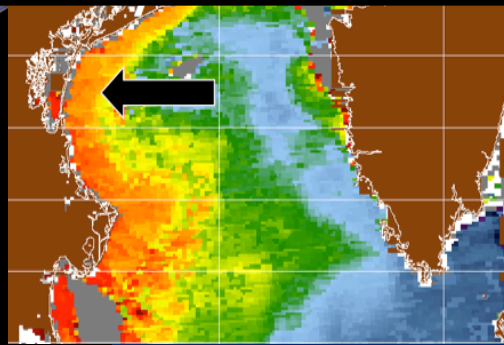
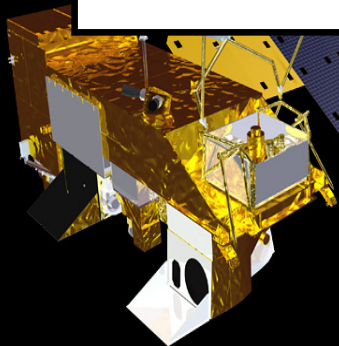




Application of AERONET-OC Data in Satellite Product match-ups: Covariance and Uncertainty index

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





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.

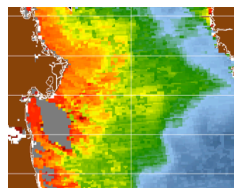


Procedure Flow Diagram:

SPATIAL DOMAIN

SPATIAL-TEMPORAL DOMAIN

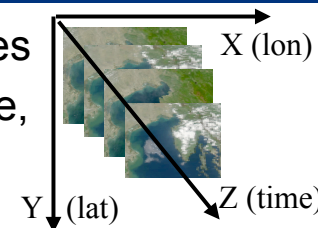
Ingestion of a satellite image
(geotiff, tiff, hadf4, hadf5,
grid or MATLAB format)



MODIS

STEP 1

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



Selection of the desired
geographic area

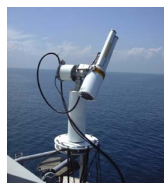
STEP 2

(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





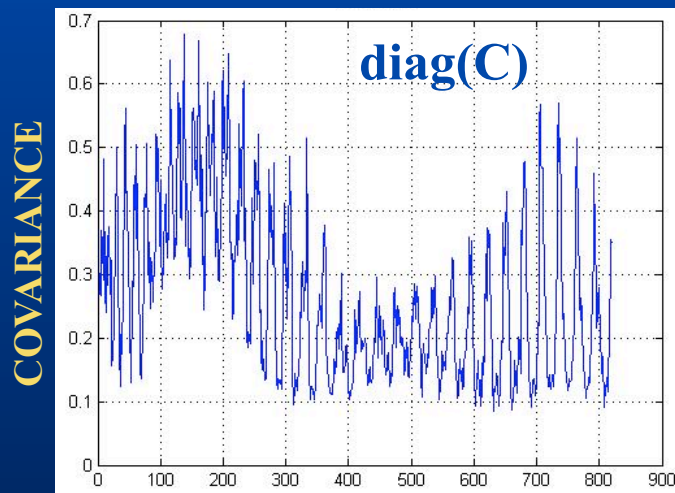
Covariance Evaluation

SPATIAL-TEMPORAL DOMAIN

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



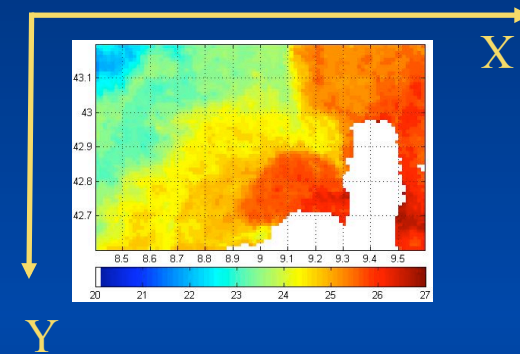
$$T_{mn} \longrightarrow C_{nn} = T^T \cdot T$$



TIME SERIES

SPATIAL DOMAIN

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



How to make statistic without temporal variations...

