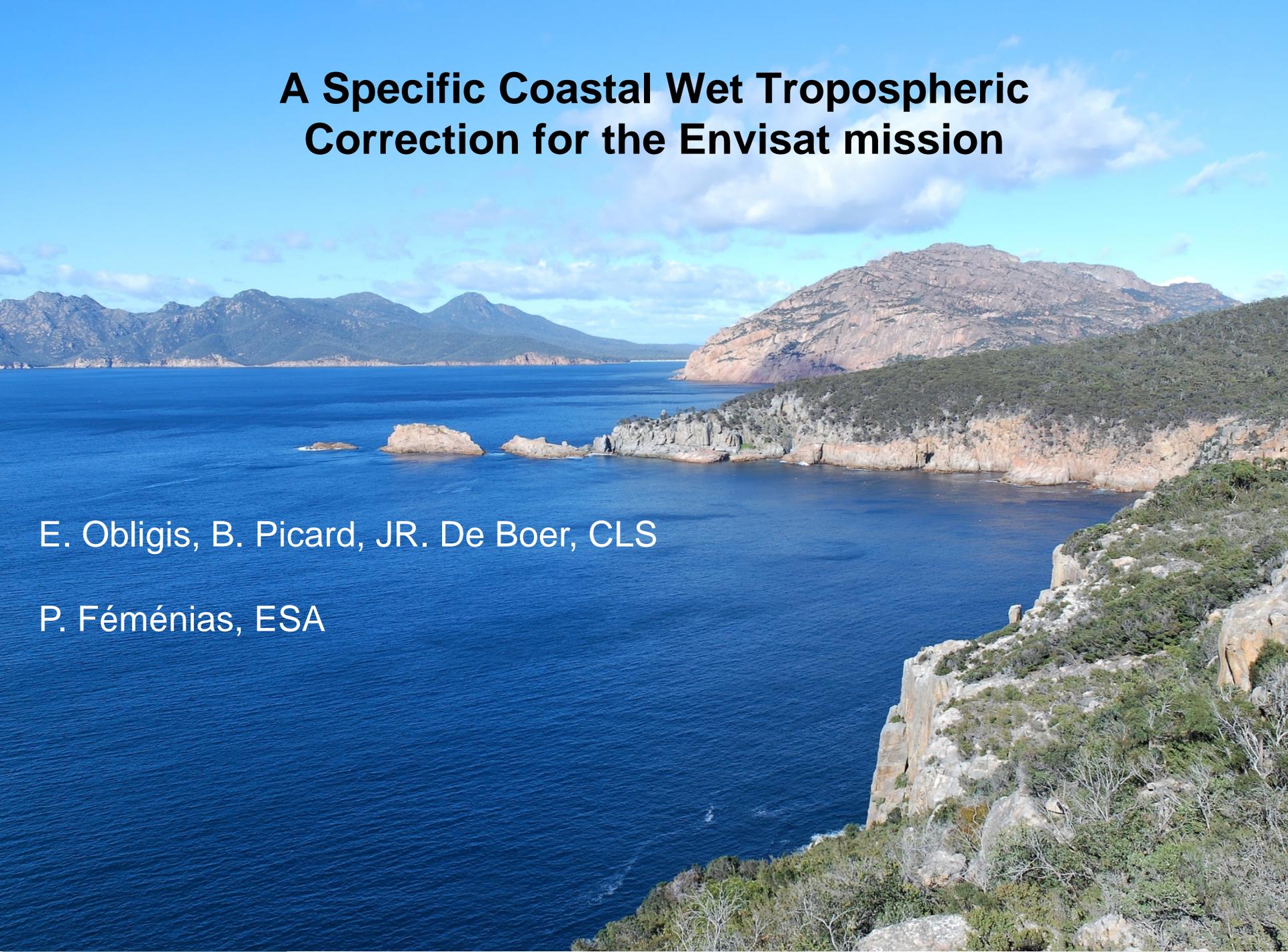


# A Specific Coastal Wet Tropospheric Correction for the Envisat mission

E. Obligis, B. Picard, JR. De Boer, CLS

P. Féménias, ESA



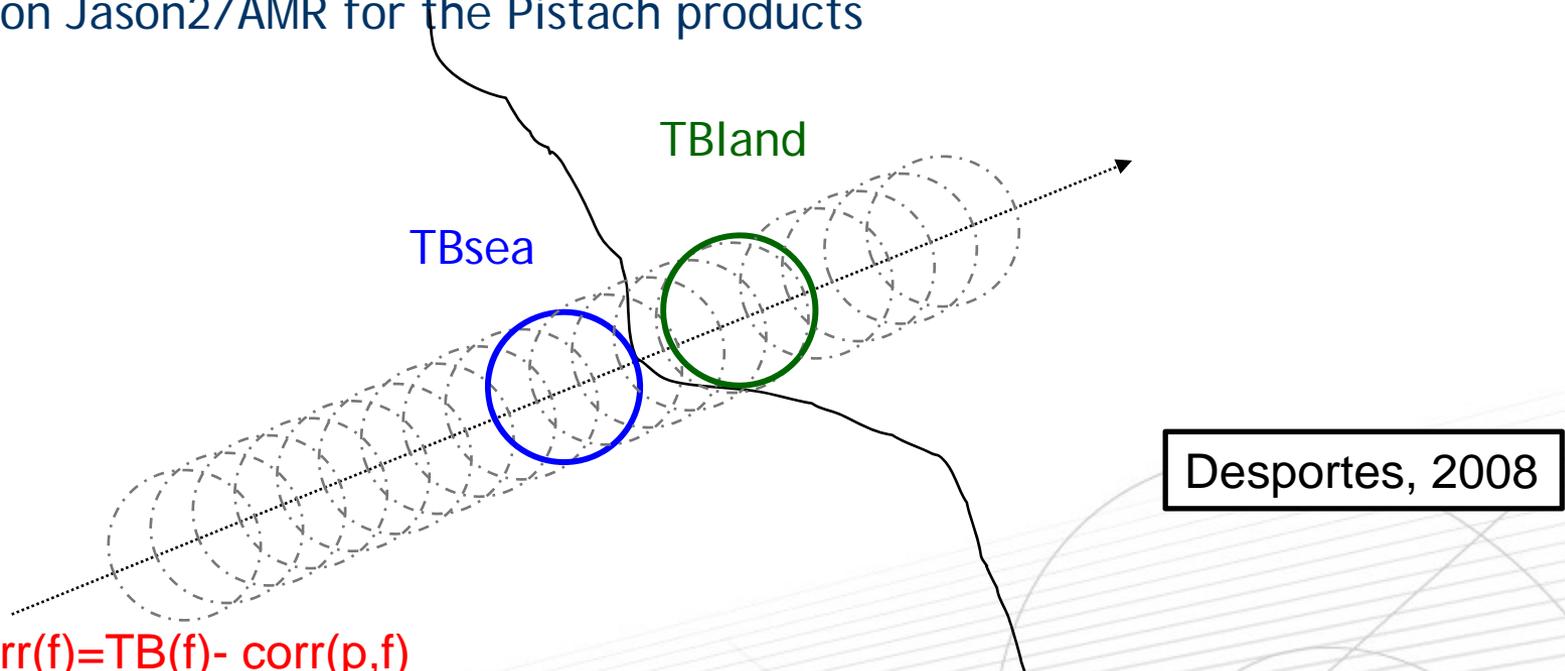
## Radiometric measurements

- Strongly contaminated by land: the same way as the other radiometers (SSM/I, TMI, AMSU...)
- Land emissivity nearly twice sea emissivity + more variable in space and time
- For a surface temperature of 300K, a 10% land contamination in the sea pixel will increase the TB by more than 10K → several centimeters !
- Classical retrieval algorithms developed assuming sea surface emissivity modeling are no more valid
- BUT only radiometer products can provide the required resolution to detect short scales SSH signals in coastal areas
- Alone or combined with other products (Mercier 2007, J. Fernandes 2011)



# Existing methods

- Correction of land contamination before application of the L2 ocean retrieval algorithm : Desportes et al, 2006
- Applied on Jason2/AMR for the Pistach products



$$TB\_corr(f) = TB(f) - corr(p, f)$$

$$corr(p, f) = [TBland(f) - TBsea(f)] \times p(f)$$

f: frequency of the 3 channels (18.7, 23.8 and 34 GHz for Jason 1-2)

p(f): proportion of land in the footprint taking into account the antenna patterns

TBland: closest TB with 100% of land

