



# Potential of new retrieval algorithms for the wet tropospheric correction of the Jason1/Jason2 missions

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## CONTEXT

- Retrieval algorithms formulated over learning database (dh, TB1, TB2, TB3)
- Formulation based on the minimization of both **global bias and the standard deviation**
  - Loglinear algorithms  $dh = c_0 + c_1 \ln(280 - TB_{18.7}) + c_2 \ln(280 - TB_{23.8}) + c_3 \ln(280 - TB_{34})$
  - Neural algorithms  $dh = NN(TB_{18.7}, TB_{23.8}, TB_{34})$
- This minimization of global bias and standard deviation ignores seasonal or regional specificities = > geographically correlated errors

## Objective

- To improve Open Ocean altimetry products
- To develop a new retrieval algorithm (Jason1 / Jason 2 configurations)
- To assess the performances with respect to classical algorithms

# Retrieval algorithm formulation

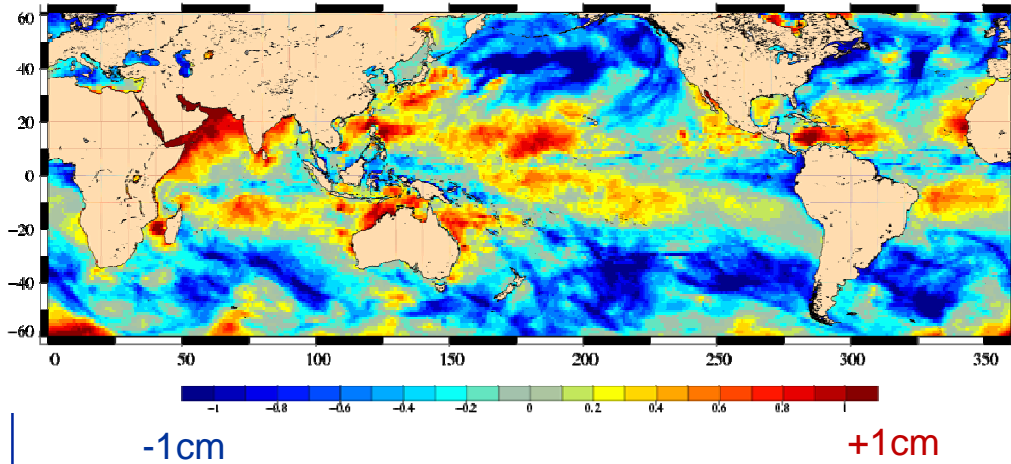
- Database built with 4 ECMWF analyses (one per season in 2008)
- For each mesh (0.5 degree):
  - Surface parameters : temperature (**SST**), wind, pressure
  - Atmospheric parameters: profiles (90 levels) of water vapor, cloud liquid
  - Wet tropospheric correction **dh** for each mesh
  - Lapse rate  $\gamma_{800}$  : temperature decrease rate between the surface and 800mb
  - Simulated **TBs at 18.7, 23.8 and 34 GHz** with a radiative transfer model
- 10% of the database for learning, the rest for testing