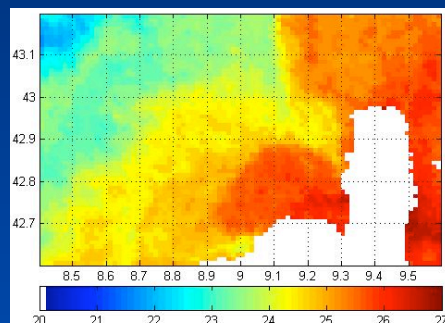


To properly evaluate the covariance (without the time domain) we have to define a statistical model.

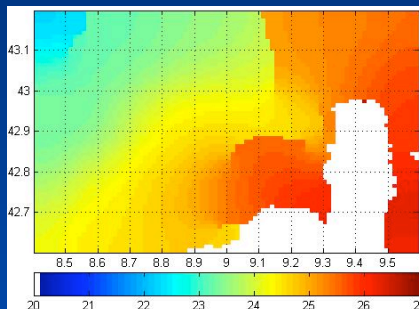


Statistical model definition

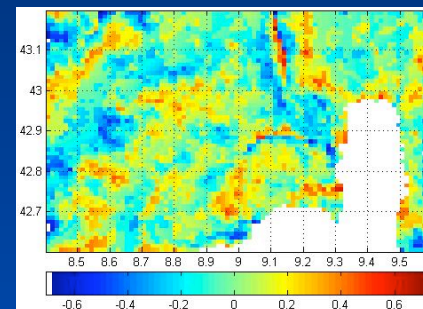
The image can be decomposed into: spatially varying mean + random Gaussian residual using an anisotropic diffusion operator that **preserves front structures**.



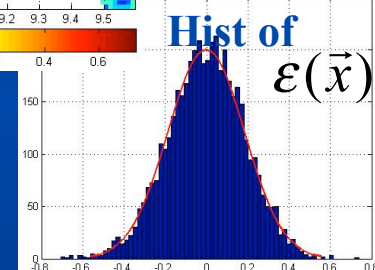
Oceanographic field $\psi(\vec{x})$
(SST original image)



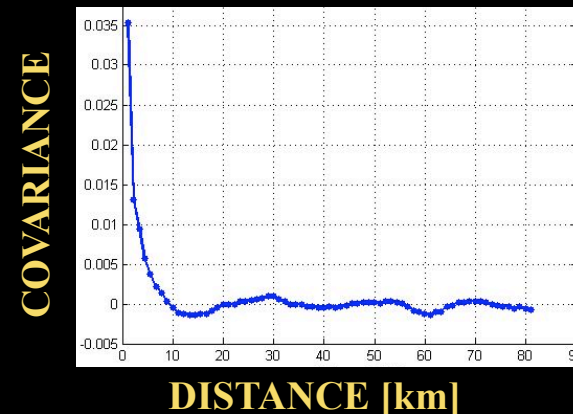
Mean field, $M\psi(\vec{x})$



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:

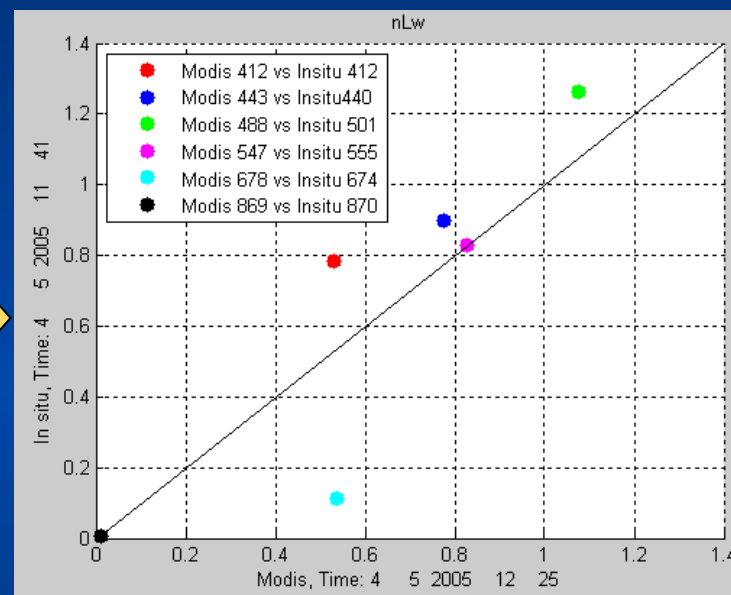


AERONET data

User responsible for the QC & QA of in situ data

EXCLUSION CRITERIA:

Max Solar Zenith Angle: 70°
Wind speed $\leq 12 \text{ ms}^{-1}$
Max time difference from
Satellite data = ± 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:

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

Diagram illustrating the components of the probability distribution equation:

- Observation vector (ψ_{obs})
- Observation matrix (H)
- Estimated pixel values (ψ_k)
- Observation error matrix (Σ_{obs}^{-1})
- Covariance matrix (C_{ϵ}^{-1})

The solution of the following equation:

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

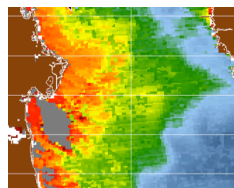
represents the **merged image**: the best estimation compatible with the available information.



The procedure:

SPATIAL DOMAIN

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

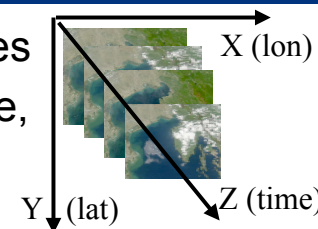


MODIS

STEP 1

SPATIAL-TEMPORAL DOMAIN

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



Data ingestion criteria (user required)

Selection of the desired
geographic area

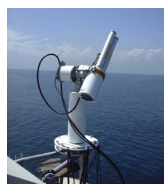
STEP 2

(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

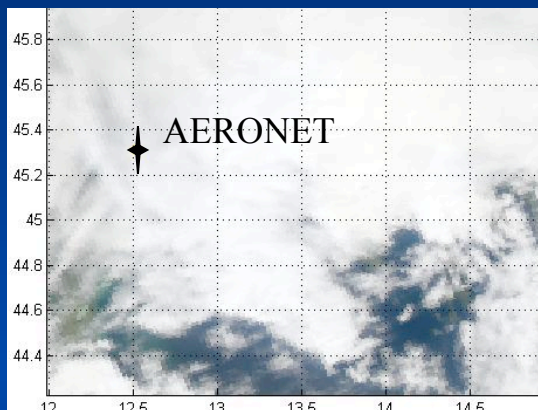
STEP 4





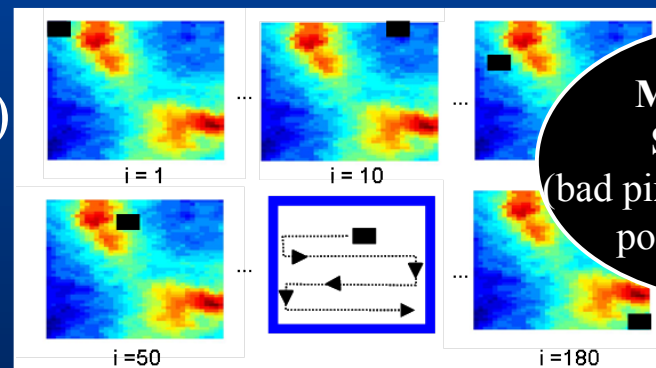
Data ingestion criteria

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

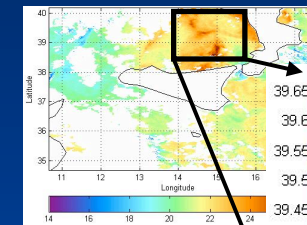
The covariance field was evaluated as a function cloud cover (bad pixels) to define a minimum amount of information for implementation of the model.



