Zenith Angle:  $70^\circ$   
Wind speed  $\leq 12 \text{ ms}^{-1}$   
Max time difference from Satellite data =  $\pm 1$  hours



If the available *in situ* data are consistent, a new image merging AERONET and satellite data can be retrieved maximizing the probability distribution (that depends on the retrieved covariance).



# Merging technique (details)



Once the covariance (C) has been obtained, a new field merging *in situ* and satellite data is obtained maximizing the probability distribution:

$$P(\psi_k) \propto \exp(\psi_{obs} - H\psi_k)^T \Sigma_{obs}^{-1} (\psi_{obs} - H\psi_k) - (\psi_k - \bar{\psi})^T C_{\epsilon}^{-1} (\psi_k - \bar{\psi})$$

Observation vector      Observation matrix      Estimated pixel values      Observation error matrix

The solution of the following equation:

$$\psi_{MERGED} = \arg \min (\psi_{obs} - H\psi_k)^T \Sigma_{obs}^{-1} (\psi_{obs} - H\psi_k) - (\psi_k - \bar{\psi})^T C_{\epsilon}^{-1} (\psi_k - \bar{\psi})$$

represents the **merged image**: the best estimation compatible with the available information.



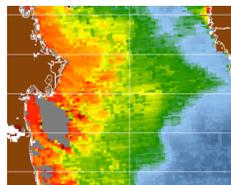
# The procedure:



## SPATIAL DOMAIN

## SPATIAL-TEMPORAL DOMAIN

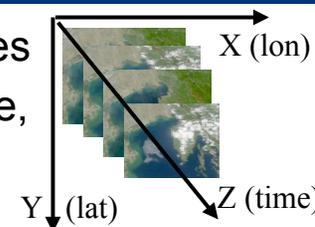
Ingestion of a satellite image  
(geotiff, tiff, hadf4, hadf5,  
grid or MATLAB format)



MODIS

STEP 1

Ingestion of a time-series  
(3D-cube with Longitude,  
Latitude and time info)



Data ingestion criteria (user required)

Selection of the desired  
geographic area

STEP 2

(with **low cloud  
percentage**)

Spatial Covariance Evaluation

STEP 3

Spatial-temporal Covariance Evaluation



Merging with *in situ* data  
(such as AERONET,  
CTD, HyperPRO, and gliders), if they are available (~same time)

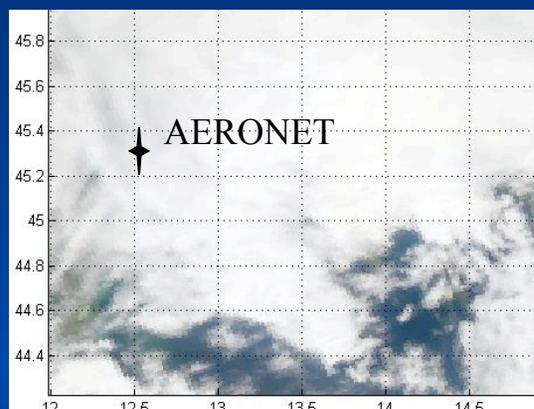
STEP 4





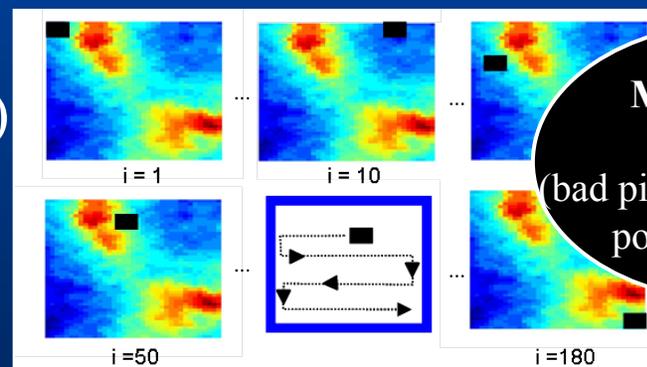
# Data ingestion criteria

It's important to note that our study is directly related to the **quality of satellite data**: results are directly related to the amount of information.