TBsea: closest TB with 100% of sea

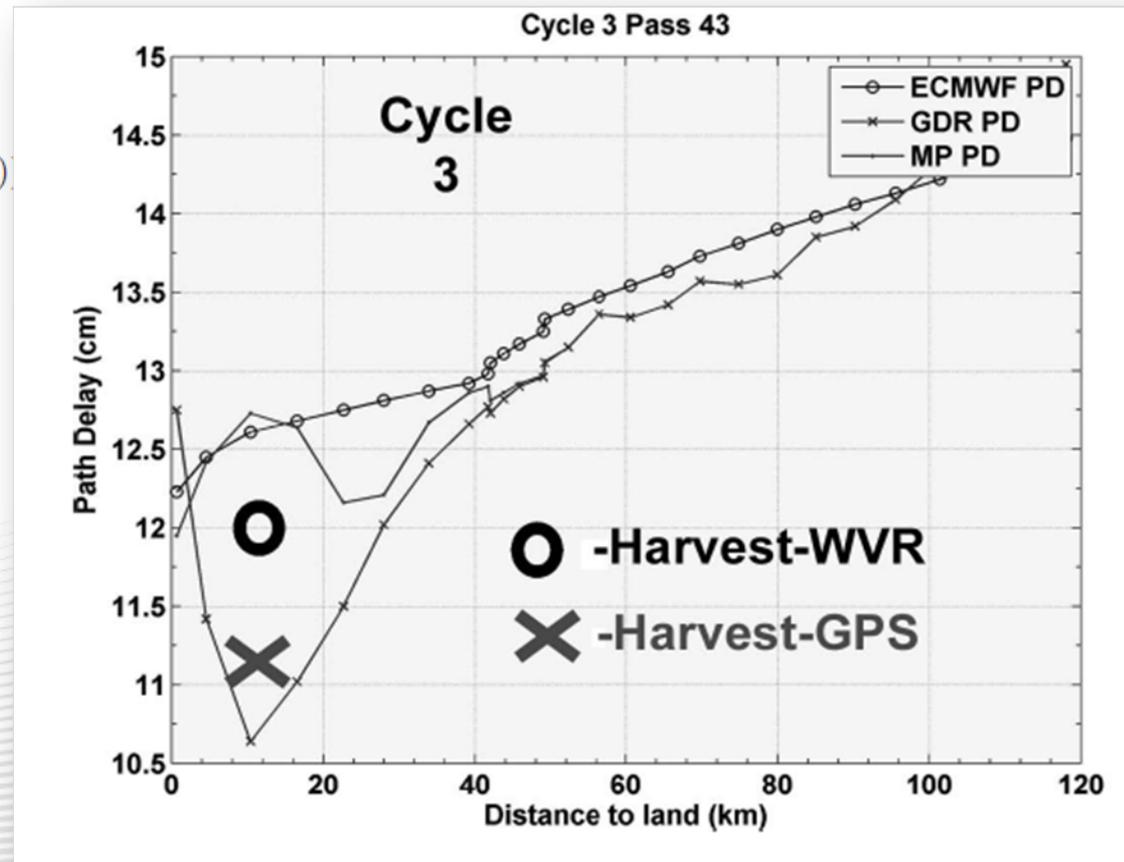
# Existing methods

- Land proportion used as external parameter in the L2 retrieval algorithm : **Brown 2010 et al**
- Applied in Jason2/AMR operational products

$$\begin{aligned}
 PD_{MP} &= c_0 (PD_0, L_F^{18.7}) \\
 &+ \sum_f c_f (PD_0, L_F^{18.7}) \log(280 - T_B(f))
 \end{aligned}$$

$$T_{MB}(f) = (1 - L_F(f))T_{Ocean}(f) + L_F(f)T_{Land}(f)$$

**Brown, 2010**



Combined MWR - ECMWF -GNSS wet tropo. corr  
through COASTALT ESA initiative (<http://www.coastalaltimetry.org/>)

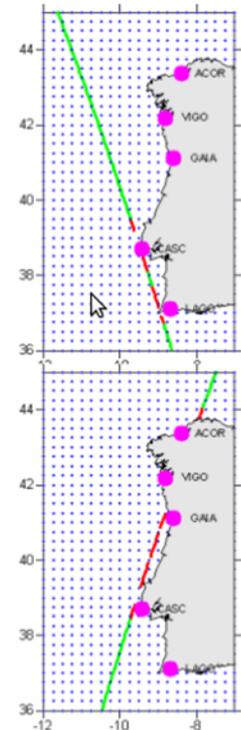
Fernandes, 2010

## Summary of the method

### GPD (GNSS-derived Path Delay)

Combines the following data sets  
(objective analysis):

- GNSS-derived zenith Zenith Total Delays (ZTD) at coastal GNSS stations
- Valid MWR measurements
- ZWD from a Numerical Weather Model, ECMWF (global grids  $0.25^\circ \times 0.25^\circ$ , every 6h)



# Proposed methodology for RA2-MWR

- RA-MWR : bi-frequency nadir radiometer : 23.8 GHz/36.5GHz
- What do we need ? [adapted from Brown 2010]
  - Measured brightness temperatures (mixed Land/Ocean)
  - Measured altimeter backscattering coefficient in Ku band (to take into account surface roughness)
  - Land proportion in the pixel at both frequencies