# Performances between different types of algorithm

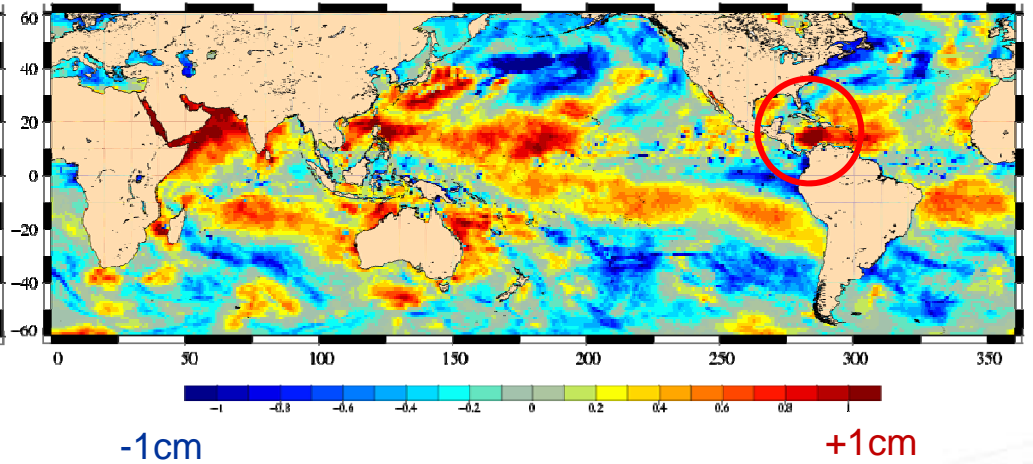
- linear bi-frequency :  $dh=c_0+c_1 \ln(280-TB_{23.8})+c_2 \ln(280-TB_{34})$
- neural bifrequency :  $dh=NN(TB_{23.8}, TB_{34})$
- linear tri-frequency :  $dh=c_0+c_1 \ln(280-TB_{18.7}) + c_2 \ln(280-TB_{23.8}) + c_3 \ln(280-TB_{34})$
- neural tri-frequency :  $dh=NN(TB_{18.7}, TB_{23.8}, TB_{34})$

