An aerial photograph of a town, likely in the Pyrenees region, is shown from a high angle. The town is surrounded by green hills and is partially obscured by a thick layer of white clouds. Overlaid on the bottom left of the image is a white weather map showing isobars (lines of equal pressure) and wind vectors (arrows). The isobars are labeled with values such as 1010, 1015, 1020, 1025, 1030, 1035, and 1040. The wind vectors are shown as arrows of varying lengths, indicating wind direction and speed. The background of the slide is a dark blue gradient with a stylized sun and cloud icon in the top left corner.

**On the use of BUOY
observations
in the Numerical Weather
Prediction data
assimilation systems at
Météo-France**

**Jean-François MAHFOUF
(and many colleagues)
CNRM/GMAP
Toulouse (France)**



Outline

1. The global model ARPEGE
2. The limited area models ALADIN
3. Use of BUOY data for atmospheric and surface analyses
(data monitoring and diagnostics)
4. Forecast sensitivity to observations with ARPEGE
5. Conclusions

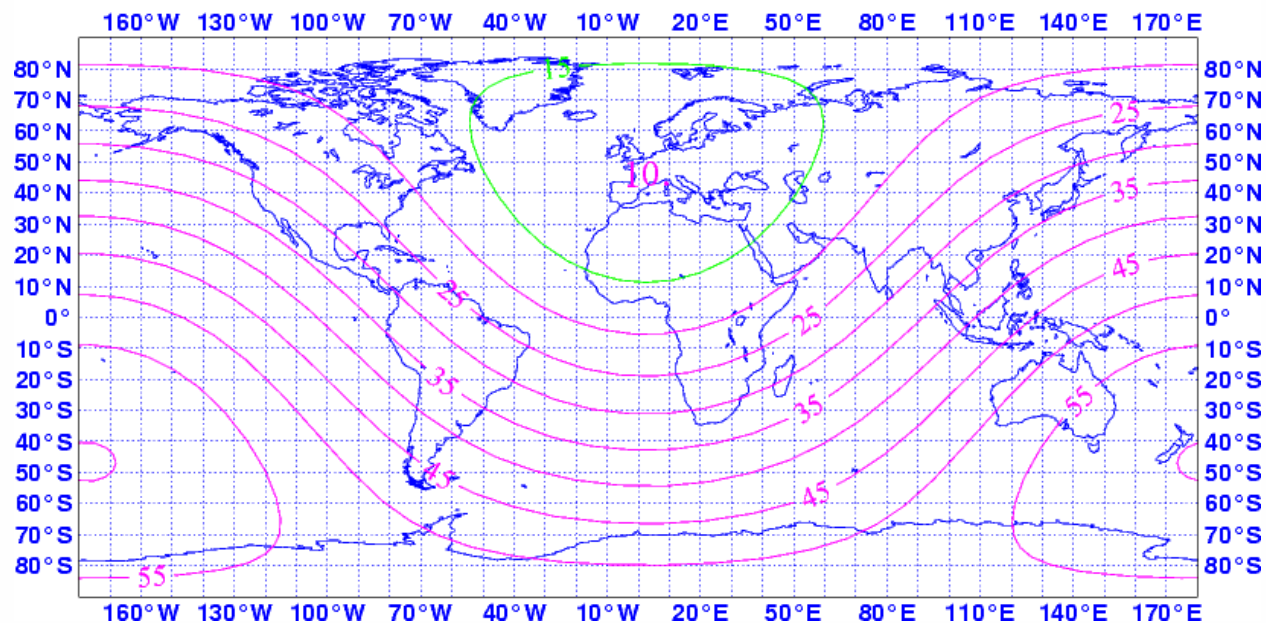


The global model ARPEGE

Global model ARPEGE (1)

- Spectral model with variable resolution : T_L798C2.4L70
 - Resolution from 10 km to 60 km, 70 levels from 17m to 0.05 hPa

ARPEGE Resolution T798 C2.4 in KM



Forecast cut-offs at 00 UTC :

Long cut-off (assimilation) : **7h30** – Short cut-off (production) : **2h15** – Very short cut-off : **1h10**



Global model ARPEGE (2)

- 4D-Var assimilation (6h window) :

- 2 loops of minimization : T107C1L70 (25 iterations) + T323C1L70 (25 iterations)
 - 2nd inner loop with simplified physics (including large scale condensation)
- Variational bias correction scheme since 2008
- Background error variances from an Ensemble Data Assimilation system (4D-Var at lower resolution) since 2008
- Assimilation time slots : 1 hour

- Data used :

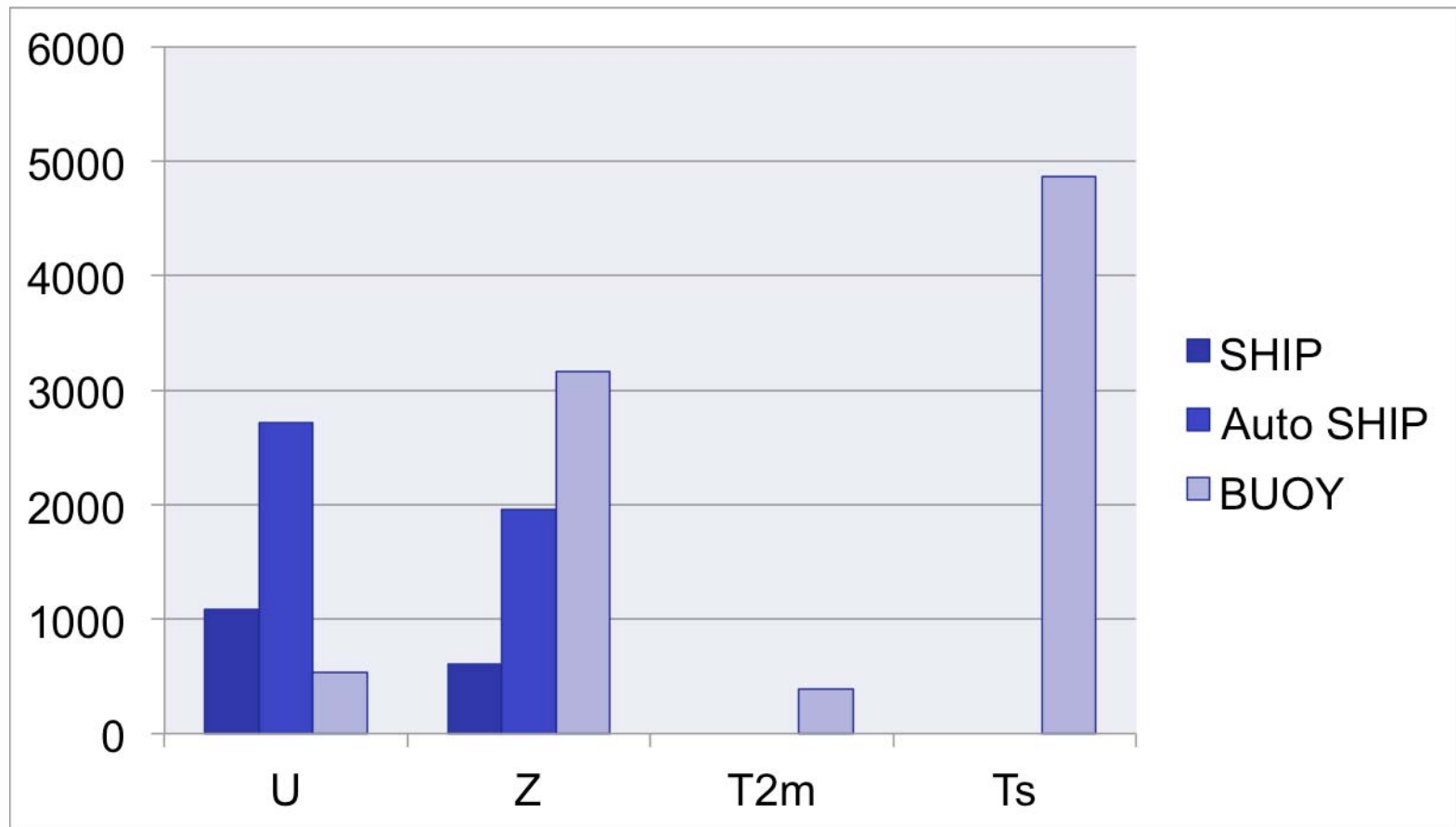
- SYNOP, SHIP, **BUOY**, AIREP, AMDAR, ACARS, TEMP, PILOT
- AMV GOES + Meteosat + MTSAT-1R, MODIS (Terra, Aqua), AVHRR/NOAA
- HIRS, AMSU-A, AMSU-B/MHS, NOAA 15, 16, 17, 18, 19, MetOp and Aqua
- SSMI/S DMSP F16, 17, 18, AIRS/AQUA, IASI/MetOp, GPS-RO, GPS-ZTD
- Sea surface winds from scatterometer ASCAT/MetOp
- Meteosat CSR
- *SST analysis from Ol scheme using SHIP and **BUOY** reports*



Type of observations over oceans

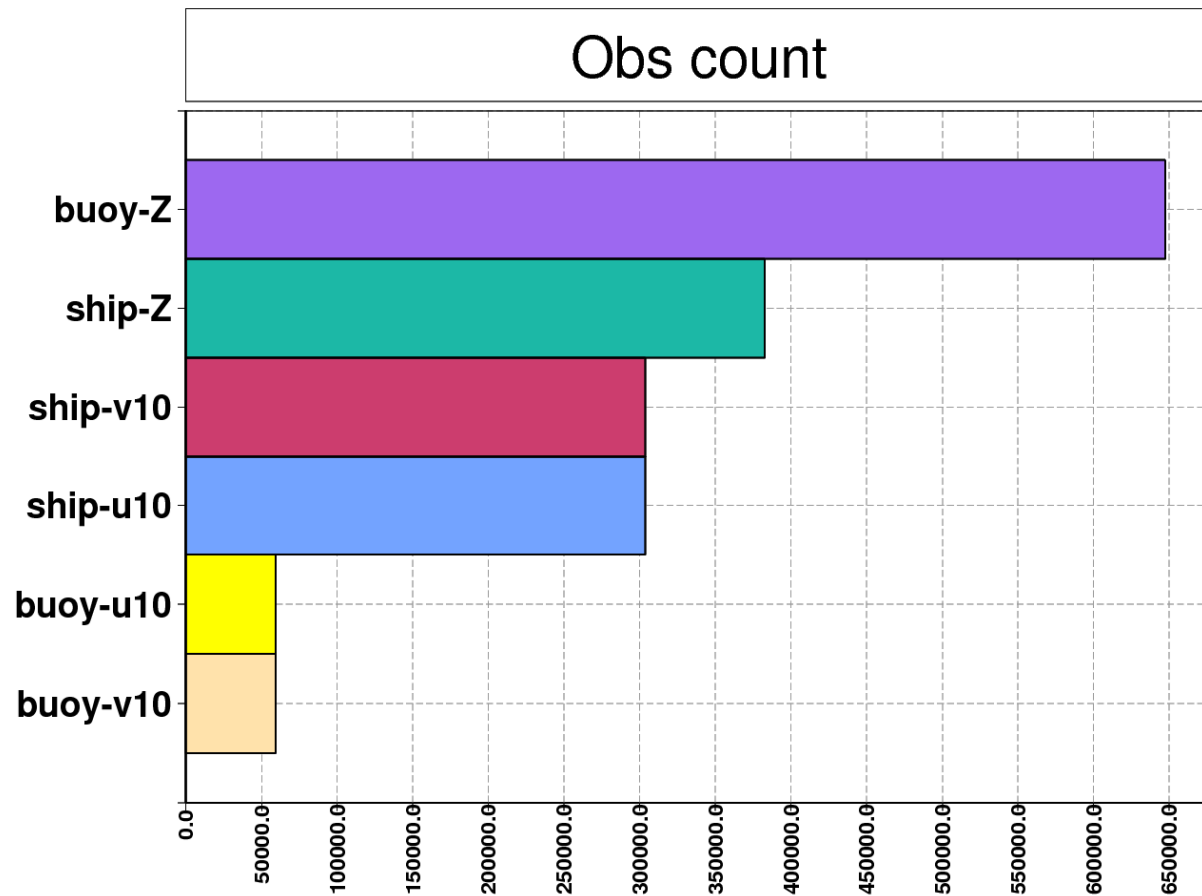
- **Conventional observations :**
 1. BUOY : surface meteorological observations from moored and drifting buoys
 2. SHIP : surface meteorological observations from ships (similar to SYNOP reports over land)
 3. BATHY : sea surface and below soundings
 4. TESAC : Argo profiling floats
 - Measurements : Surface pressure (geopotential height), wind components, sea surface temperature, screen-level temperature and relative humidity.
- **Satellite observations** informative about the ocean surface : scatterometer derived winds (ASCAT), skin temperature (AVHRR, MODIS)