**EVALUATION OF THE METHOD TO RESPECT CLOUDS/NOISE/BAD PIXELS**

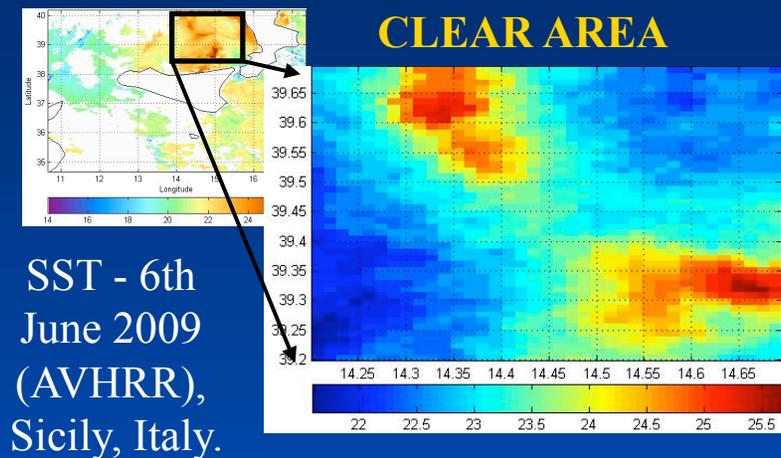
The covariance field was evaluated as a function cloud cover (bad pixels) to define a minimum amount of information for implementation of the model.



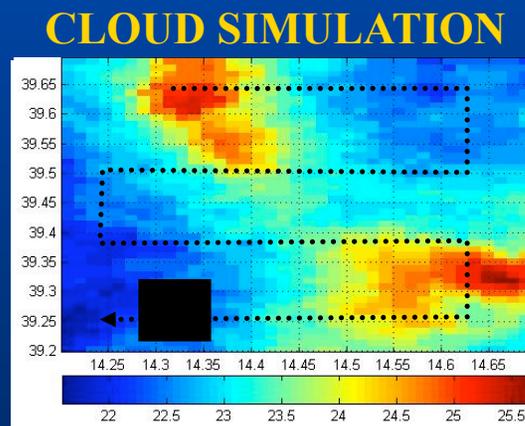
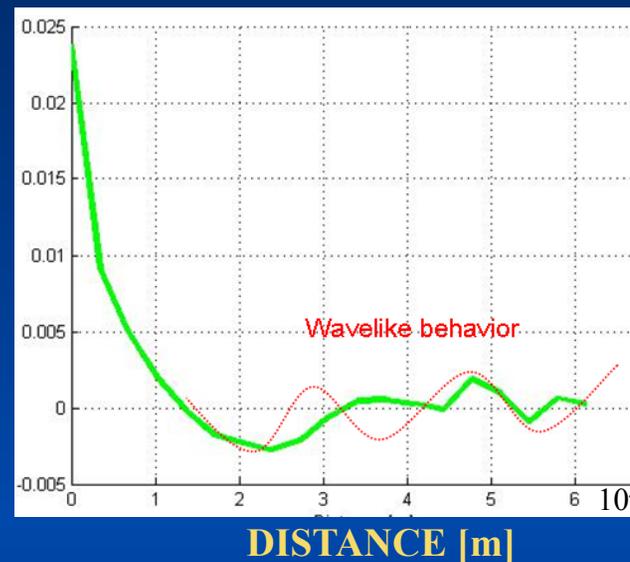
**Monte Carlo Simulation**  
(bad pixels with different position & size)



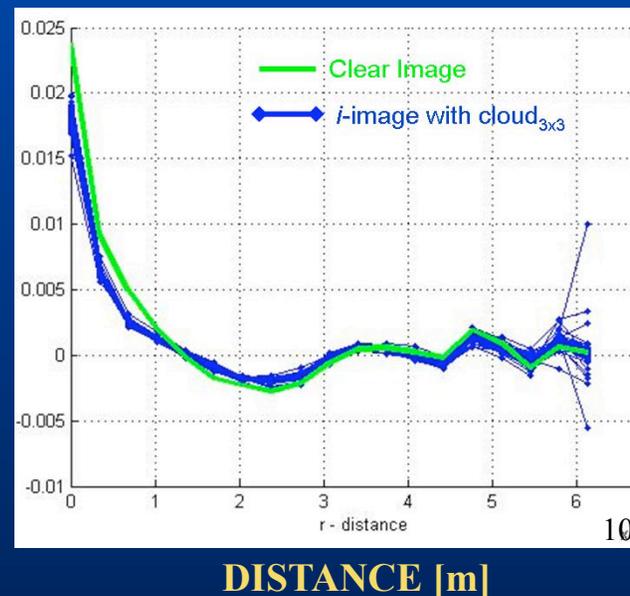
# Covariance as function of clouds (1/2)