$dh = NN (TB_{23.8}, TB_{36.5}, \sigma_{0\_Ku}, land\_prop_{23.8}, land\_prop_{36.4})$

**C**oastal **N**eural **N**et algorithm

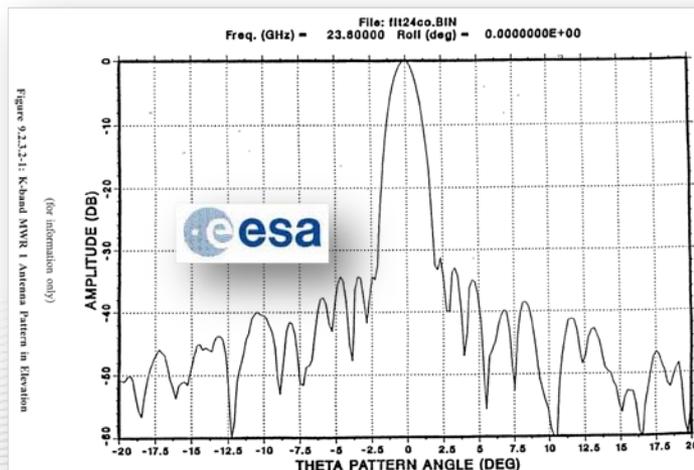
- Algorithm formulation
  - Building of the learning database
  - Learning of the neural net

# Building of the learning database

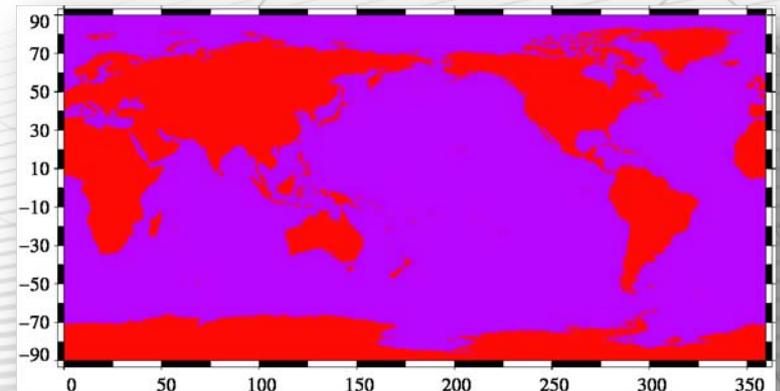
dh = NN (TB23.8, TB36.5,  $\sigma_0$ \_Ku, land\_prop23.8, land\_prop36.5)

land proportion

- Weighted mean of a 1/30° land sea mask by a sampling of Envisat MWR true antenna pattern