Measured parameters according to data type



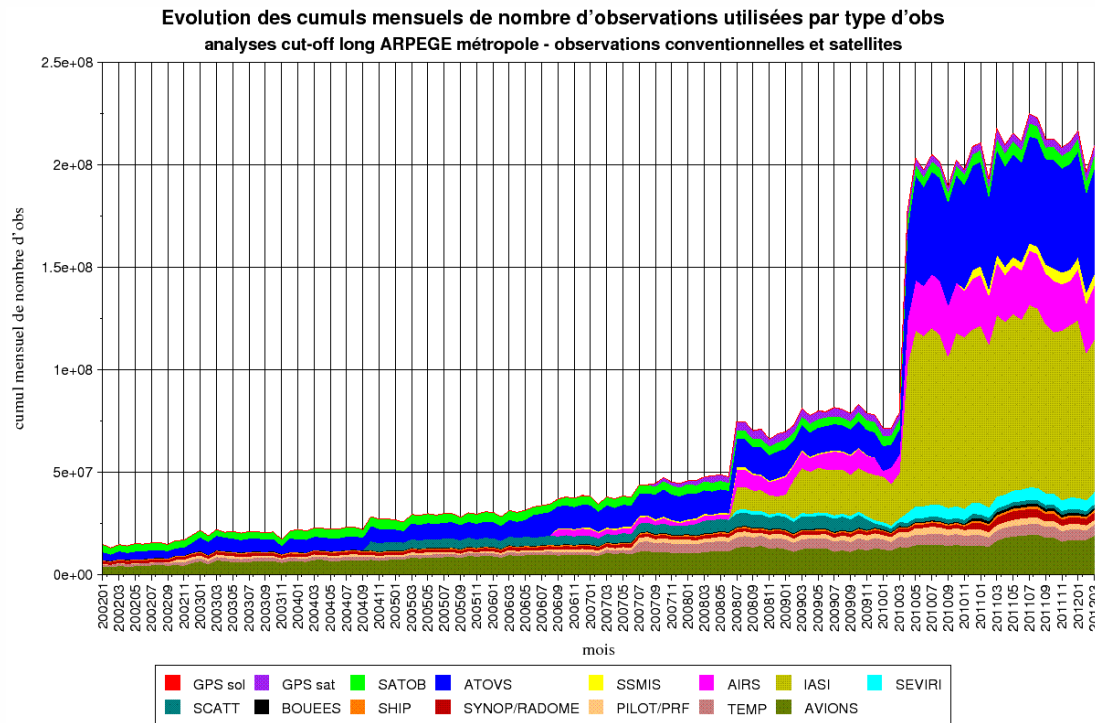
Global data available over a 6 hour time window

Measured parameters according to data type



2010/12/02 to 2011/01/15 : assimilated in ARPEGE 4D-Var system

Number of observations assimilated in ARPEGE

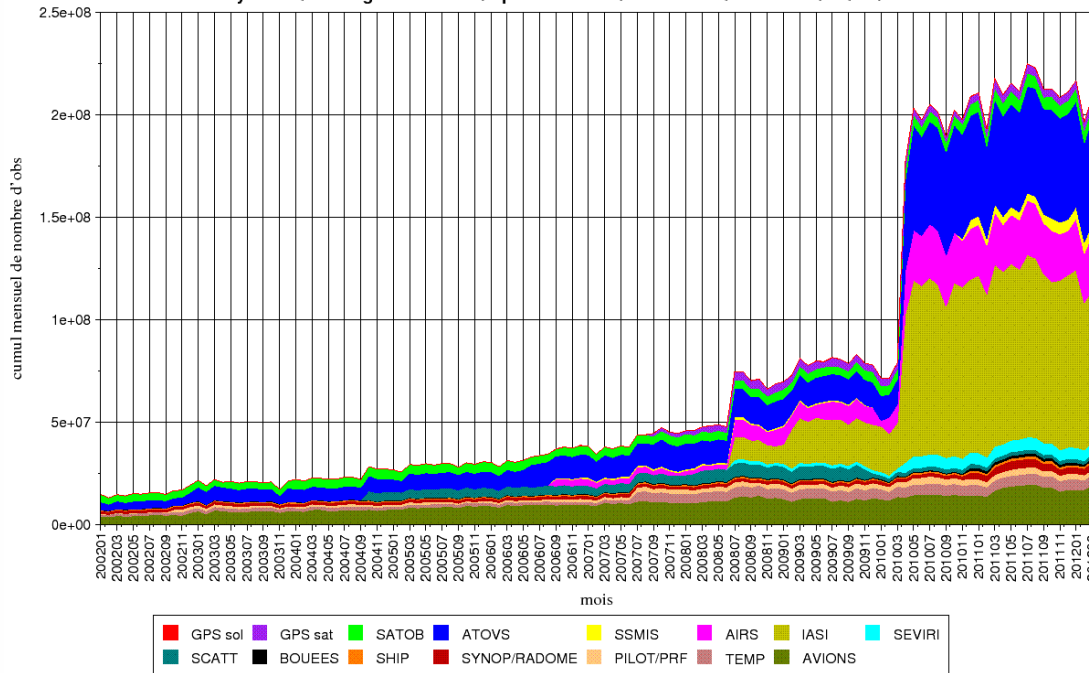


All observations



Number of observations assimilated in ARPEGE

Evolution des cumuls mensuels de nombre d'observations utilisées par type d'obs
analyses cut-off long ARPEGE métropole - observations conventionnelles et satellites



All observations



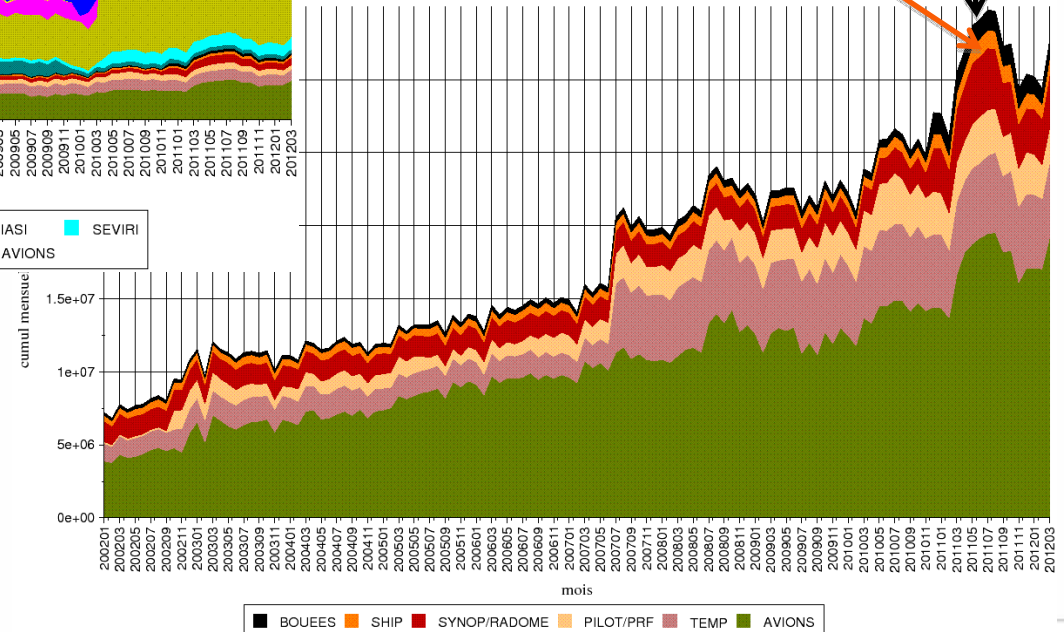
2.7 %

SHIP

1.7 %

BUOY

Evolution des cumuls mensuels de nombre d'observations utilisées par type d'obs
analyses cut-off long ARPEGE métropole - observations conventionnelles



Only conventional





Degrees of Freedom for Signal

1. Degrees of freedom of an assimilation system are controlled by the observations assimilated :

$$\begin{aligned} DFS &= Tr \left(\frac{\partial(H\mathbf{x}^a)}{\partial(\mathbf{y}^o)} \right) \\ &= Tr(\mathbf{HK}) \end{aligned}$$

Under linear assumption

- Sum of impacts of a given observation on its analyzed equivalent : self-sensitivity of an observation on the analysis
- When \mathbf{R} is block-diagonal, the DFS can be derived for each observation type :

$$DFS_i = Tr(\Pi_i \mathbf{HK} \Pi_i^T)$$

2. References : Chapnik et al. (2004), Fourrié et al. (2005)

Degrees of Freedom for Signal (DFS)

1. Sensitivity of the analysis to a perturbation of a particular subset of observations : weight of these observations in the analysis

$$DFS = Tr \left(\frac{\partial(H\mathbf{x}^a)}{\partial(\mathbf{y}^o)} \right)$$
$$= Tr(\mathbf{HK})$$

Under linear assumption

When \mathbf{R} is block-diagonal, DFS can be computed for each observational dataset

1. Estimation from a randomization technique (Desroziers and Ivanov, 2001)

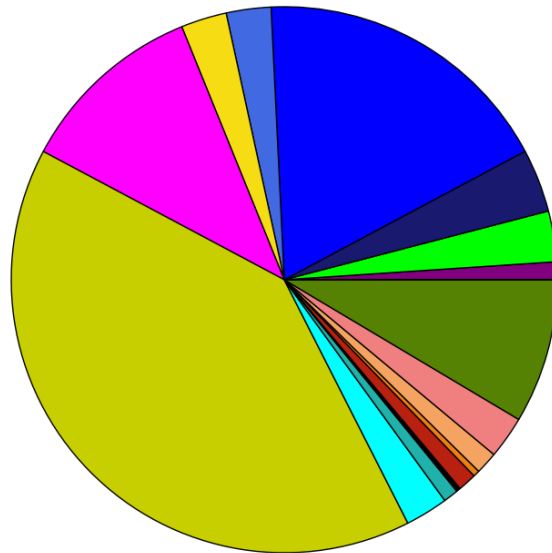
$$Tr(\mathbf{C}) \approx \boldsymbol{\eta}^T \mathbf{C} \boldsymbol{\eta}$$

Reference analysis (x_a)
Analysis with perturbed obs (x_a^*)

$$Tr(\mathbf{HK}) \approx (\mathbf{R}^{-1/2} \boldsymbol{\eta})^T (\mathbf{H} \delta x_a^* - \mathbf{H} \delta x_a)$$

Information content of observations in ARPEGE

Proportions des nombres d'observations utilisées par type d'obs
analyses cut-off long - ARPEGE metropole oper
observations conventionnelles et satellites
cumul du nombre d'observations utilisées sur la période 2011110300 - 2011110318 : 6984383



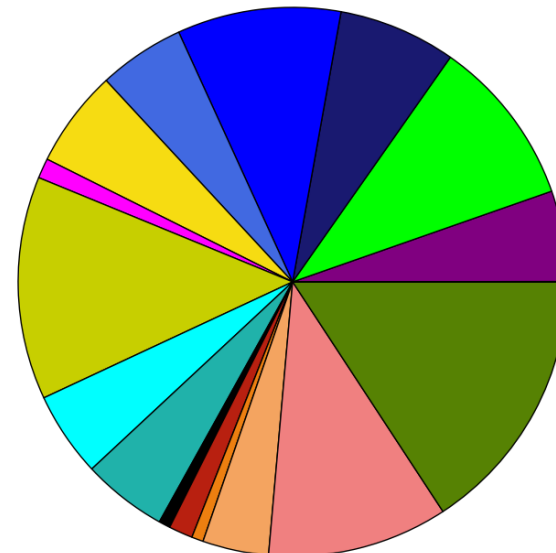
GPS ground	0.01%	AIRES	11.09%	PILOT/PRF	1.32%
GPS sat	1.04%	IASI	40.26%	TEMP	2.51%
SATOB	3.00%	SEVIRI	2.53%	AIRCRAFTS	8.55%
ATOVS HIRS	3.75%	SCATT	0.89%	RADAR Vr	0.00%
ATOVS AMSU-A	17.97%	BUOY	0.25%	RADAR Hur	0.00%
ATOVS AMSU-B	2.65%	SYNOP/SYNOR/RADOME	1.10%	BOGUS	0.00%
SSMIS	2.71%	SHIP	0.38%		

Number of observations

BUOYS = 0.25 %

SHIP = 0.38 %

Part des DFS par type d'obs
analyses cut-off long - ARPEGE metropole oper
observations conventionnelles et satellites
cumul du DFS sur la période 2011110300 - 2011110318 : 188312