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

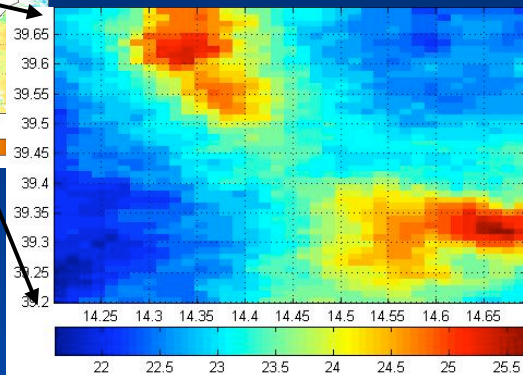


Covariance as function of clouds (1/2)

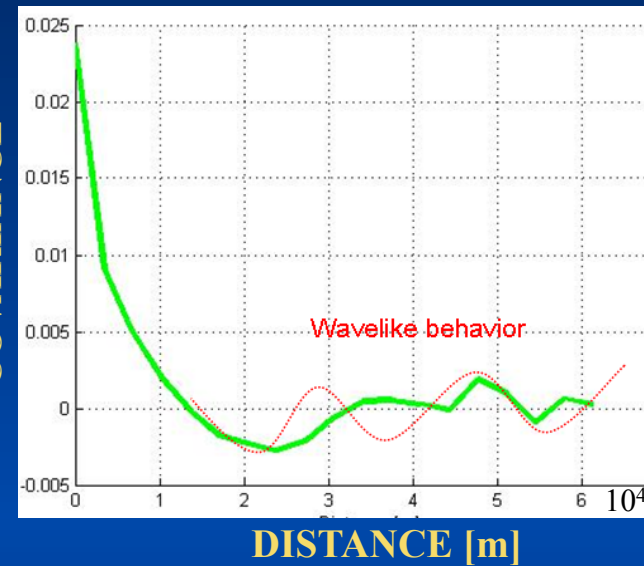


SST - 6th
June 2009
(AVHRR),
Sicily, Italy.