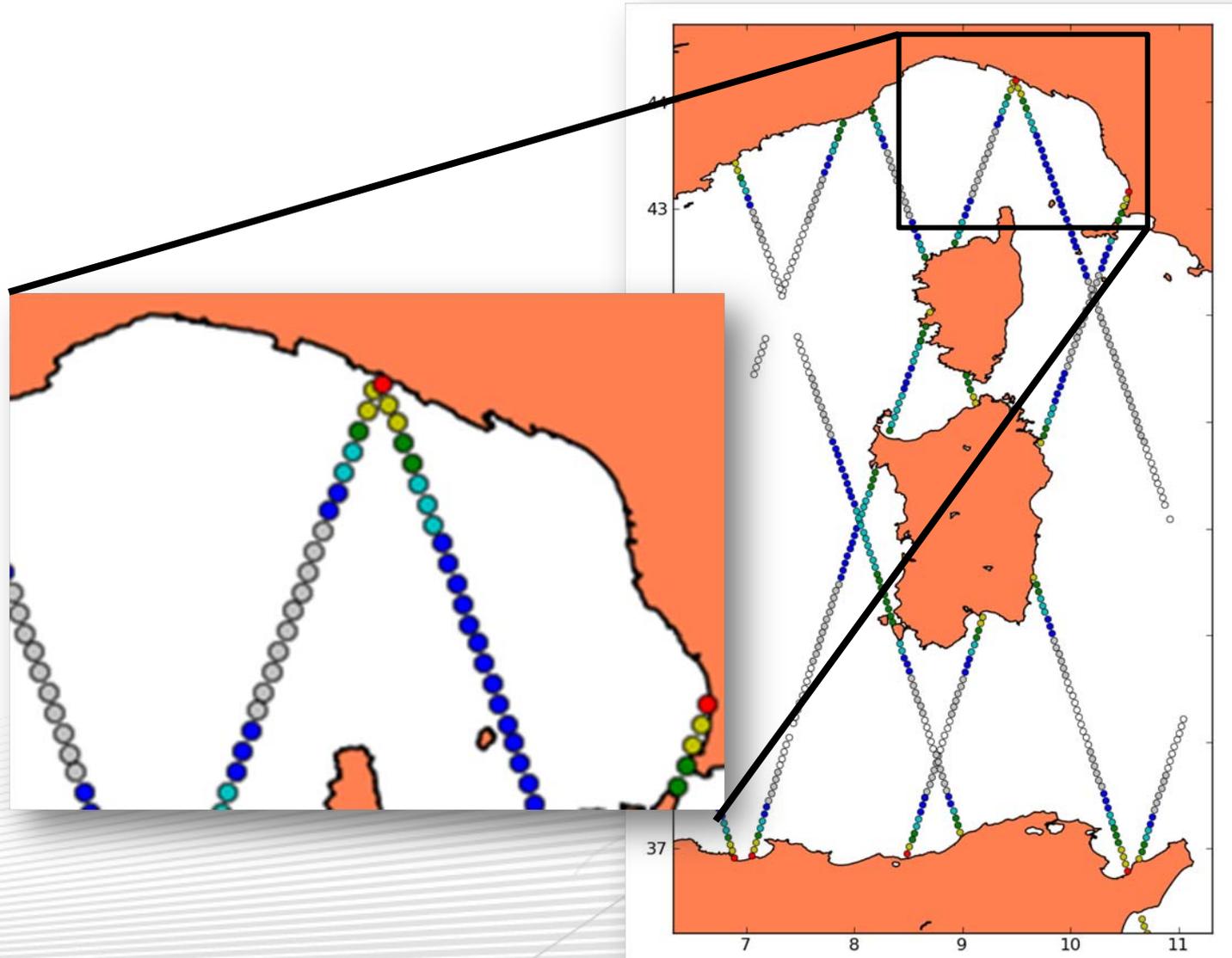


\*



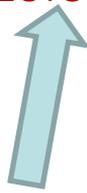
# Land proportion

- < 0.1
- < 0.2
- < 0.3
- < 0.4
- < 0.5
- > 0.5



# Building of the learning database

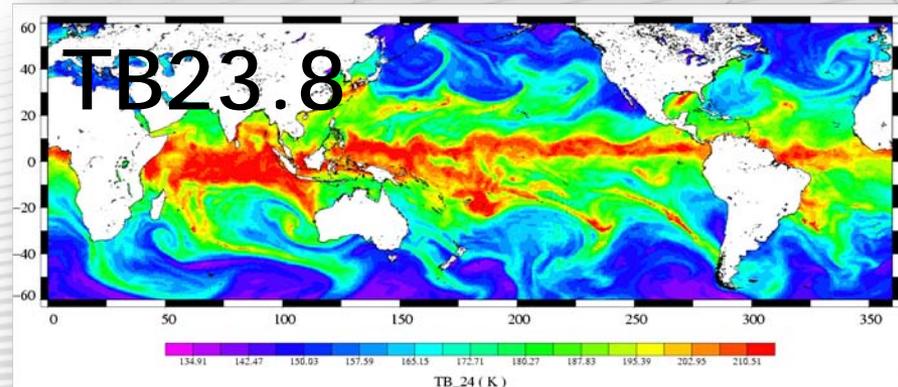
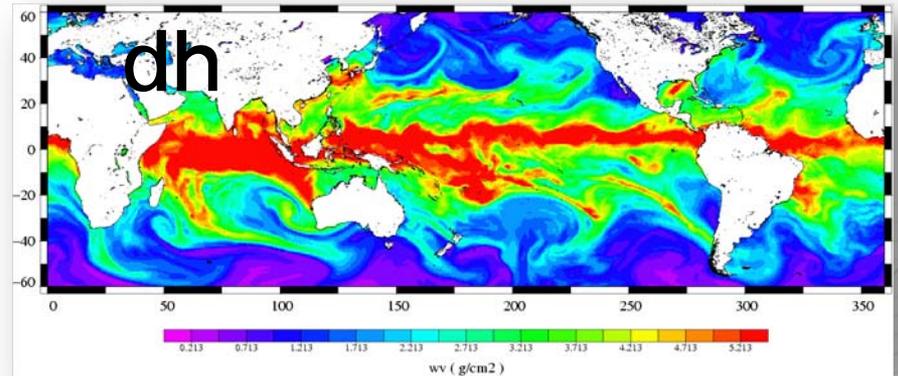
dh = NN (TB23.8, TB36.5,  $\sigma_0$ \_Ku, land\_prop23.8, land\_prop36.4)



## 1rst STEP : simulation of Ocean TBs

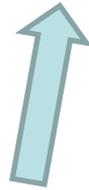
➤ A set of ECMWF analyses over sea with wet tropospheric correction, and other needed geophysical parameters: surface temperature and pressure, temperature and humidity profiles, surface wind speed

➤ Simulation over sea of brightness temperatures at 23.8 and 36.5 GHz thanks to a radiative transfer model



# Building of the learning database

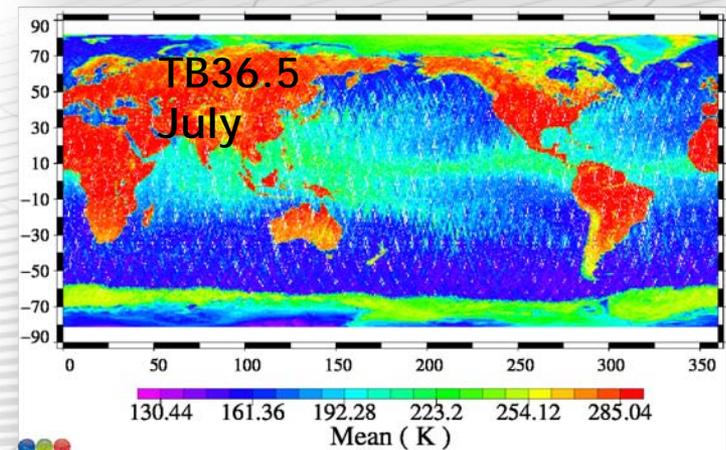
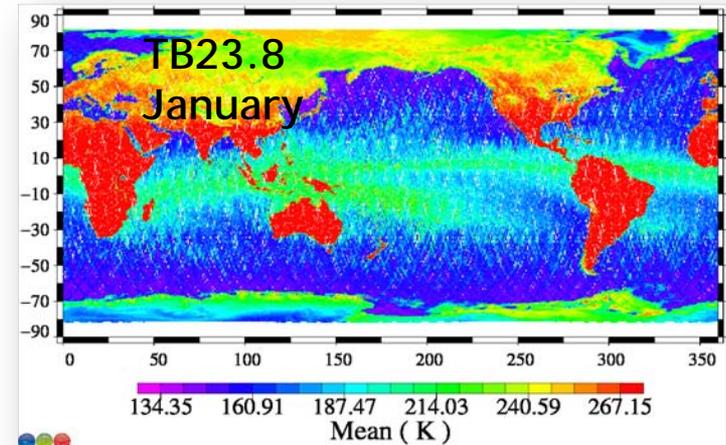
dh = NN (TB23.8, TB36.5,  $\sigma_0$ \_Ku, land\_prop23.8, land\_prop36.4)



## 2nd STEP : simulation of Mixed TBs

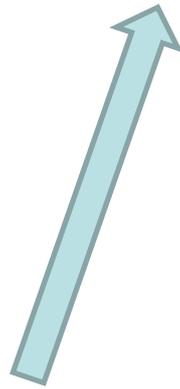
$$TB_{mixed} = (1-LP) * TB_{Ocean} + LP * TB_{Land}$$

- LP randomly chosen in a realistic distribution (obtained from one data cycle)
- TB\_Ocean simulated by the radiative transfer model
- TB\_Land : real measurement randomly picked up in a 10° latitude band