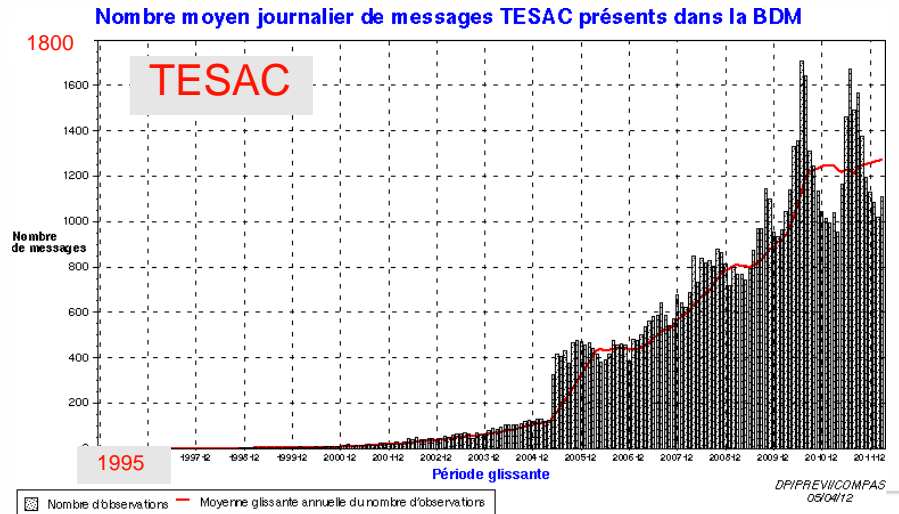
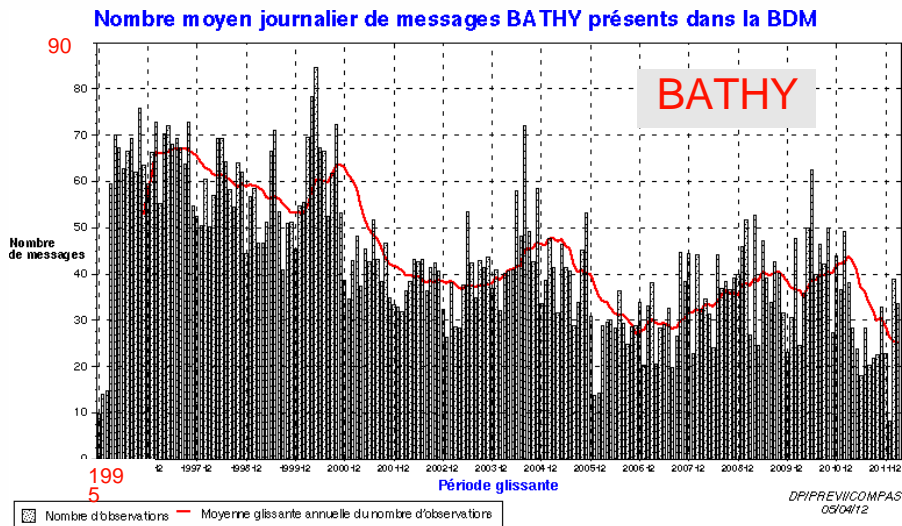
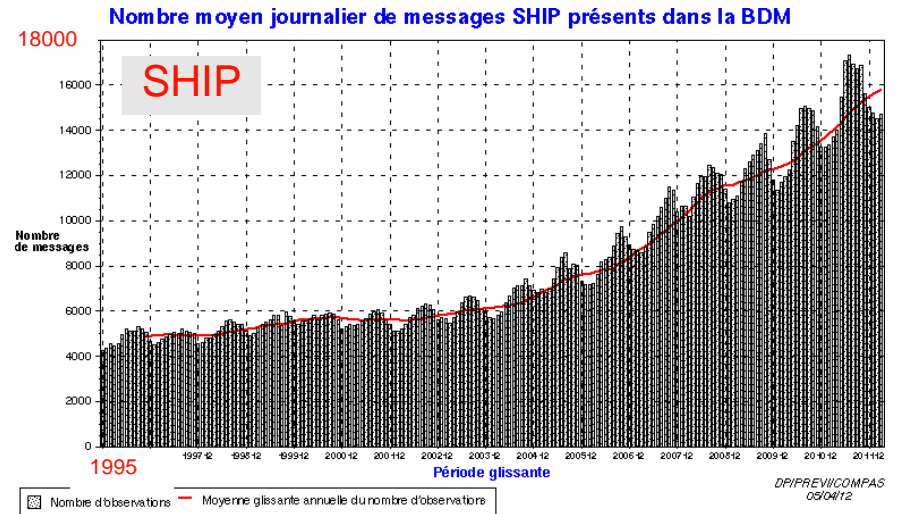
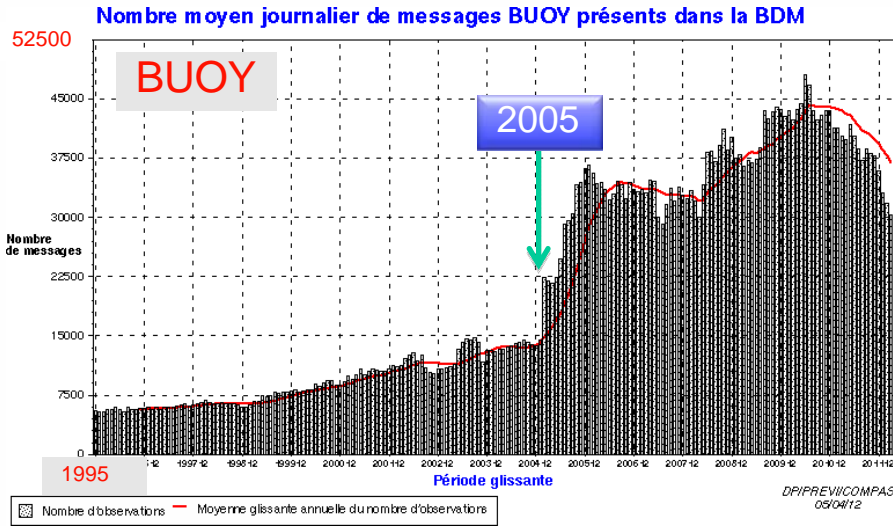
GPS ground	0.02%	AIRES	1.17%	PILOT/PRF	3.89%
GPS sat	5.34%	IASI	13.14%	TEMP	10.63%
SATOB	9.89%	SEVIRI	4.99%	AIRCRAFTS	15.77%
ATOVS HIRS	6.92%	SCATT	5.00%	RADAR Vr	0.00%
ATOVS AMSU-A	9.64%	BUOY	0.68%	RADAR Hur	0.00%
ATOVS AMSU-B	5.08%	SYNOP/SYNOR/RADOME	1.41%	BOGUS	0.00%
SSMIS	5.74%	SHIP	0.69%		

Degrees of Freedom for Signal

BUOYS = 0.68 %

SHIP = 0.69 %

Number of daily archived messages in the MF data base

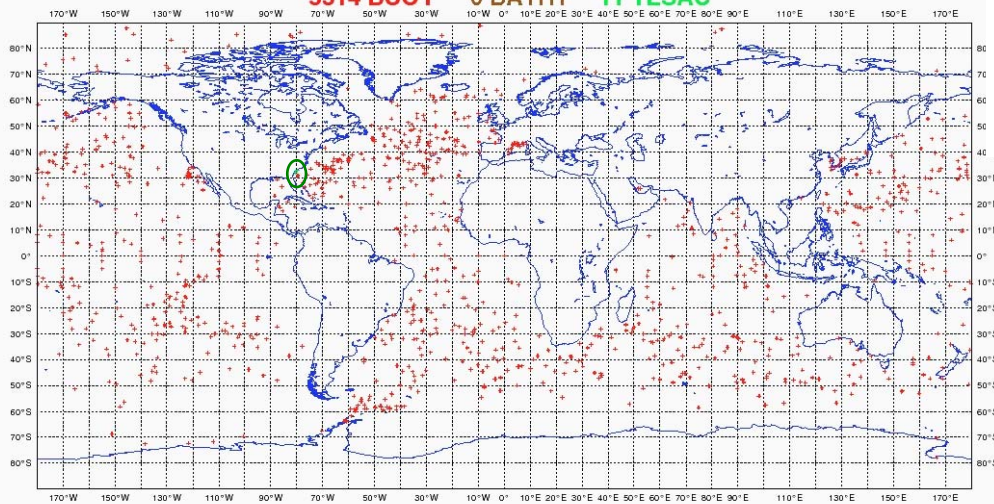


Number of observations assimilated in ARPEGE

80 % of available data are used

BUOY = 5314

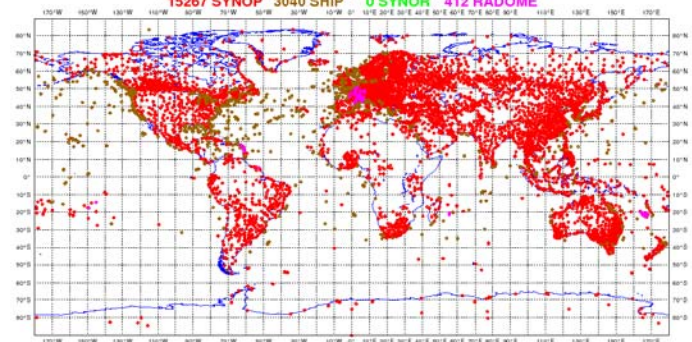
METEO-FRANCE couverture de donnees - BUOY - 2012/05/02 00H UTC cut-off long
Nombre total d'observations apres screening : 5325