COVARIANCE



COVARIANCE





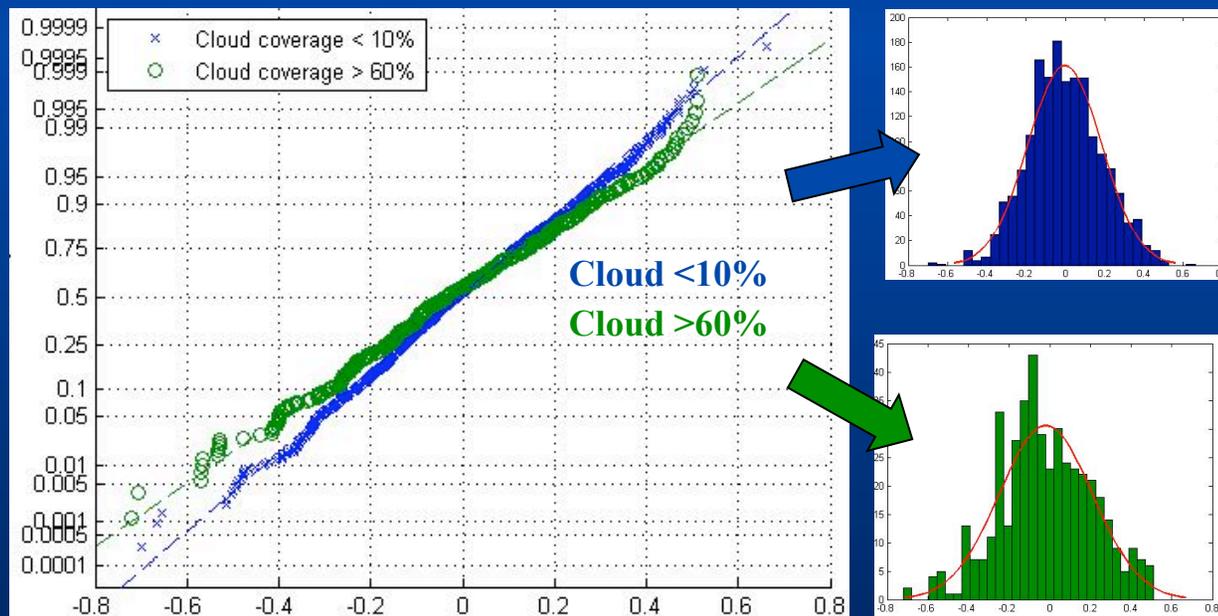
## Covariance as function of clouds (2/2)



The results show that the covariance analysis is directly dependent on cloud size/position. Clouds cause a loss of information, which means:

1. Decreased correlation length;
2. Decreased goodness of the realization (Normal distribution)

### PROBABILITY PLOT FOR NORMAL DISTRIBUTION (data vs prob)



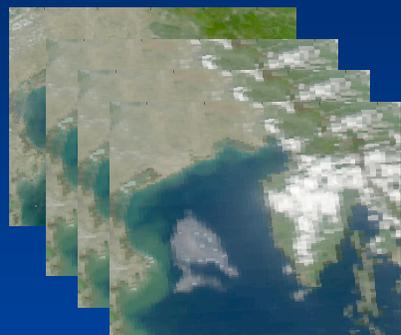
### QUALITY INDICES DEFINITION:

**Index 1** takes into account of the # of structures (~amount of information);

**Index 2** takes into account Of the goodness of the residual field realization.



# An example of MODIS time series (1/2)



MODIS time-series images.  
Month : May  
Years : from 2005 to 2009

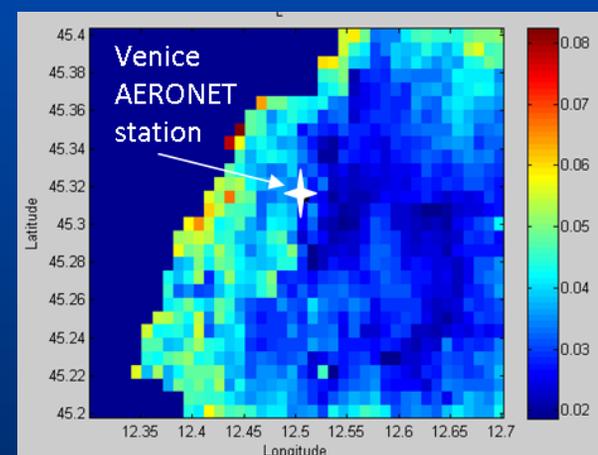
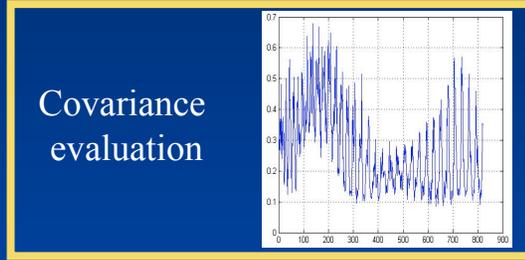