# Building of the learning database

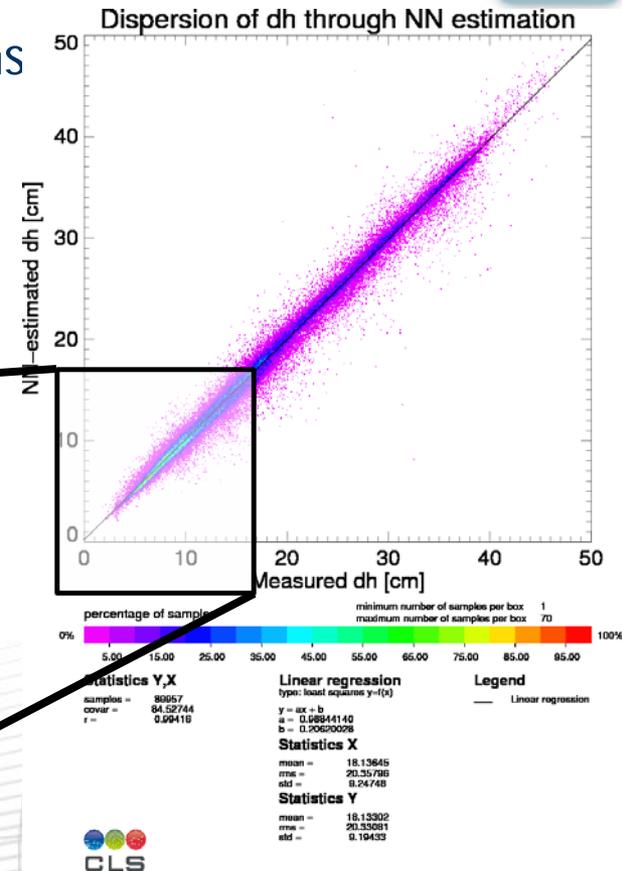
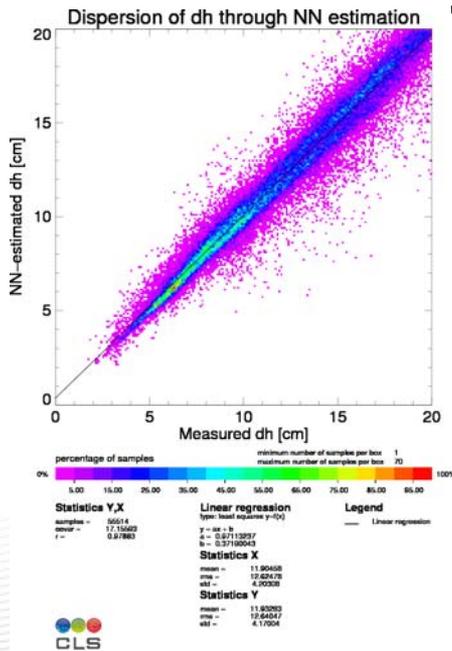
dh = NN (TB23.8, TB36.5,  $\sigma_0$ \_Ku, land\_prop23.8, land\_prop36.4)



Simulated with the radiative transfer model assuming a sea surface, smaller resolution

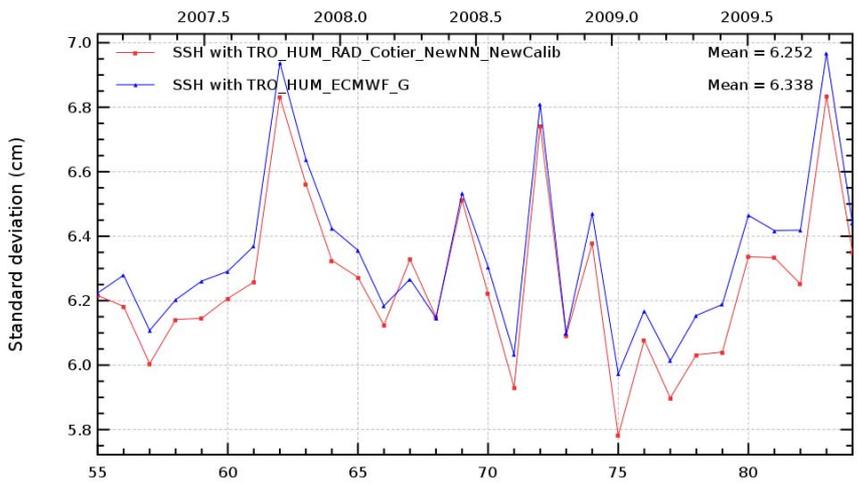
# Algorithm formulation

- Neural Net formalism to estimate weights and biases that minimize the differences (bias and rms between estimated and reference dh
- Architecture with 1 hidden layer of 8 neurons
- ➔ allows an optimal regression taking into account non linearities

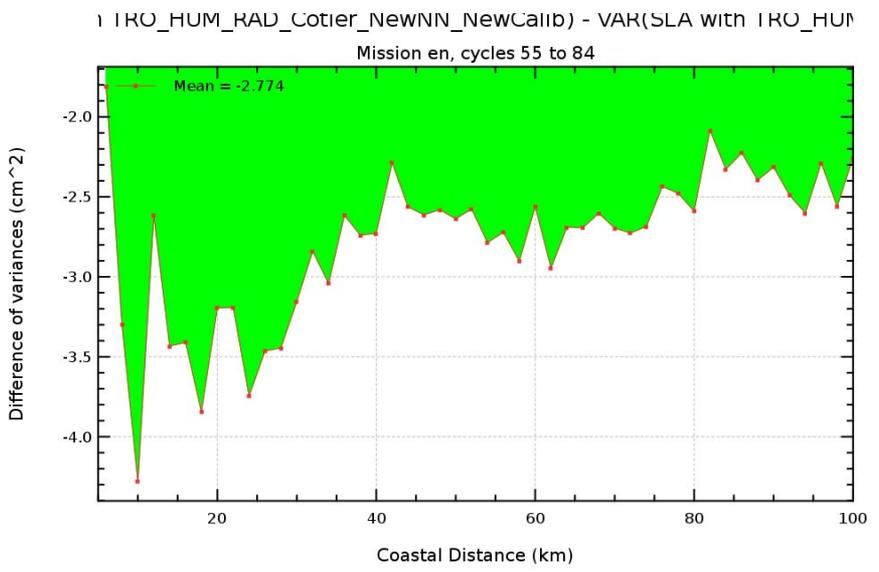
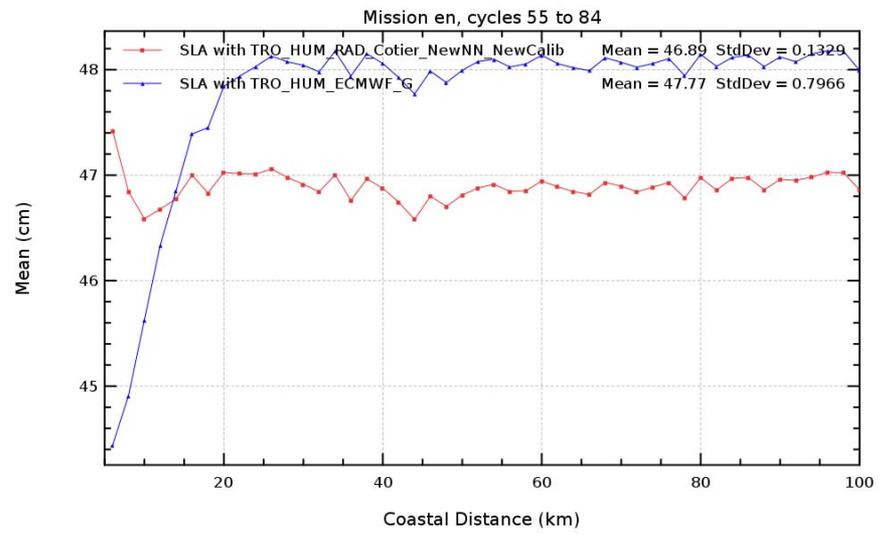
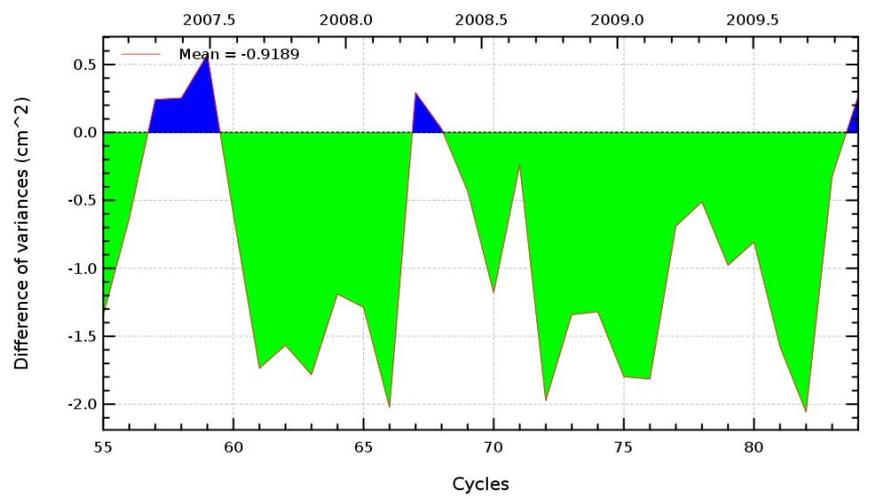