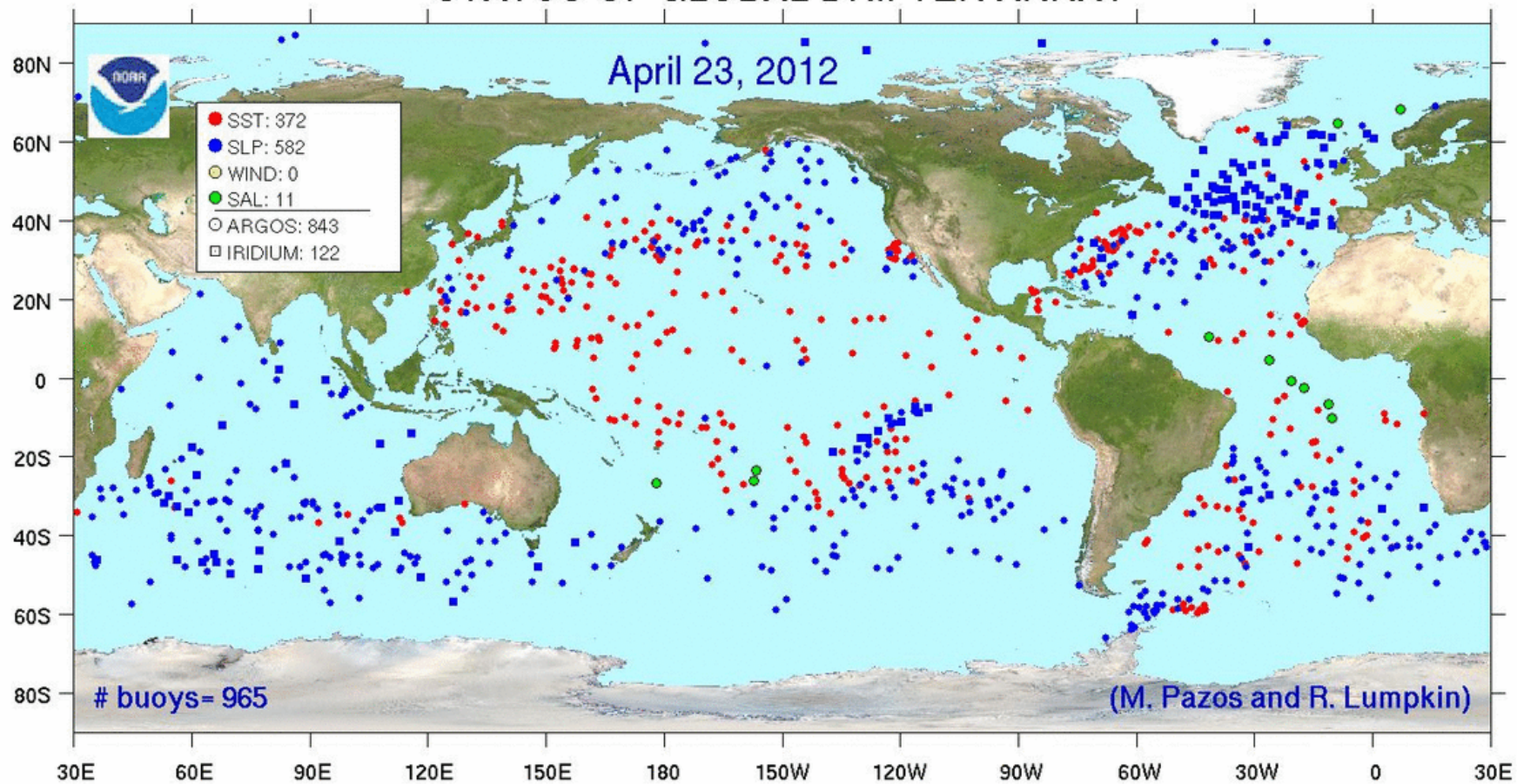
80 % of available data are used

SHIP = 3040

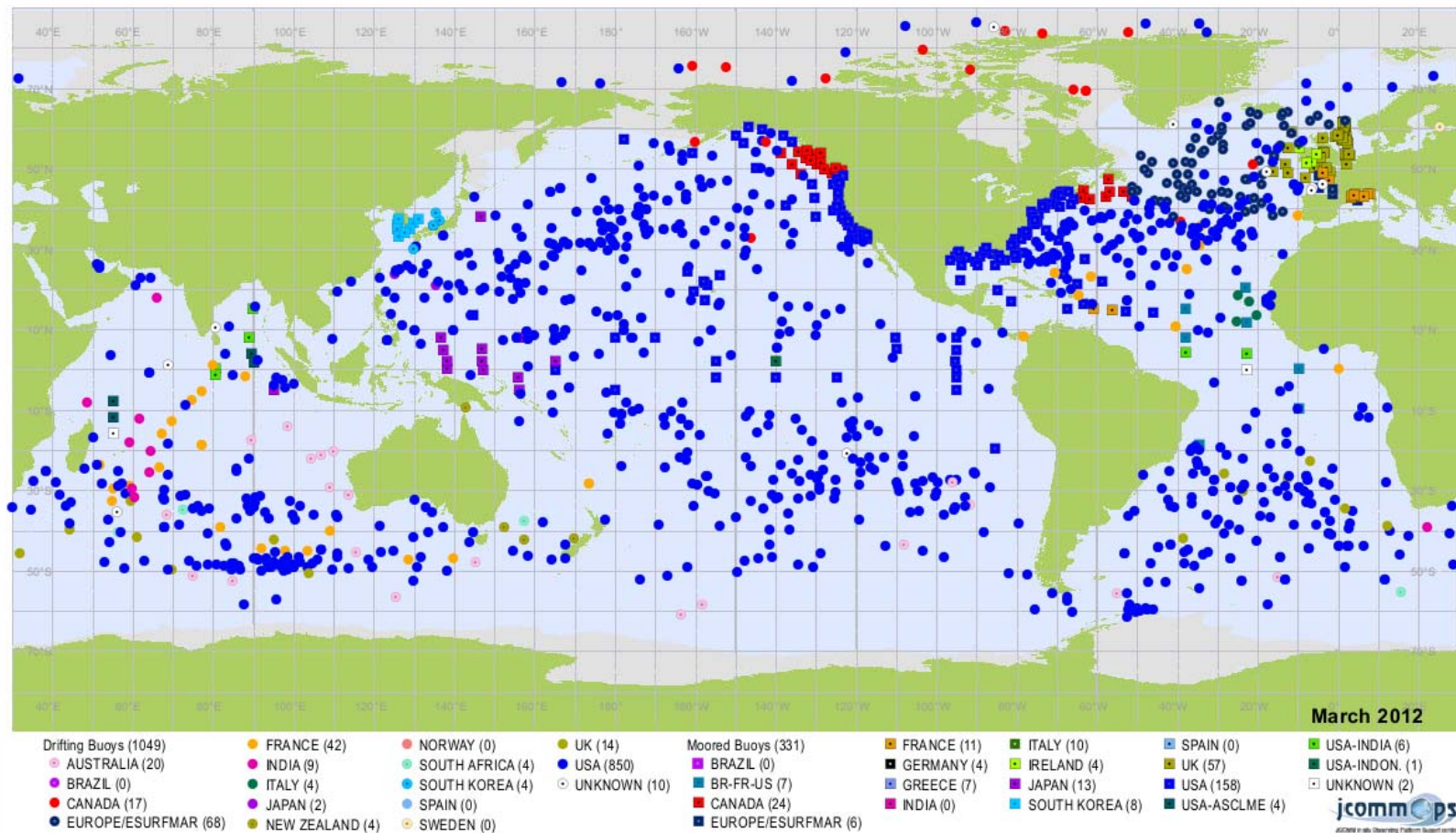
METEO-FRANCE couverture de donnees - SYNOP/SHIP - 2012/05/02 00H UTC cut-off long
Nombre total d'observations apres screening : 18719



STATUS OF GLOBAL DRIFTER ARRAY



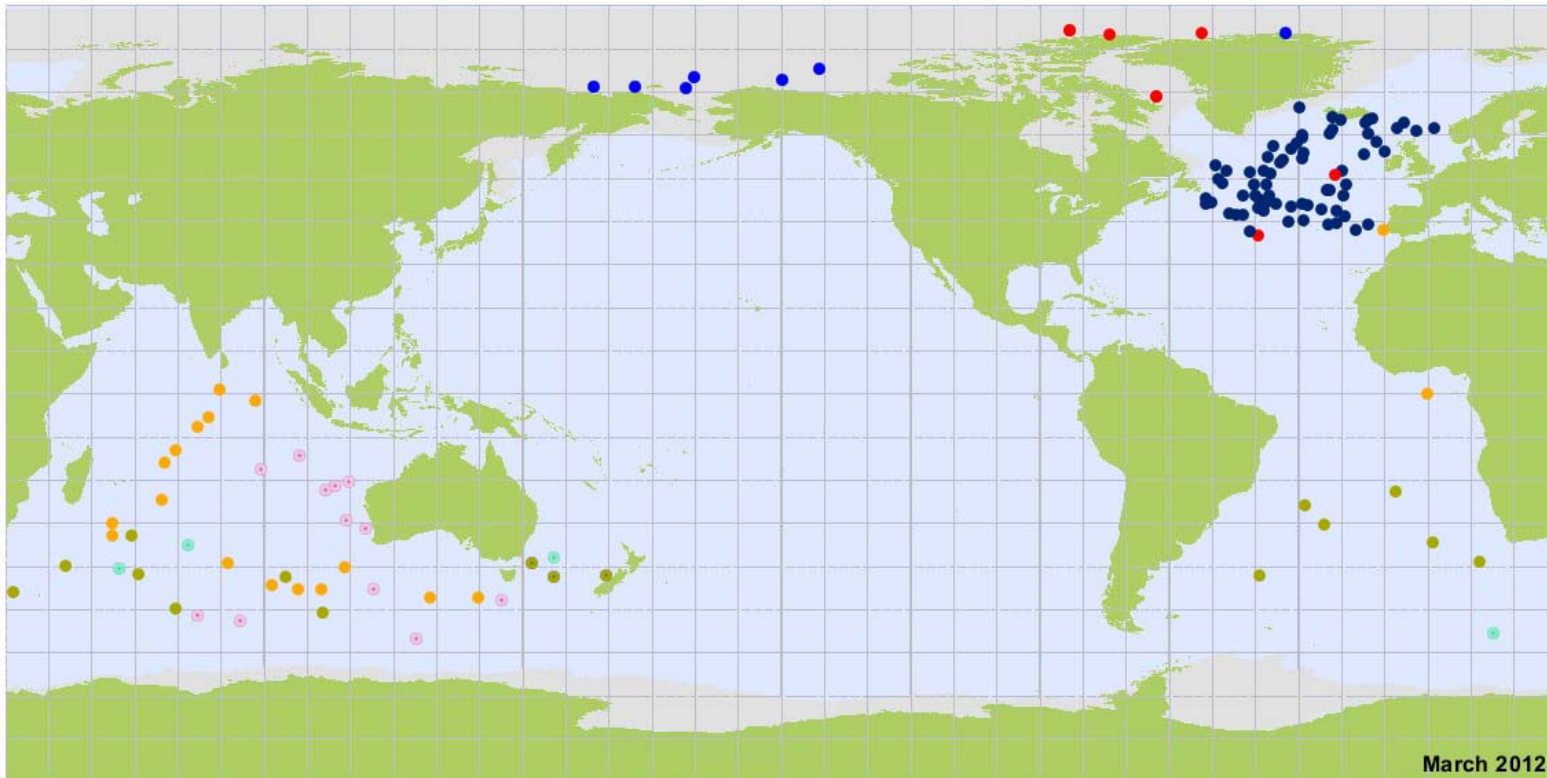
Global network of ocean buoys



● Drifting buoys (1049)

■ Moored buoys (331)

Improved transmission mode (133 buoys)



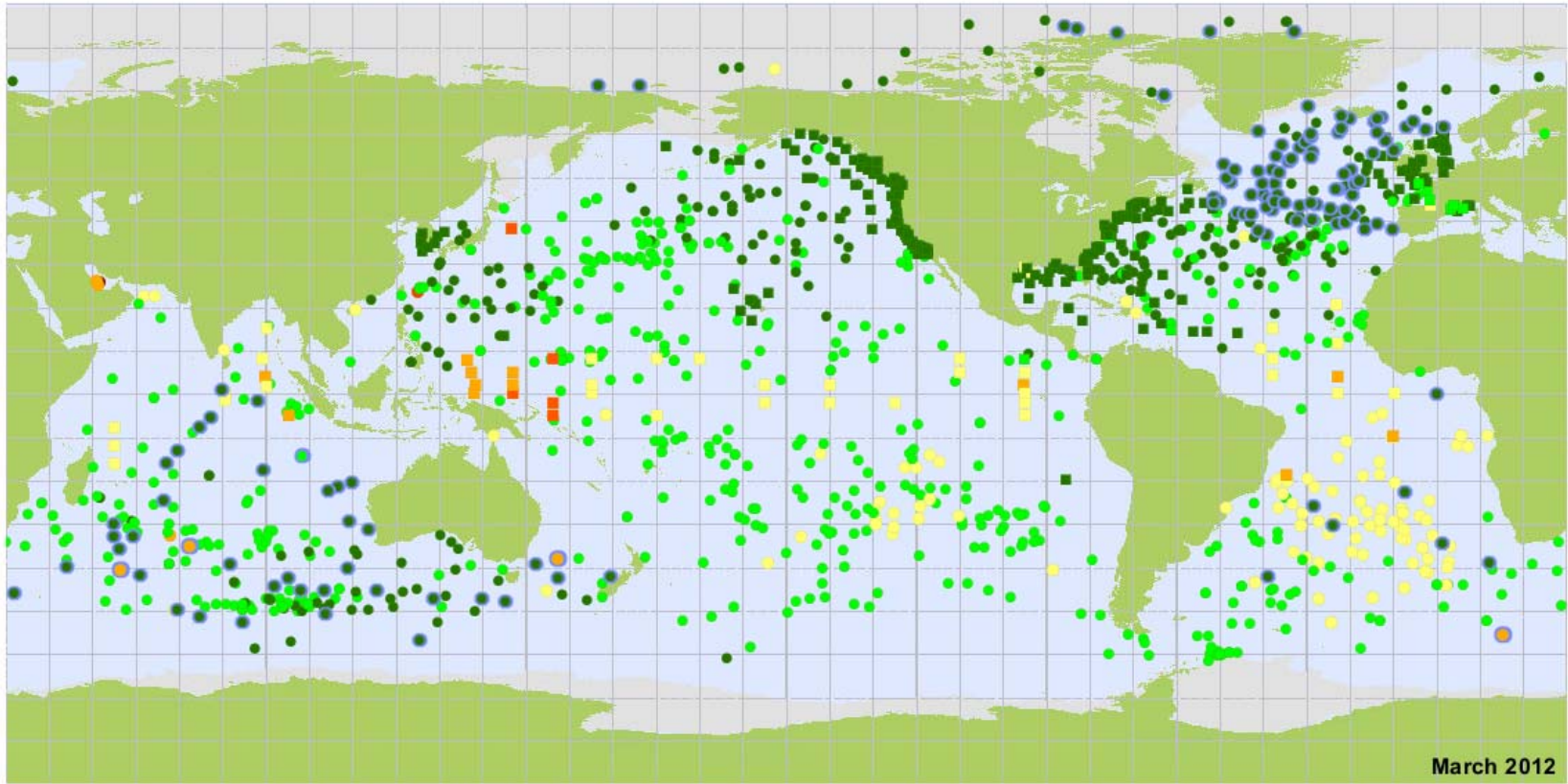
DBCP Pilot Projects (Iridium and Argos-3 Drifters by Country)