# Residuals = Differences between retrieved and reference dh

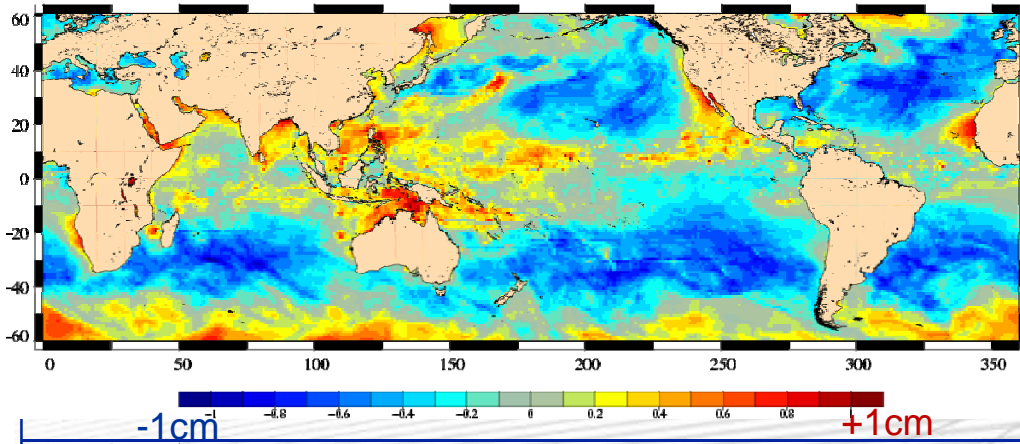
LIN 23.8 34 =>  $\sigma=8.4$  mm



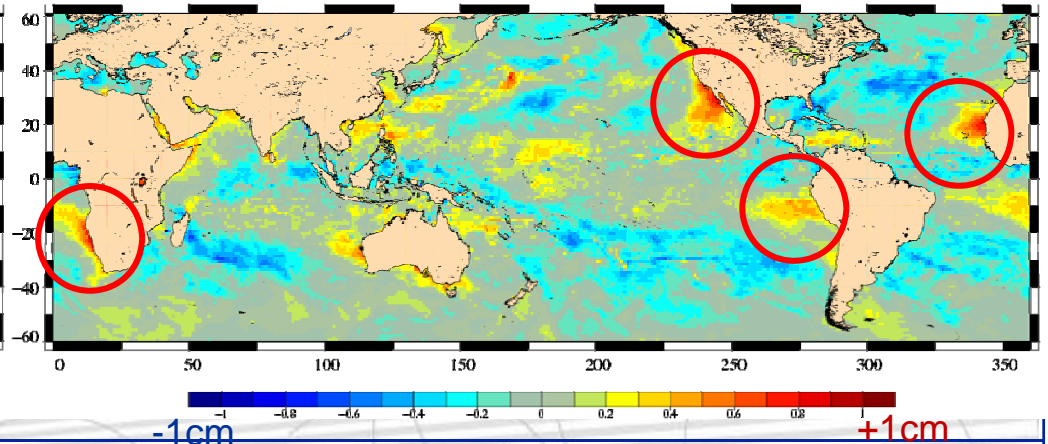
NN 23.8 34 =>  $\sigma=5.6$  mm



LIN 18.7 23.8 34 =>  $\sigma=4.1$  mm



NN 18.7 23.8 34 =>  $\sigma=2.2$  mm



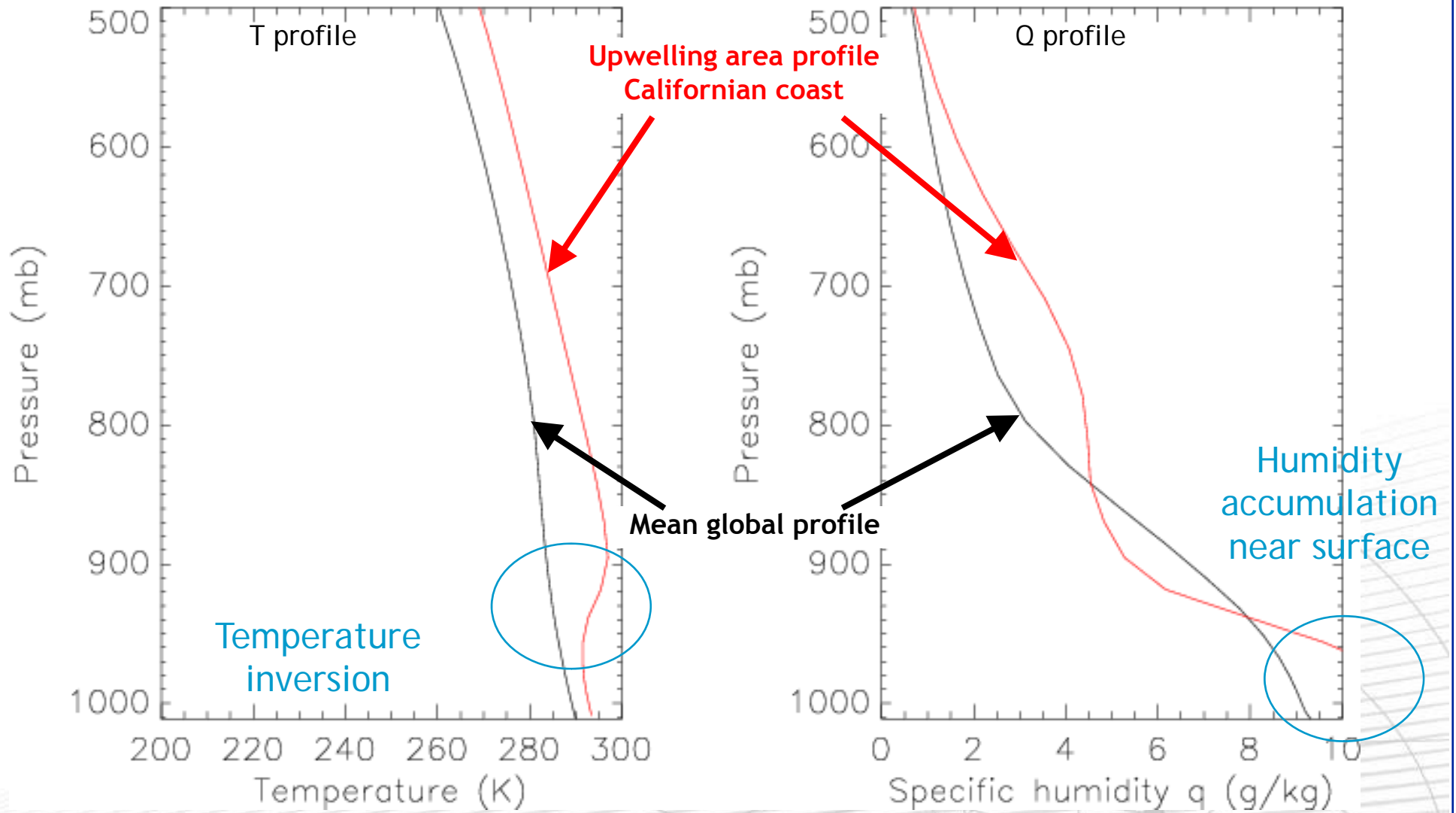
OSTST – Seattle – June 2009



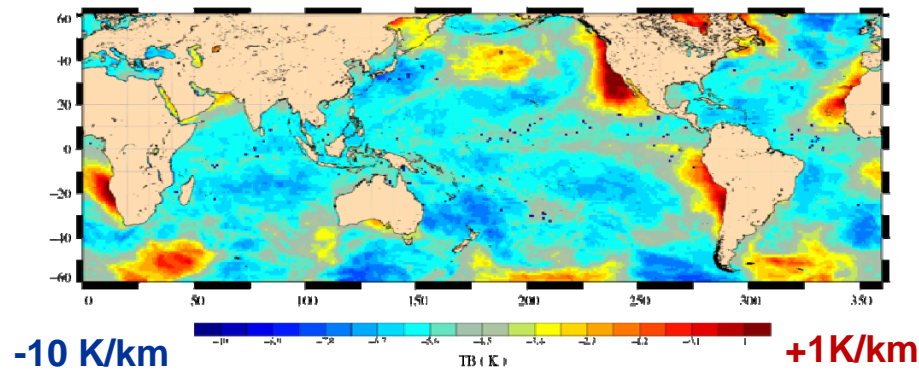
# Performances of the different algorithms

- A linear tri-frequency is better than a neural bi-frequency.