# CNN vs ECMWF estimation

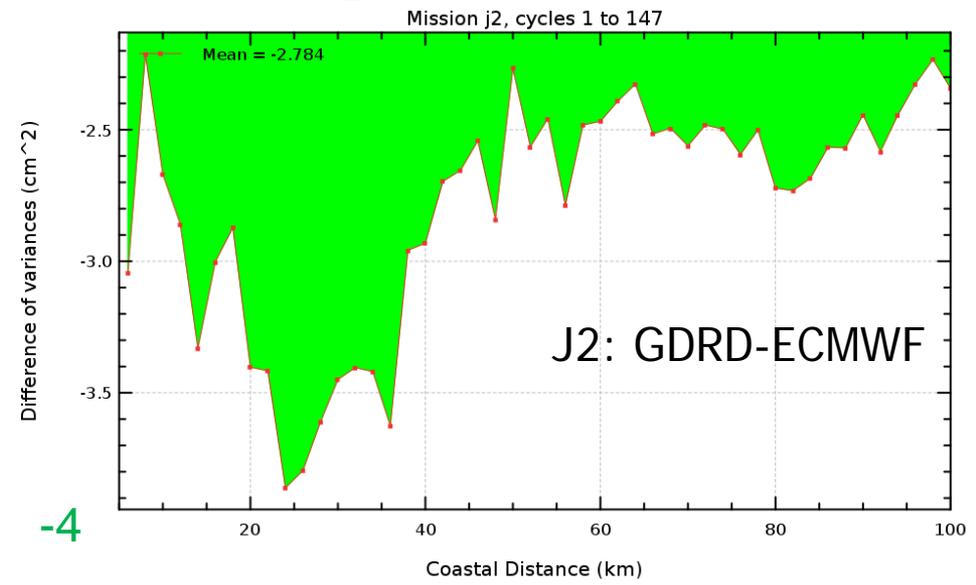
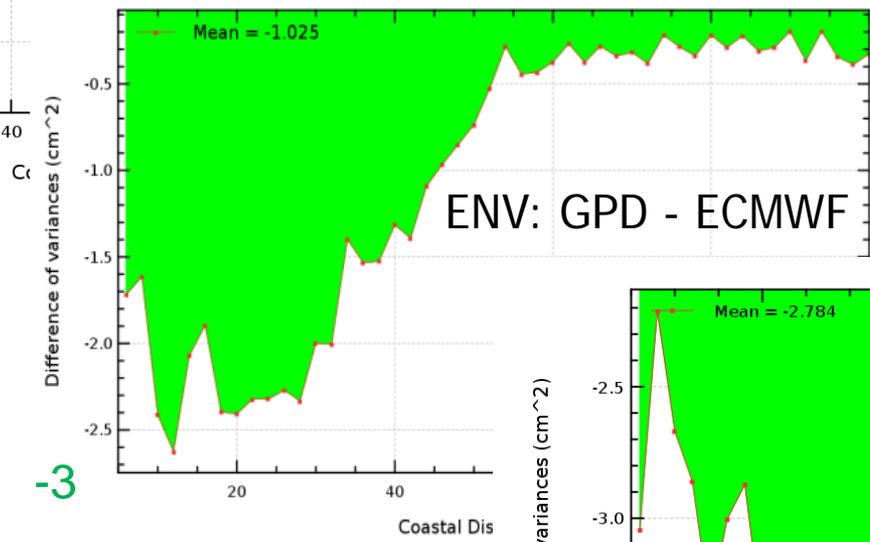
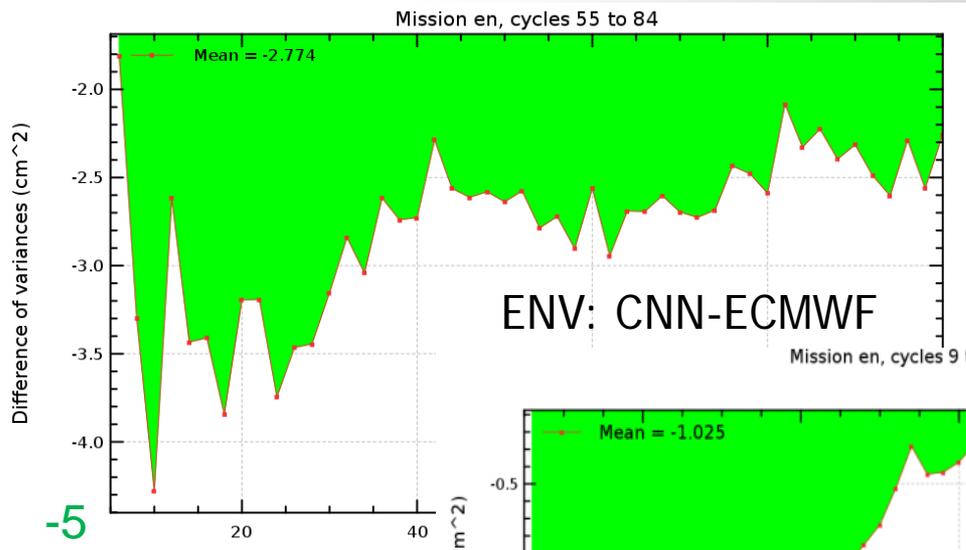
Standard deviations of SSH crossovers  
Mission en, cycles 55 to 84