Iridium Drifting Buoys (133)

- | | | | |
|------------------|------------------------|--------------------|-----------|
| ● CANADA (8) | ● FRANCE (18) | ● SOUTH AFRICA (4) | ● USA (7) |
| ● AUSTRALIA (12) | ● EUROPE/ESURFMAR (68) | ● NEW ZEALAND (3) | ● UK (13) |



Retransmission delays



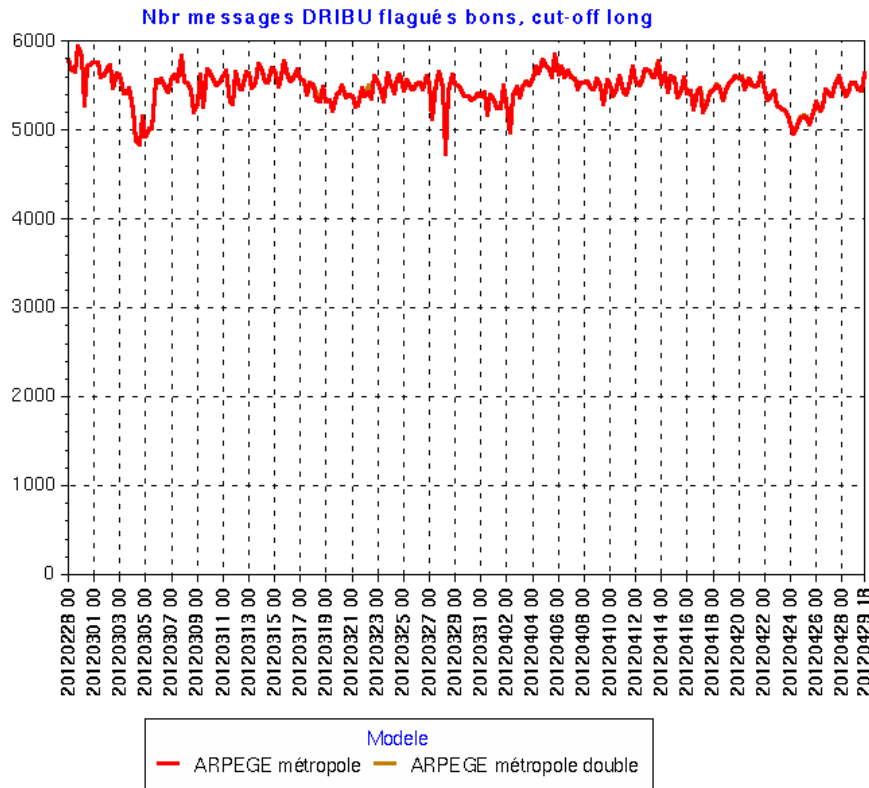
March 2012

Average Delay (Minutes) ● < 60 ● 61 - 120 ● 121 - 180 ● 181 - 240 ● 241 - 300 ● > 300



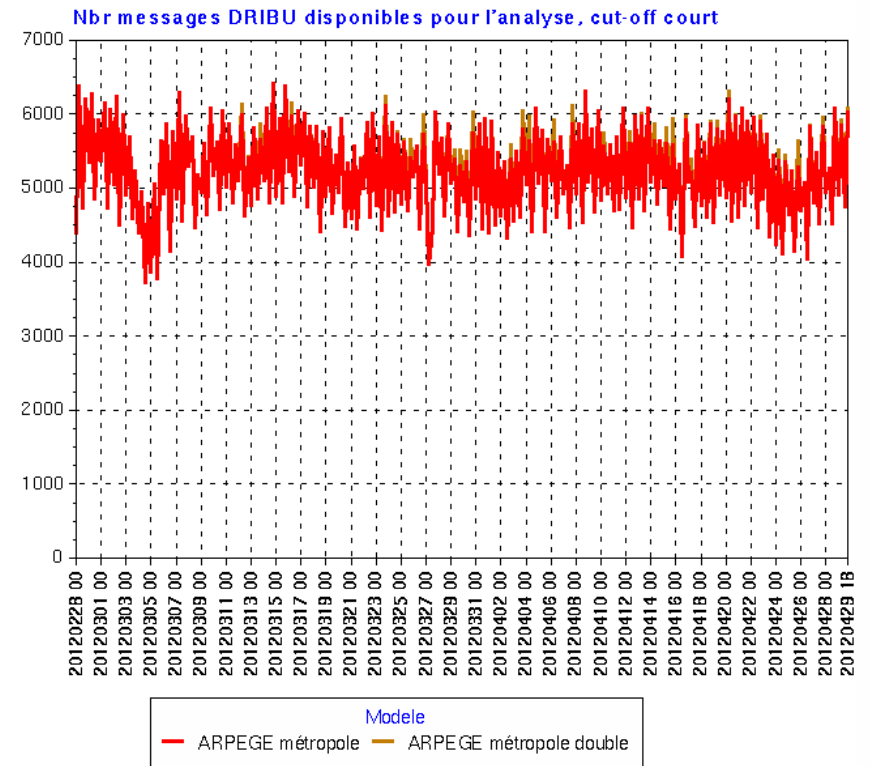
Number of « good quality » messages

DPrévilCOMPAS
30/04/12



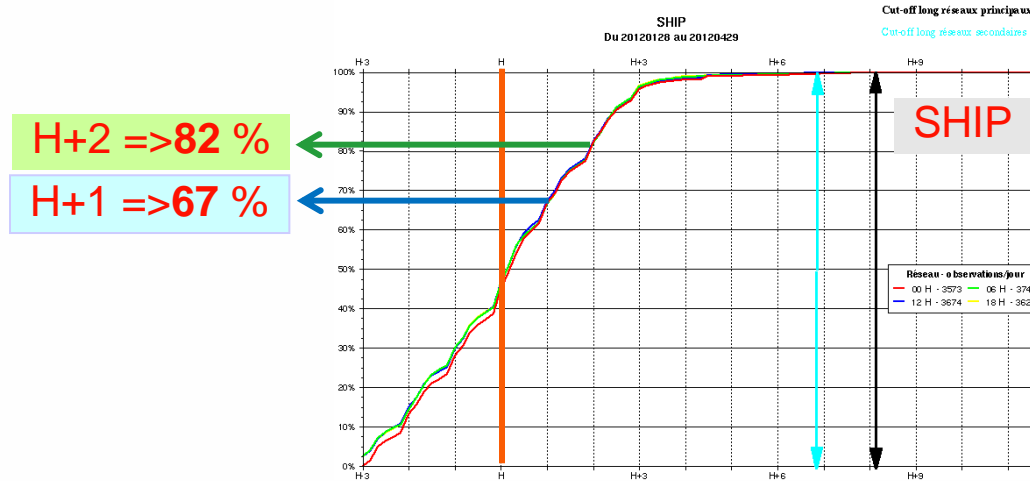
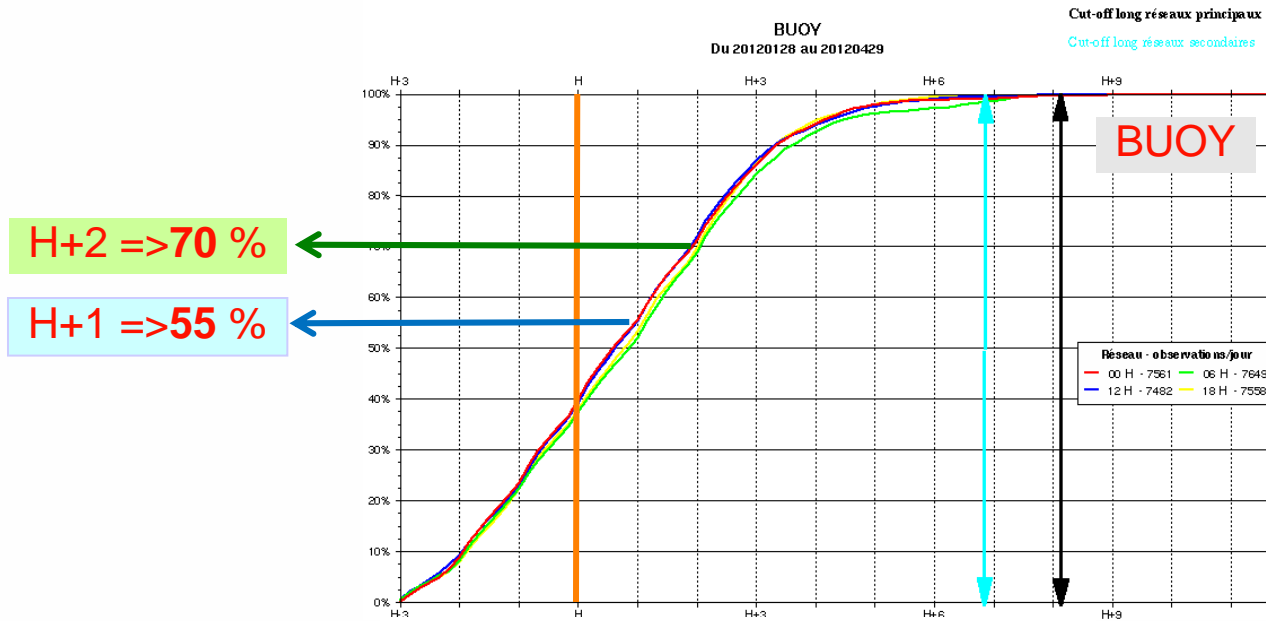
Long cut-off

DPrévilCOMPAS
30/04/12



Short cut-off

Timeliness of observations



Comparison of BUOY data with other data types

Wind intensity at 10 m : comparison against ARPEGE 6h-forecasts (FG) and analyses (AN) – 31 days (20/08/2011 – 19/09/2011)

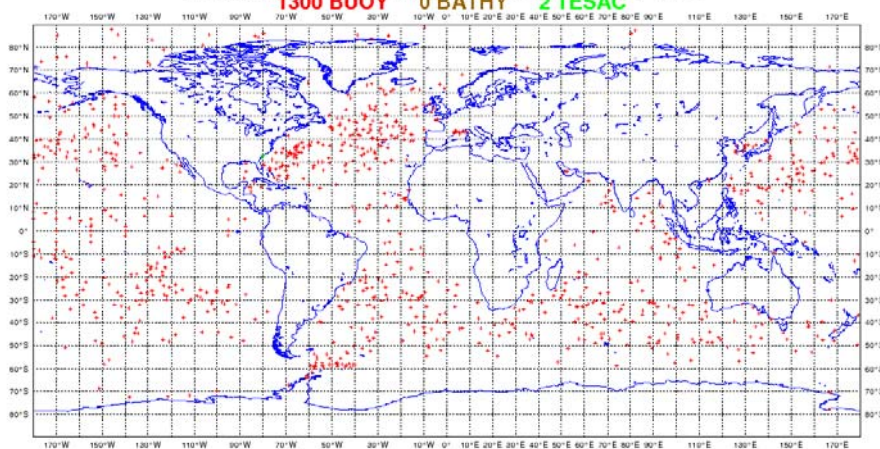
Data type	BUOY	ASCAT	SHIP
Number	43159	1118889	225564
Bias (O-FG)	-0.32	-0.08	-0.15
Std (O-FG)	1.68	1.27	2.4
Bias (O-AN)	-0.16	0.04	-0.13
Std (O-AN)	1.34	0.73	2.0

Use of BUOY observations for surface analysis

Before screening : 1302

After screening : 866 (66 %)

METEO-FRANCE couverture de donnees - BUOY - 2012/04/30 00H UTC cut-off long
Nombre total d'observations avant screening : 1302



BUOYS = 6.4 %

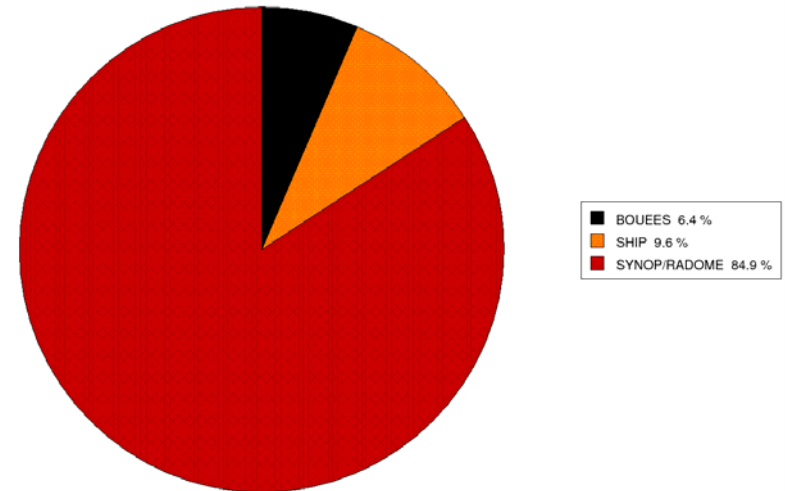
SHIP = 9.4 %

SYNOP = 84.9 %

Global OI analysis CANARI

SST, T2m, RH2m

Proportions de nombre d'observations utilisées par type d'obs - analyses de surface
cut-off long ARPEGE métropole
cumul mensuel de nombre d'observations utilisées pour 201203 : 2006040