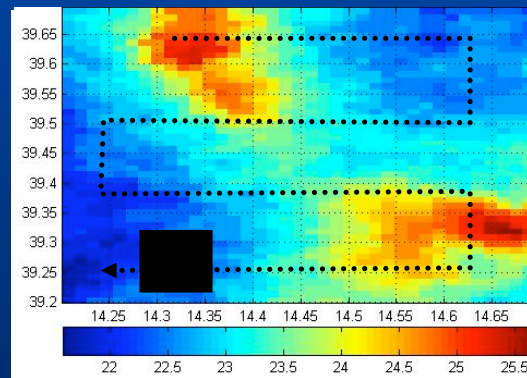
CLEAR AREA



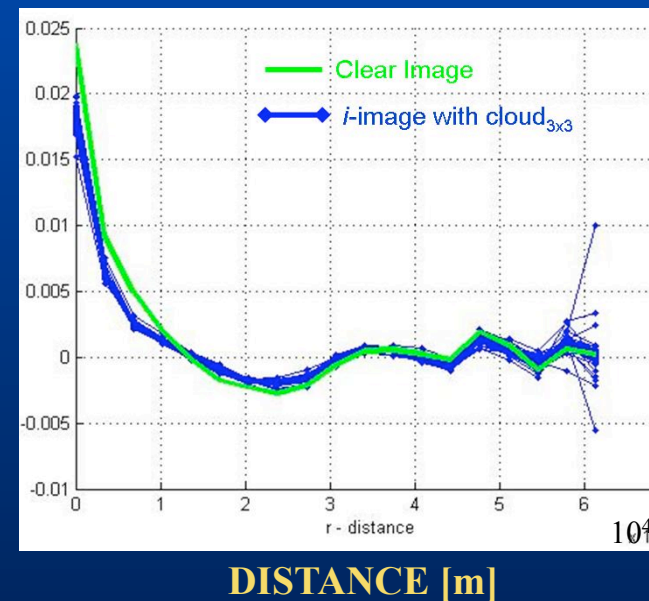
COVARIANCE



CLOUD SIMULATION



COVARIANCE



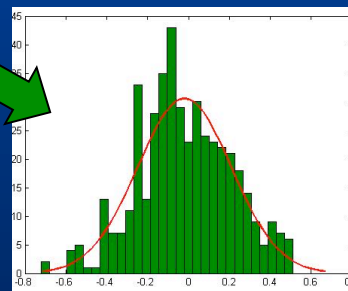
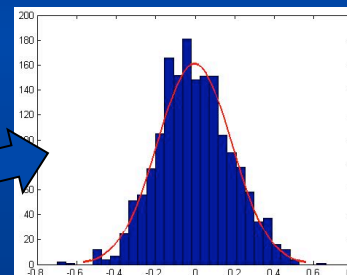
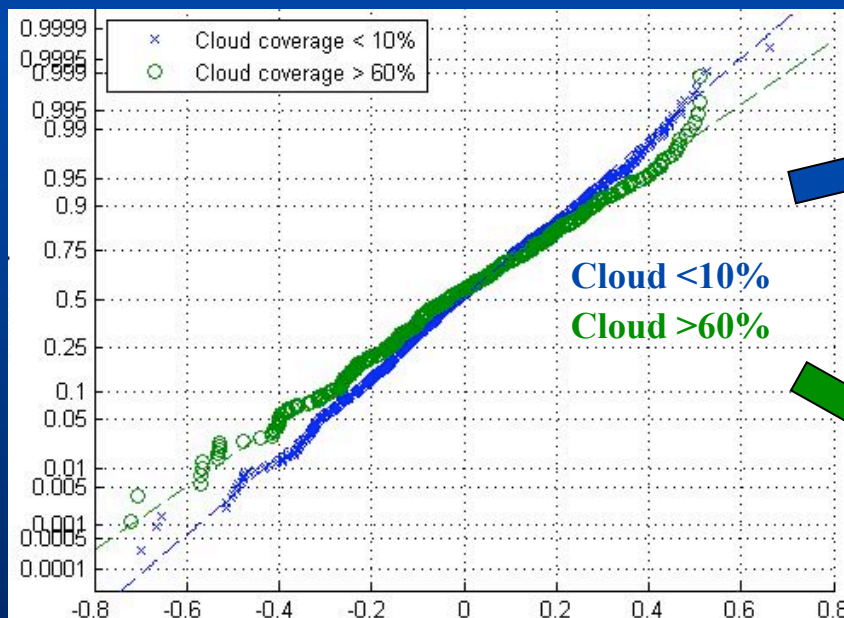


Covariance as function of clouds (2/2)

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)



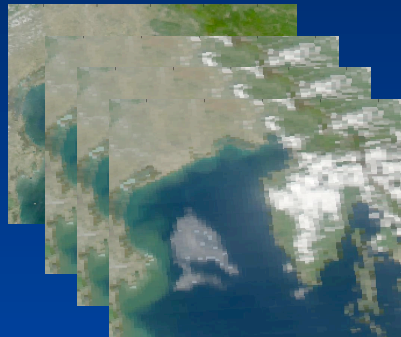
QUALITY INDICES DEFINITION:

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

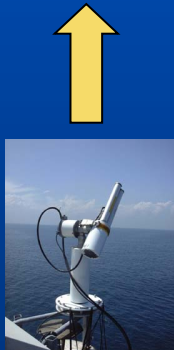
Index 2 takes into account Of the goodness of the residual field realization.