AERONET data  
Integration (water  
leaving radiance)

205  
images

Automatic + user  
required procedures  
to select "clear"  
images (data quality)

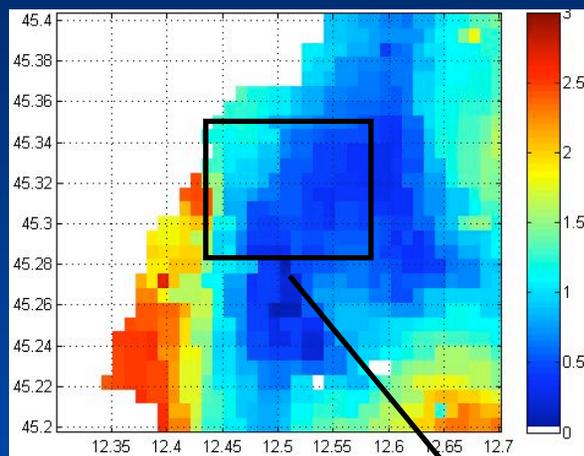
15 images



Satellite Error Estimation  
(uncertainty)

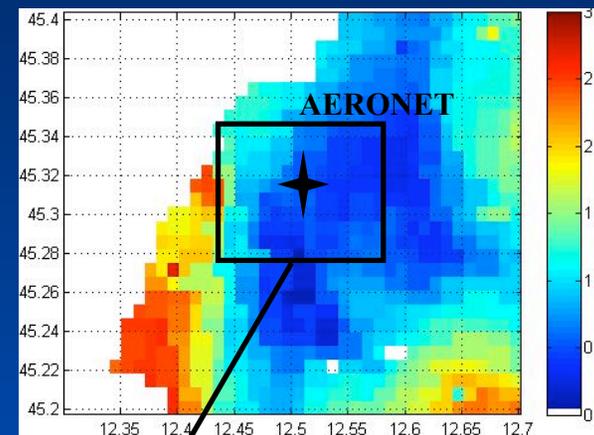


# An example of MODIS time series (2/2)



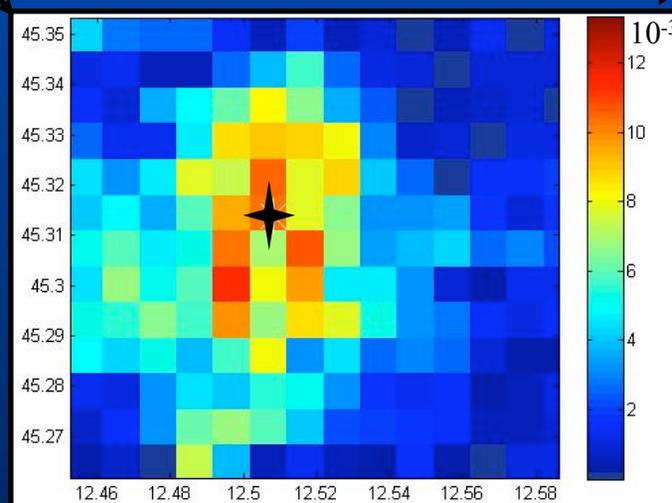
Water Leaving radiance (547 nm) from MODIS (4 May 05, h12.25)