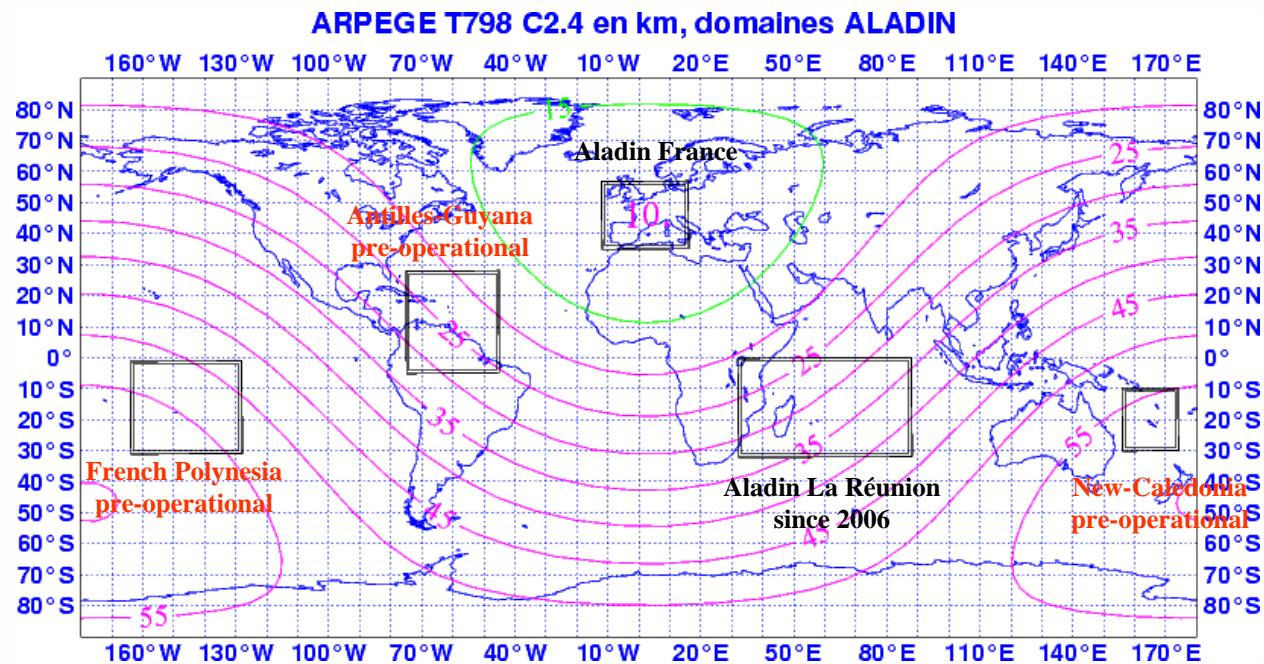




The limited area models ALADIN

Regional model ALADIN

- Spectral limited area model : E199x199L70
 - 70 levels from 17m to 0.05 hPa, horizontal resolution 7.5 km
- 3D-Var assimilation (6h window) :
 - Same data as ARPEGE plus SEVIRI radiances (ALADIN Réunion)
- Current operational domains :

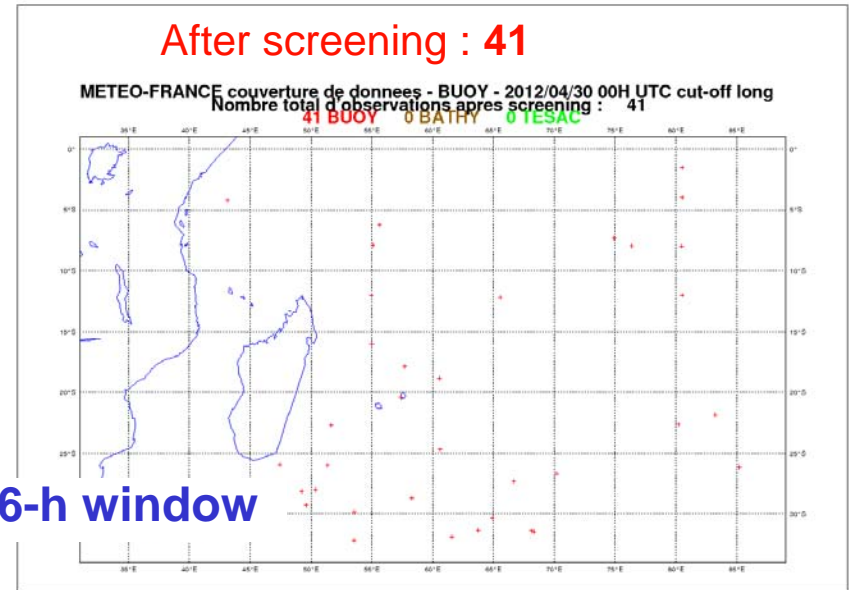
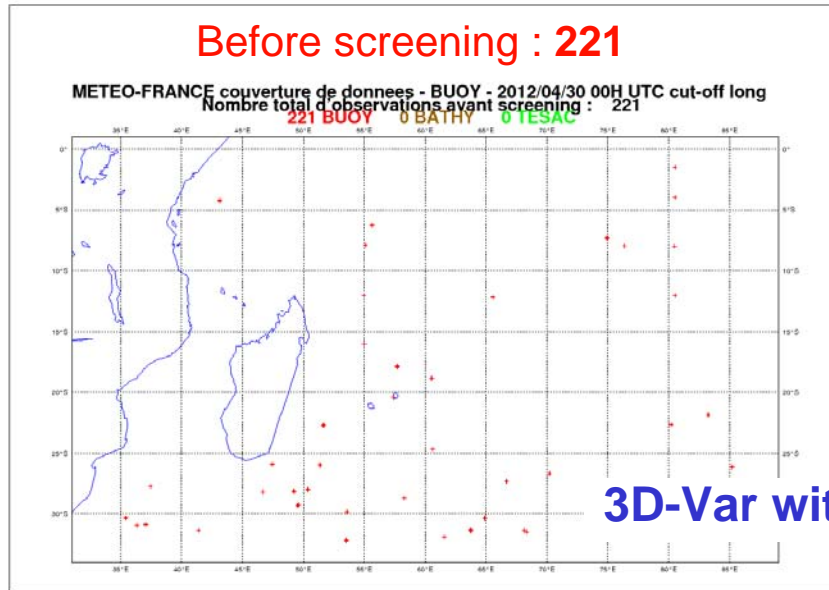


Forecast ranges and cut-offs:

FC+54 (00 UTC) [2h15], FC+48 (06 UTC) [3h],

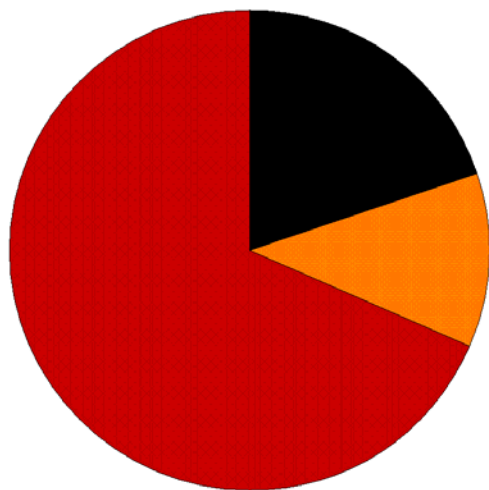
FC+42 (12 UTC) [1h50], FC+36 (18 UTC) [3h]

Use of BUOY observations in ALADIN Réunion



3D-Var with 6-h window

Proportions de nombre d'observations utilisées par type d'obs - analyses de surface
 ALADIN Réunion
 cumul mensuel de nombre d'observations utilisées pour 201203 : 24673



BUOYS = 19.9 %

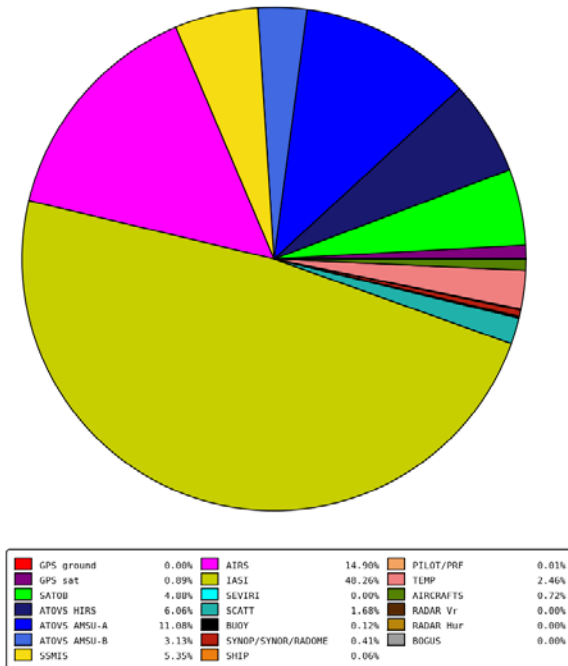
SHIP = 11.5 %

SYNOP = 68.5 %

OI regional surface analysis

Information content of observations in ALADIN Réunion

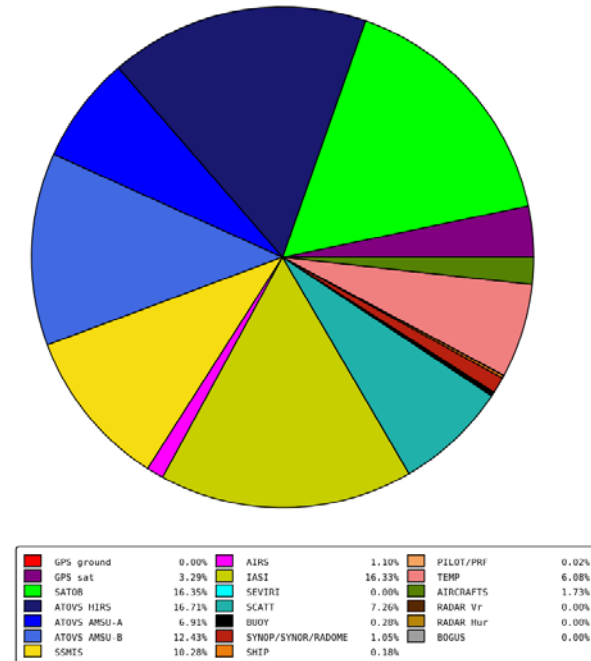
Proportions des nombres d'observations utilisées par type d'obs
analyses cut-off long - ALADIN Reunion
observations conventionnelles et satellites
cumul du nombre d'observations utilisées sur la période 2011110300 - 2011110318 : 199666



Number of observations

BUOYS=0.12 %

Part des DFS par type d'obs
analyses cut off long - ALADIN Reunion
observations conventionnelles et satellites
cumul du DFS sur la période 2011110300 - 2011110318 : 19798



Degrees of Freedom for Signal

BUOYS=0.28 %

ALADIN Réunion vs ARPEGE : larger fraction of oceans
but less used data (3D-Var vs. 4D-Var)



Forecast sensitivity to observations with ARPEGE

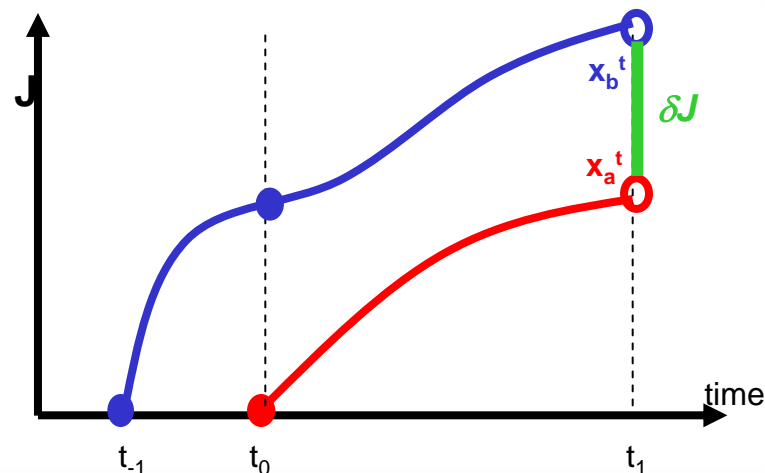
Linear estimate of observation impact

How?