An example of MODIS time series (1/2)



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



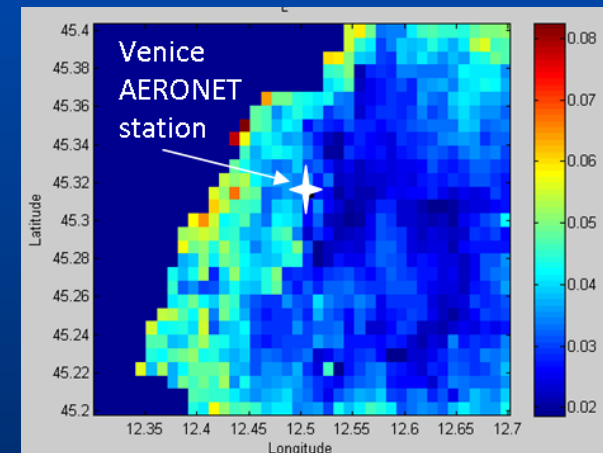
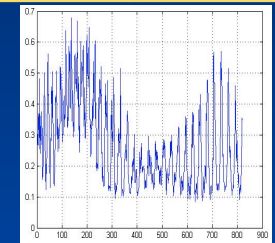
AERONET data
Integration (water
leaving radiance)

205
images

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

15 images

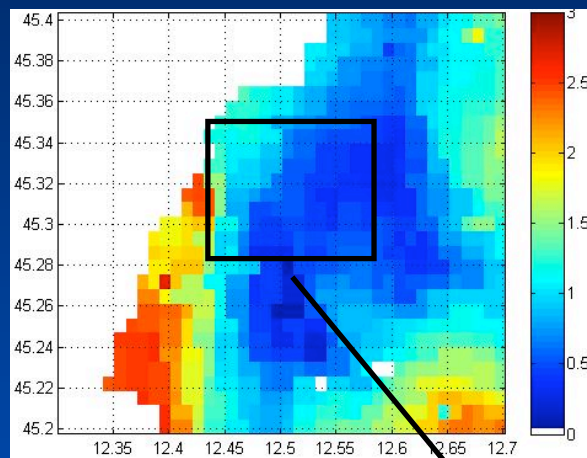
Covariance
evaluation



Satellite Error Estimation
(uncertainty)

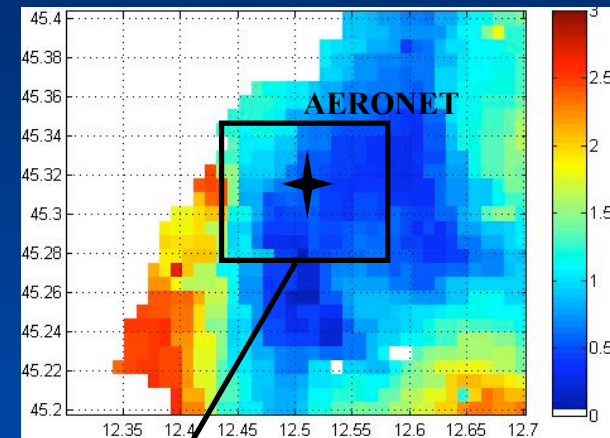


An example of MODIS time series (2/2)



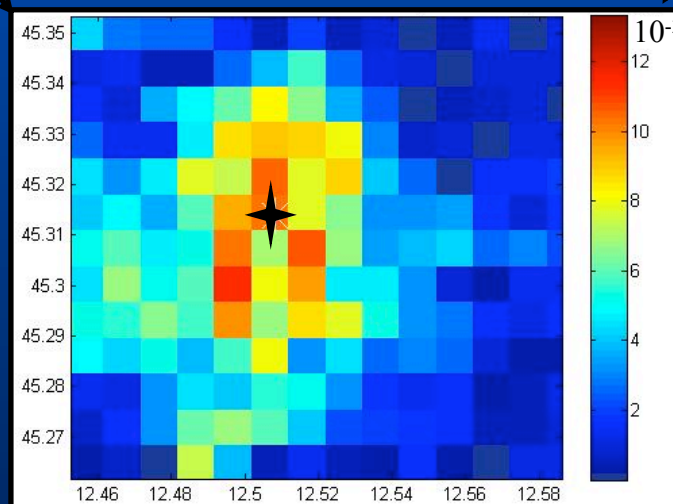
Water Leaving radiance (547 nm)
from MODIS (4 May 05, h12.25)