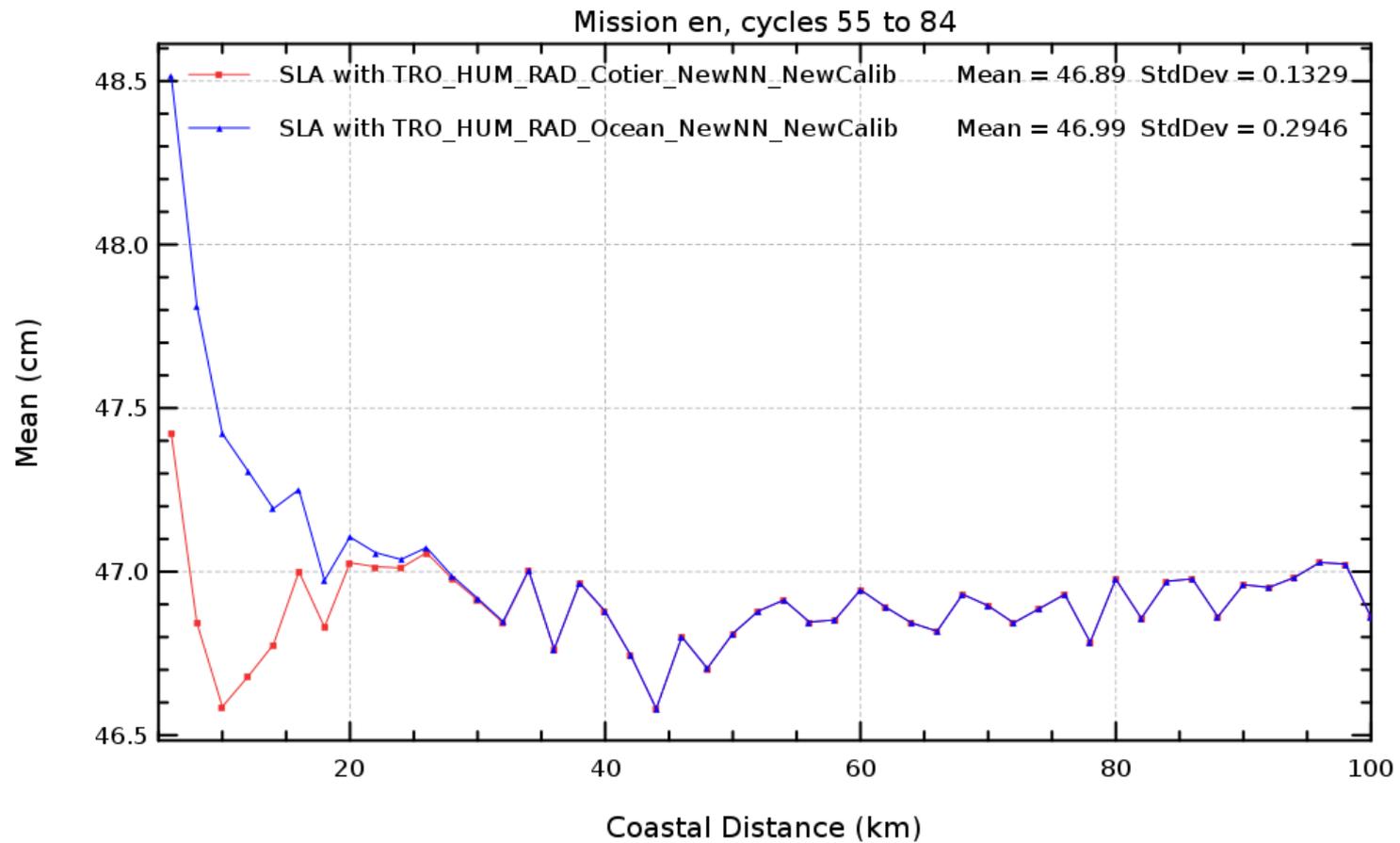
VAR(SLA with TRO\_HUM\_RAD\_Cotier\_NewNN\_NewCalib) - VAR(SLA with TRO\_HUM\_ECMWF\_G)  
Mission en, cycles 55 to 84