- Implemented in IFS (ECMWF) by C. Cardinali and M. Fisher.
- J : short-range forecast error = difference between a 24h forecast and an analysis (dry energy norm)
- Observation impact:

$$\delta J = J(x_b^t) - J(x_a^t)$$

$$\delta J \approx \frac{\partial J}{\partial y} \delta y = \left(\frac{\partial x_a}{\partial y} \right) \left(\frac{\partial J}{\partial x_a} \right) (y - Hx_b)$$



$$\left(\frac{\partial J}{\partial x_a} \right) = \frac{1}{2} \left[\frac{\partial J(x_a^t)}{\partial x_a} + \frac{\partial J(x_b^t)}{\partial x_b} \right]$$

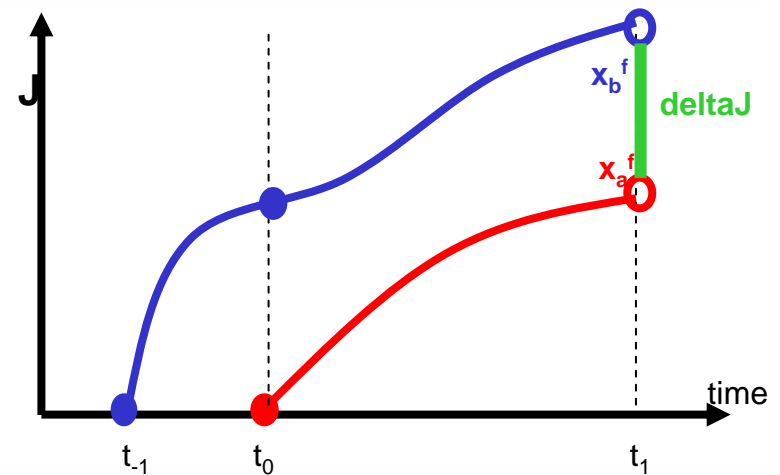
$$\left(\frac{\partial x_a}{\partial y} \right) = \mathbf{K}^T = \mathbf{R}^{-1} \mathbf{H} \mathbf{A}$$

- Contribution of individual observations to the reduction of J

Linear estimate of observation impact

How?

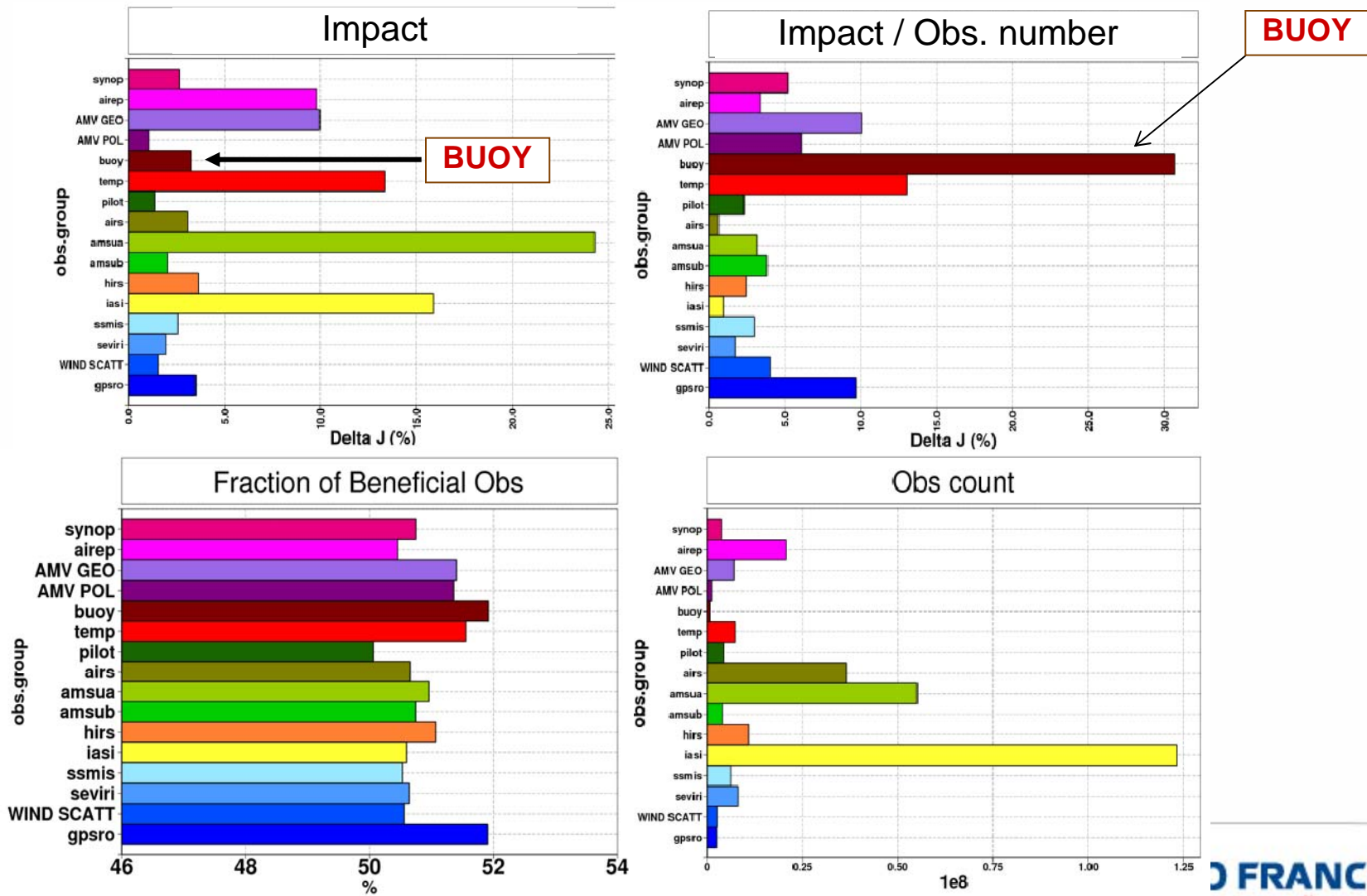
- Implemented in IFS (ECMWF) by C. Cardinali and M. Fisher.
- J : 3D integrated dry total energy of the difference between the 24h forecast and a reference state
- Observation impact:

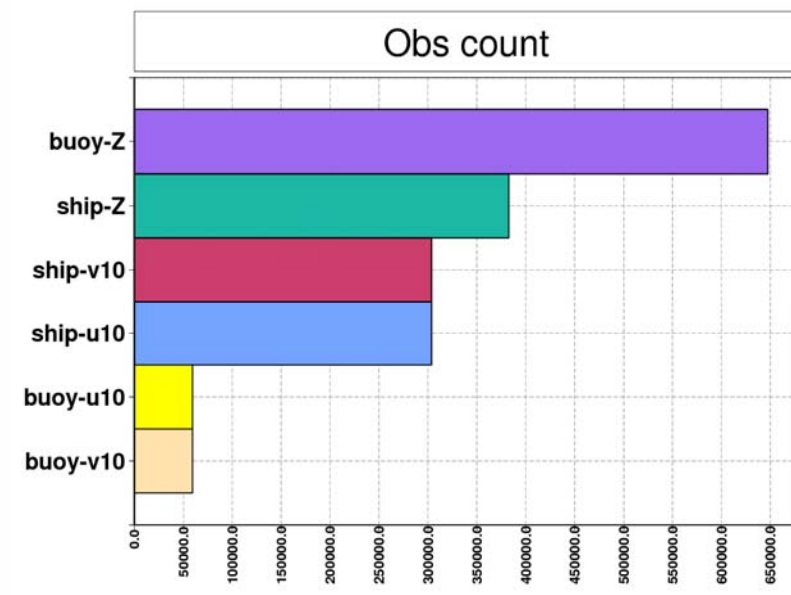
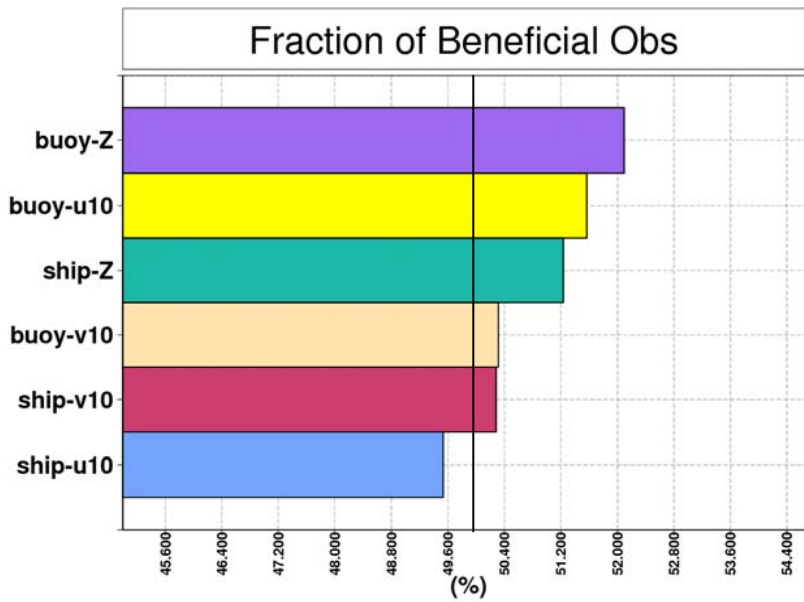
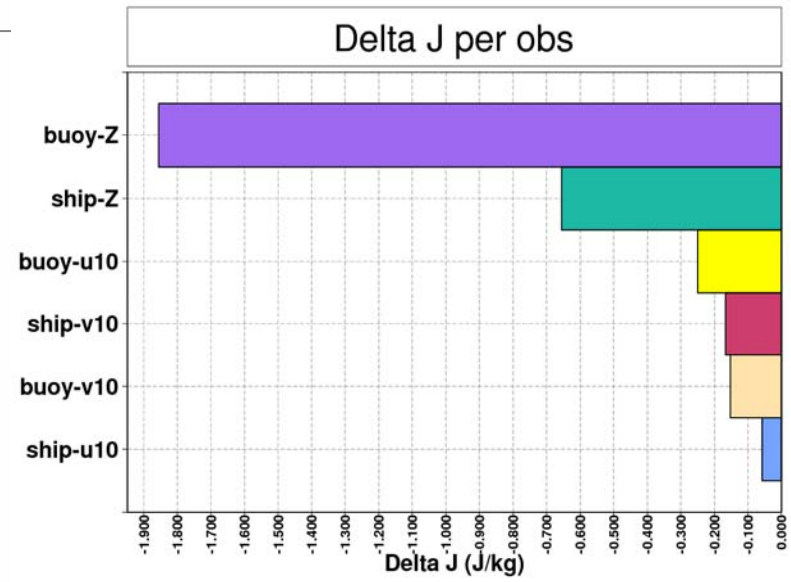
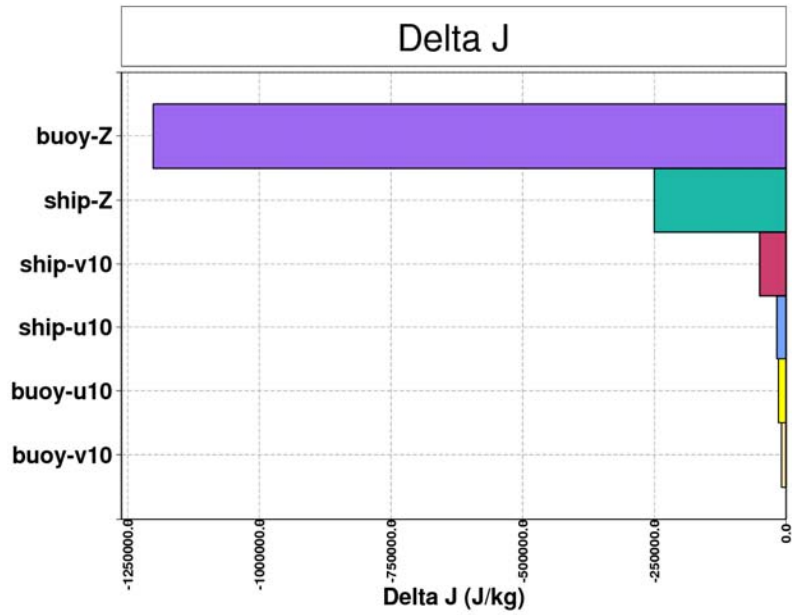


$$\delta J = \frac{1}{2} (\mathbf{R}^{-1} \mathbf{H} \mathbf{A}) \left(\mathbf{M}_a^T \frac{\partial J^b}{\partial x_b^f} + \mathbf{M}_b^T \frac{\partial J^a}{\partial x_a^f} \right) (y - H x_b)$$

- second order approximation (Errico, 2007).