**MERGING**  
Using C and AERONET measurement:  
Lat = 45.3139;  
Lon = 12.5083  
Time = 4 May 2005, h12.21

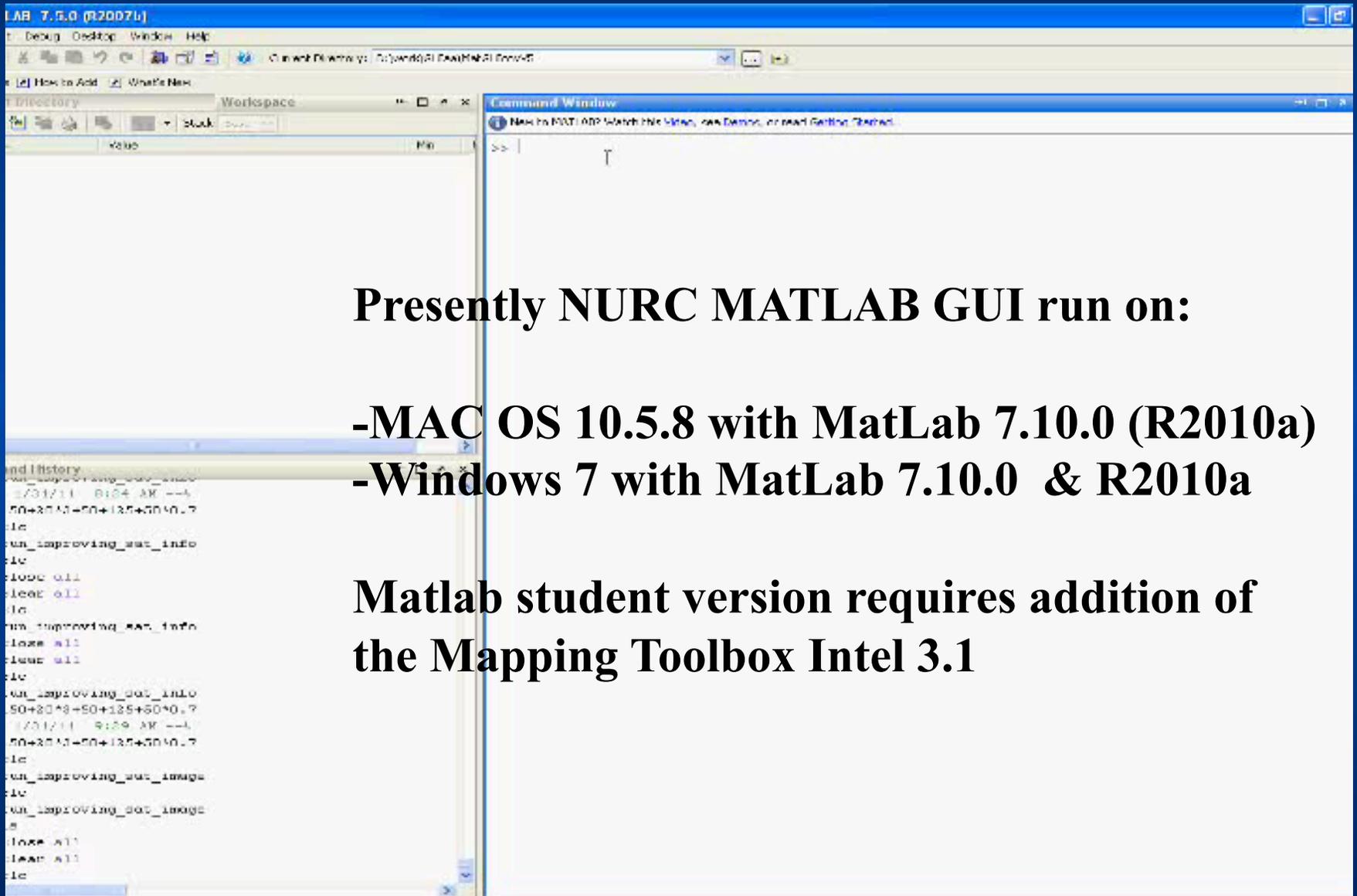


Merged Image

Difference between original and merged



# GUI demo



**Presently NURC MATLAB GUI run on:**

- MAC OS 10.5.8 with MatLab 7.10.0 (R2010a)**
- Windows 7 with MatLab 7.10.0 & R2010a**

**Matlab student version requires addition of  
the Mapping Toolbox Intel 3.1**



## Conclusion:

The procedure developed for the project Calibration/Validation (cal/val) Plan for Oceans Environmental Data Records (ERDS) for the Visible Infrared Spectrometer (VIIRS) have been described.

In particular:

- ✓ Covariance computation procedures;
- ✓ Evaluation of the covariance dependence on clouds size/position;
- ✓ Description of the merging procedure;
- ✓ Demo of the software developed (GUI version 3 – Jan 2011).

## Work in progress:

- Integration of the spatial-temporal procedure in the GUI (Version 4 – April 2011);
- Utilization of VIIRS proxy data to estimate optimal *in situ* sampling strategies (one basis for a defined Uncertainty Index-UI).