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

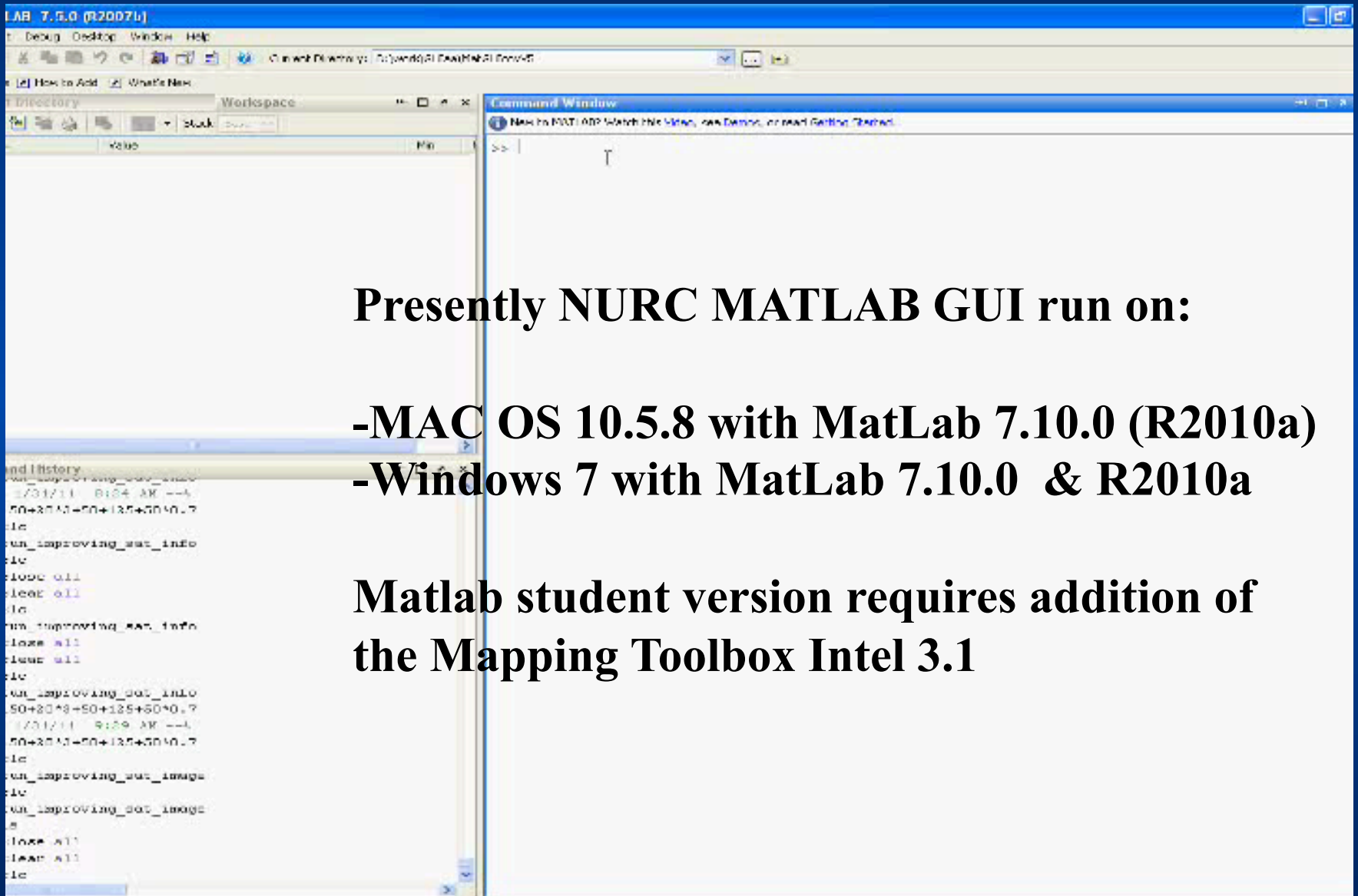


Merged Image

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).