- Necessity to add the altimeter backscattering information in case of bi-frequency radiometers (Envisat, AltiKa, Sentinel 3)
- Even with a neural regression
  - Some underestimations
  - Over estimations in the Eastern part of the subtropical bassins
- New parameters added as inputs in the retrieval algorithm (Obligis et al, 2009)
  - decrease rate of the temperature
  - SST

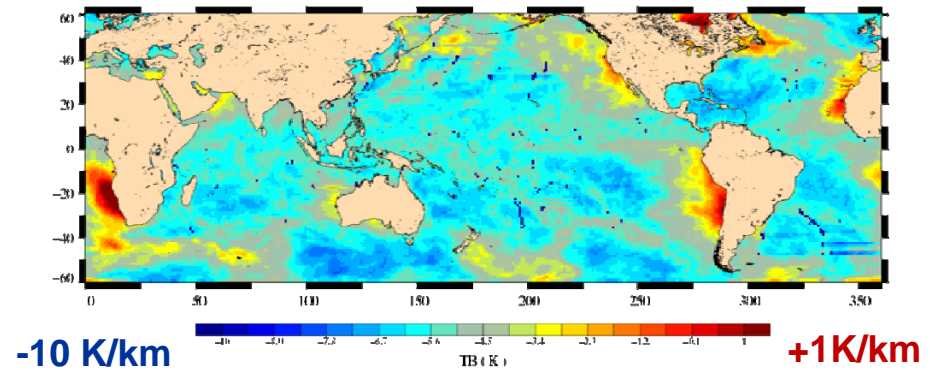
# Example of atmospheric profiles near the californian coast