# SLA variance difference vs ECMWF

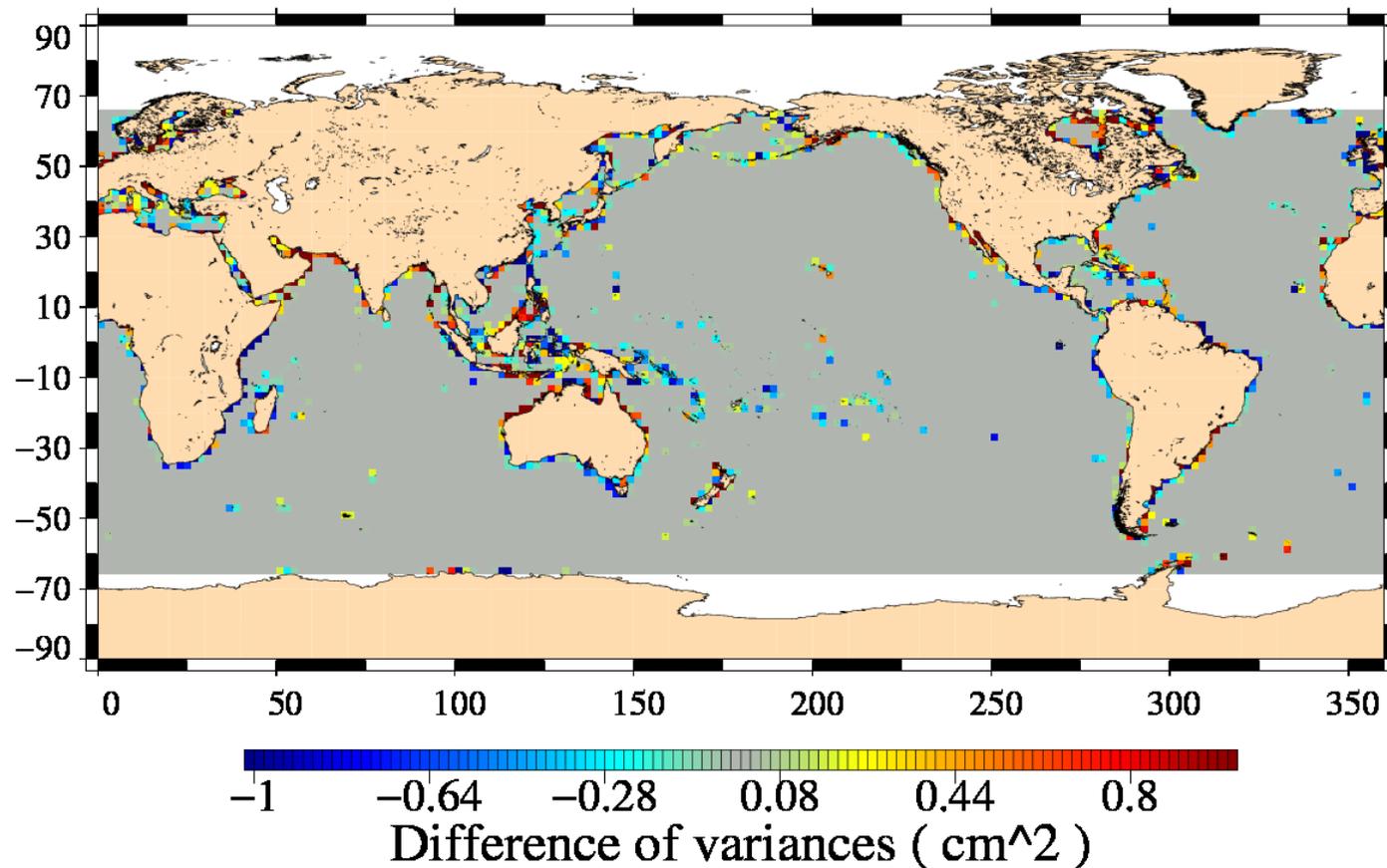


# CNN SLA when approaching the coast



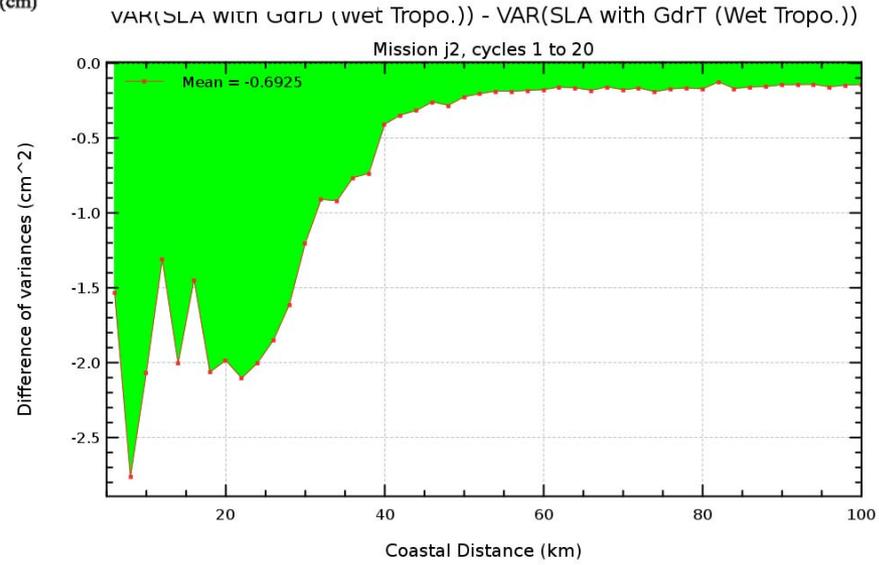
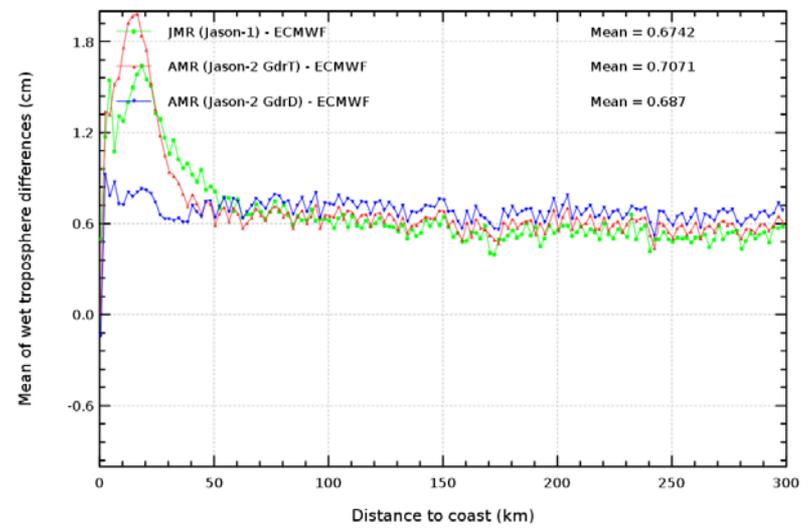
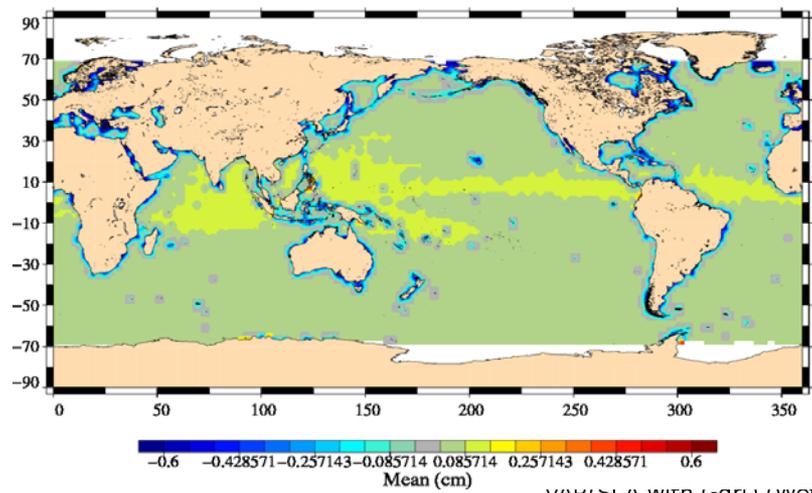
# SLA variance difference vs GDR

$M\_RAD\_Cotier\_NewNN\_NewCalib) - VAR(SLA \text{ with } TRO\_HUM\_RAD\_C$   
Mission en, cycles 55 to 84



# J2 GDR-D coastal dh

Mean of GdrD (Wet Tropo.) – GdrT (Wet Tropo.)  
Mission j2, cycles 1 to 20



# Conclusions & Perspectives

- Future altimetry missions defined to increase resolution and accuracy in altimetry measurements (S3/SRAL, SARAL/AltiKa, SWOT/Karin, ...) => will allow a better characterization of SSH coastal variability
- A global, high resolution and accurate wet tropospheric correction will be needed to take advantage of these new instruments
- Only the radiometer estimation, alone or combined with other products (models, GPS) will allow to reach this goal, models presenting insufficient spatial resolution and poor temporal sampling
- We developed a new algorithm derived from previous studies (Desportes, Brown) to improve the coastal wet tropospheric correction
- NN are used to easily and accurately take into account the required additional geophysical parameters

# Conclusions & Perspectives

- First results show a significant reduction of SLA variance with respect to the model and reduction of standard deviation of SSH at cross-overs
- For future altimetry missions, other aspects of processing and design of the radiometers should be analyzed and possibly improved:
  - Quality of the side-lobe correction (L1 processing)
  - Potential of the “original” measurements of the instrument (7 Hz for Envisat)
  - Review of the interpolation processing between radiometer and altimeter measurement
  - Enhancement of the radiometer resolution either through better antenna or innovative algorithm (currently used in imagery)
  - Potential of high frequency radiometers (higher spatial resolution, much smaller land impact)