## Interannual variability of intensity/extent of the upwelling



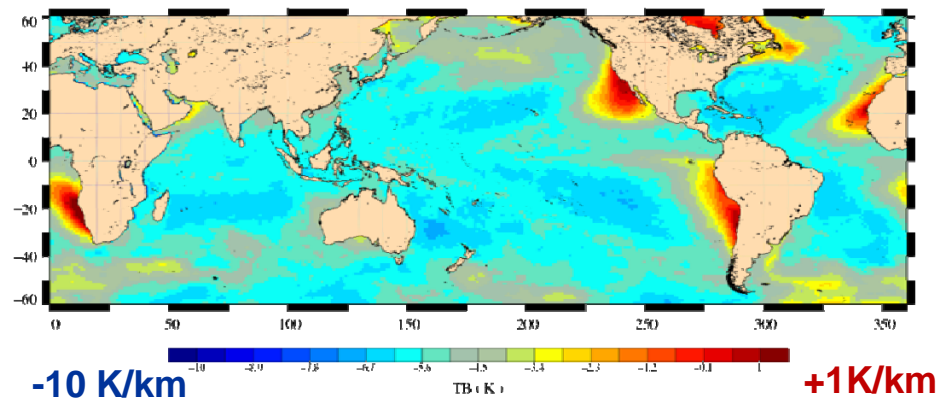
2008



2004

- Auxiliary information computed from 5 years of ECMWF atmospheric profiles

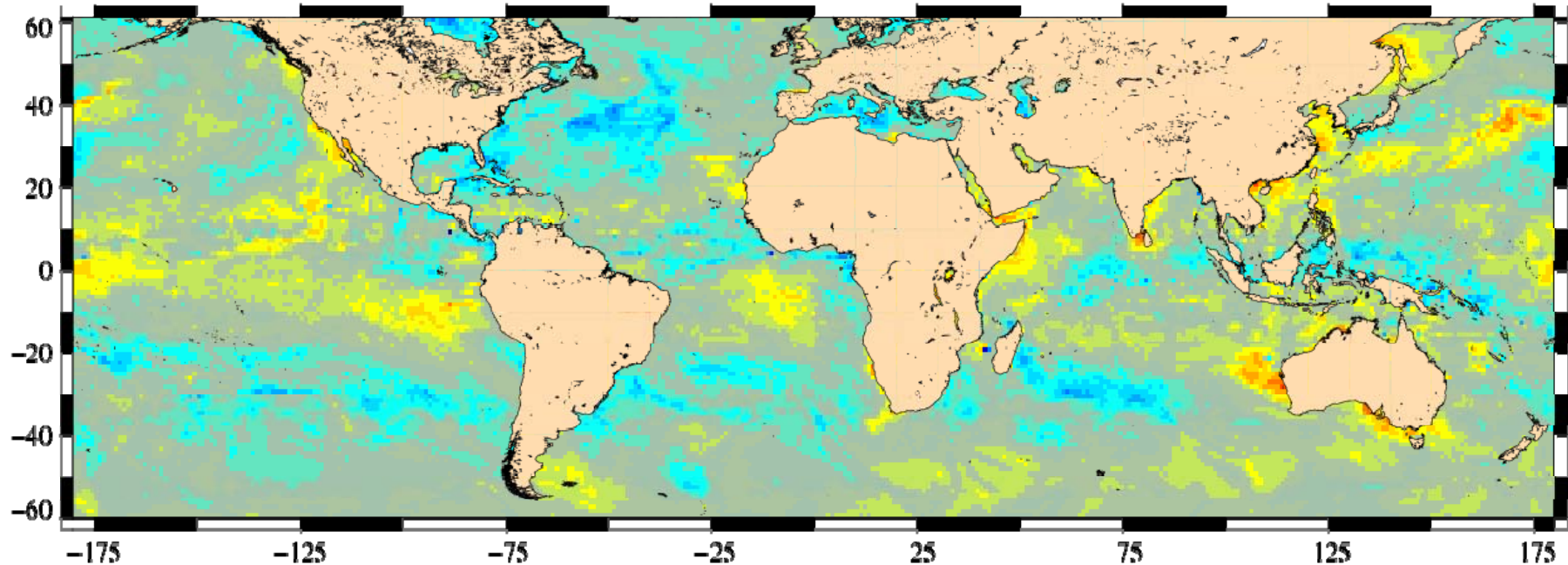
GAMMA\_800 2003-2008





# Performances of a new algorithm on simulated database

$dh=NN(TB18.7, TB23.8, TB34, \gamma 800, SST)$



Nb of data	: 43039	St. Dev	: 0.1658828	Skewness	: 0.2591651	Minimum	: -1.1421746
Mean	: -0.0002088	Rms	: 0.1658829	Kurtosis	: 1.5086549	Maximum	: 1.3581051

# Conclusions

- Evaluation of a new algorithm with SST and  $\gamma 800$  as new input parameters
- SST does not bring additional information in case of a 3-frequency radiometer. Already contained in the low frequency TB (18.7 GHz)
- $\gamma 800$  allows considering specificities of atmospheric profiles in upwelling areas
- With the new algorithm, large scale differences with the ECMWF model are reduced (not shown)