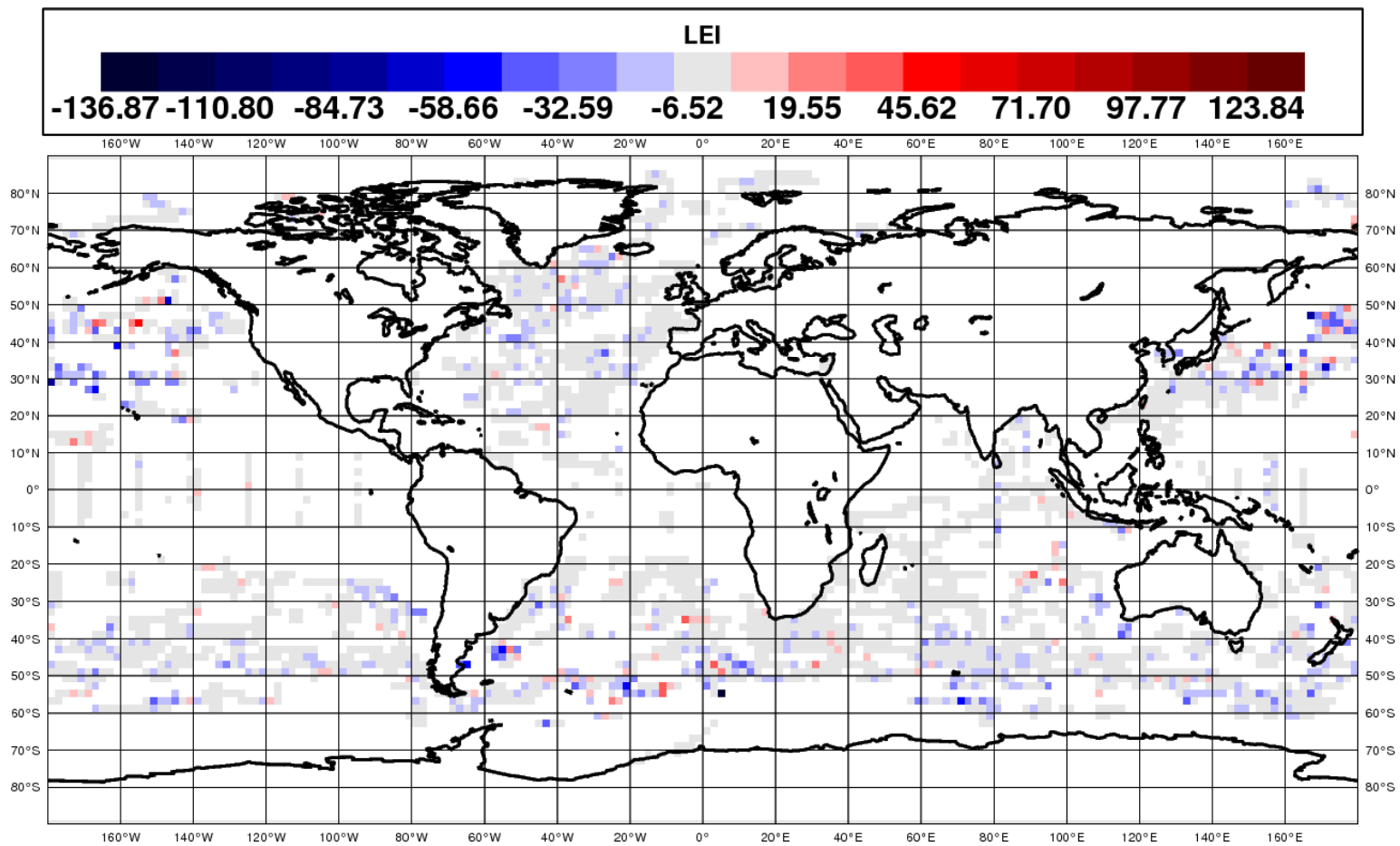
Forecast impact experiment from Dec. 2010 to Jan. 2011





BUOY all parameters

Averaged Linear Estimate of Impact (LEI): -6806.31 J/kg
Experiment: B281 / From 2010/12/02 at 00UTC to 2011/01/15 at 18UTC / H UTC cycles / Grid: 2° × 2° .
Objective function : Total energy of the 24h forecast error.
Observation type: buoy, parameter: All, level : all



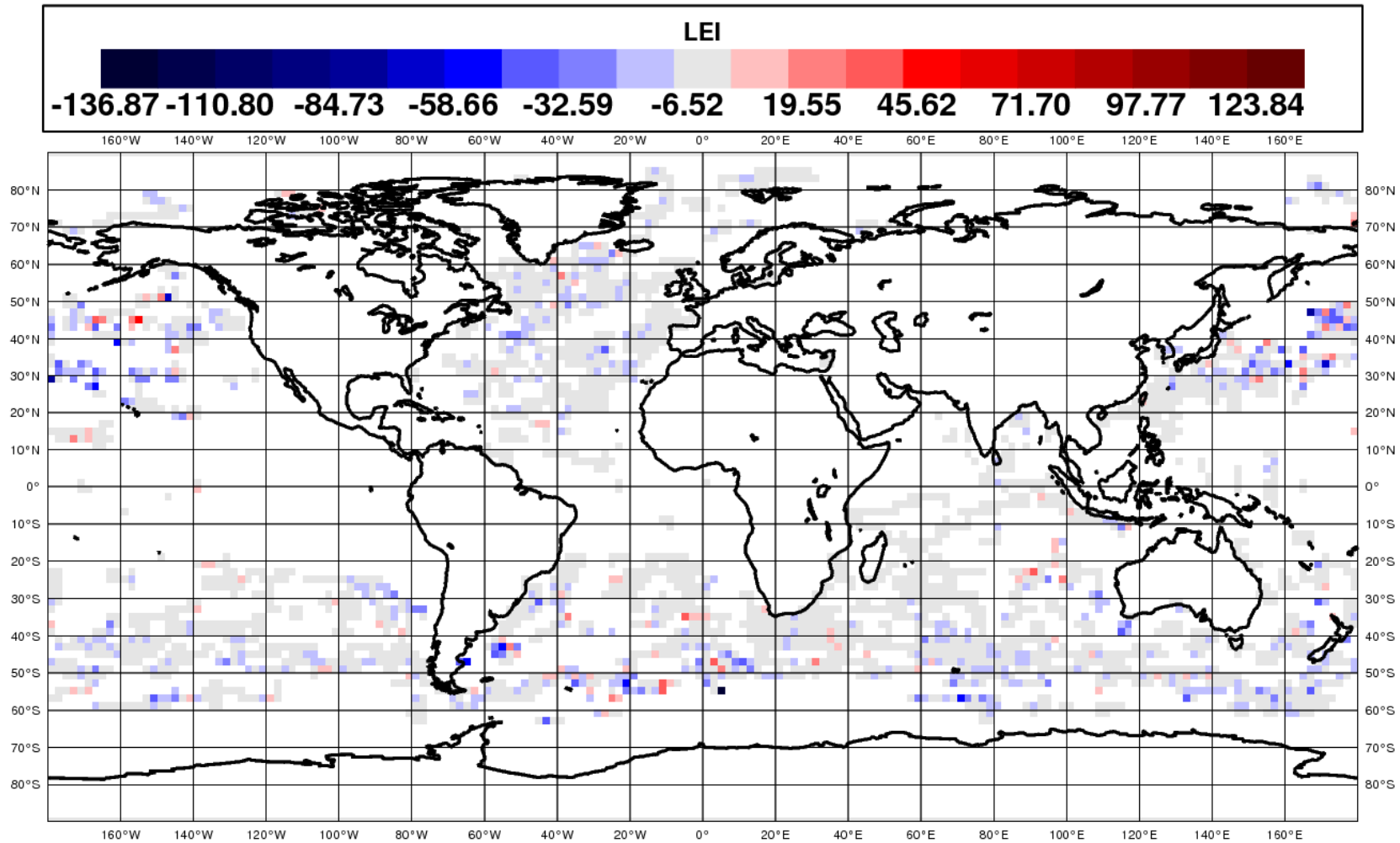
BUOY geopotential

Averaged Linear Estimate of Impact (LEI) **-6675.23 J/kg**

Experiment: B281 / From 2010/12/02 at 00UTC to 2011/01/15 at 18UTC / H UTC cycles / Grid: 2° × 2°.

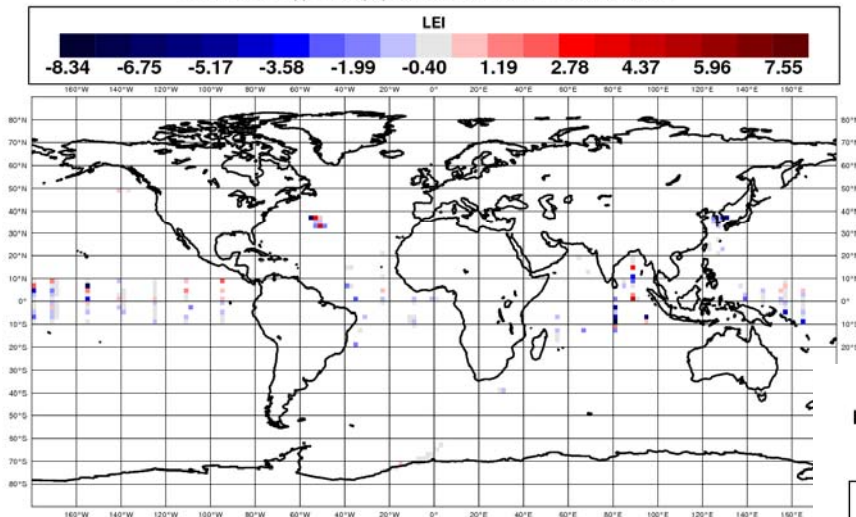
Objective function : Total energy of the 24h forecast error.

Observation type: buoy, parameter: Geopotential, level : all



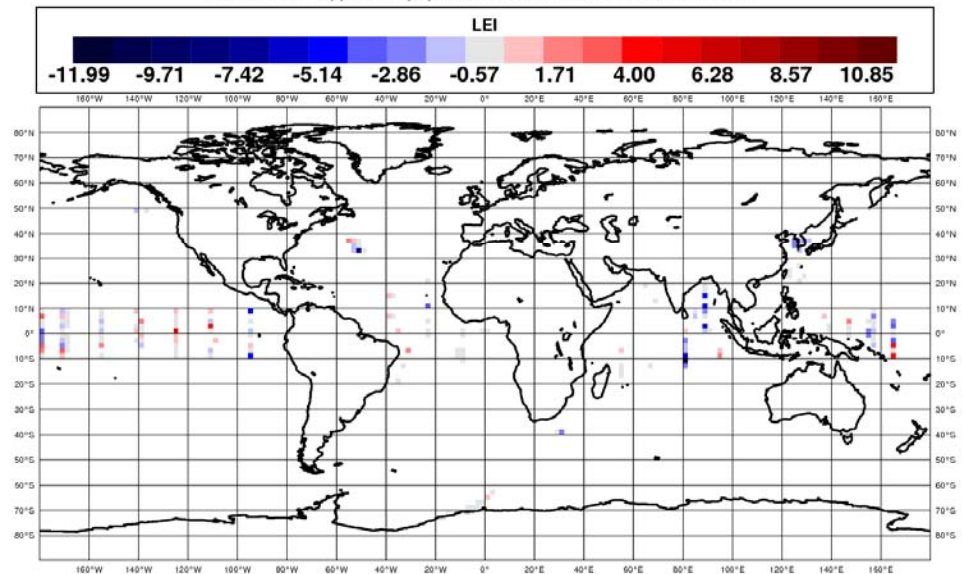
BUOY wind

Averaged Linear Estimate of Impact (LEI): -81.6788 J/kg
 Experiment: B281 / From 2010/12/02 at 00UTC to 2011/01/15 at 18UTC / H UTC cycles / Grid: 2° × 2°.
 Objective function : Total energy of the 24h forecast error.
 Observation type: buoy, parameter: Surface U wind, level : all



U component
 $dJ = -61.76 \text{ J/kg}$

Averaged Linear Estimate of Impact (LEI): -49.4005 J/kg
 Experiment: B281 / From 2010/12/02 at 00UTC to 2011/01/15 at 18UTC / H UTC cycles / Grid: 2° × 2°.
 Objective function : Total energy of the 24h forecast error.
 Observation type: buoy, parameter: Surface V wind, level : all



V component
 $dJ = -49.40 \text{ J/kg}$

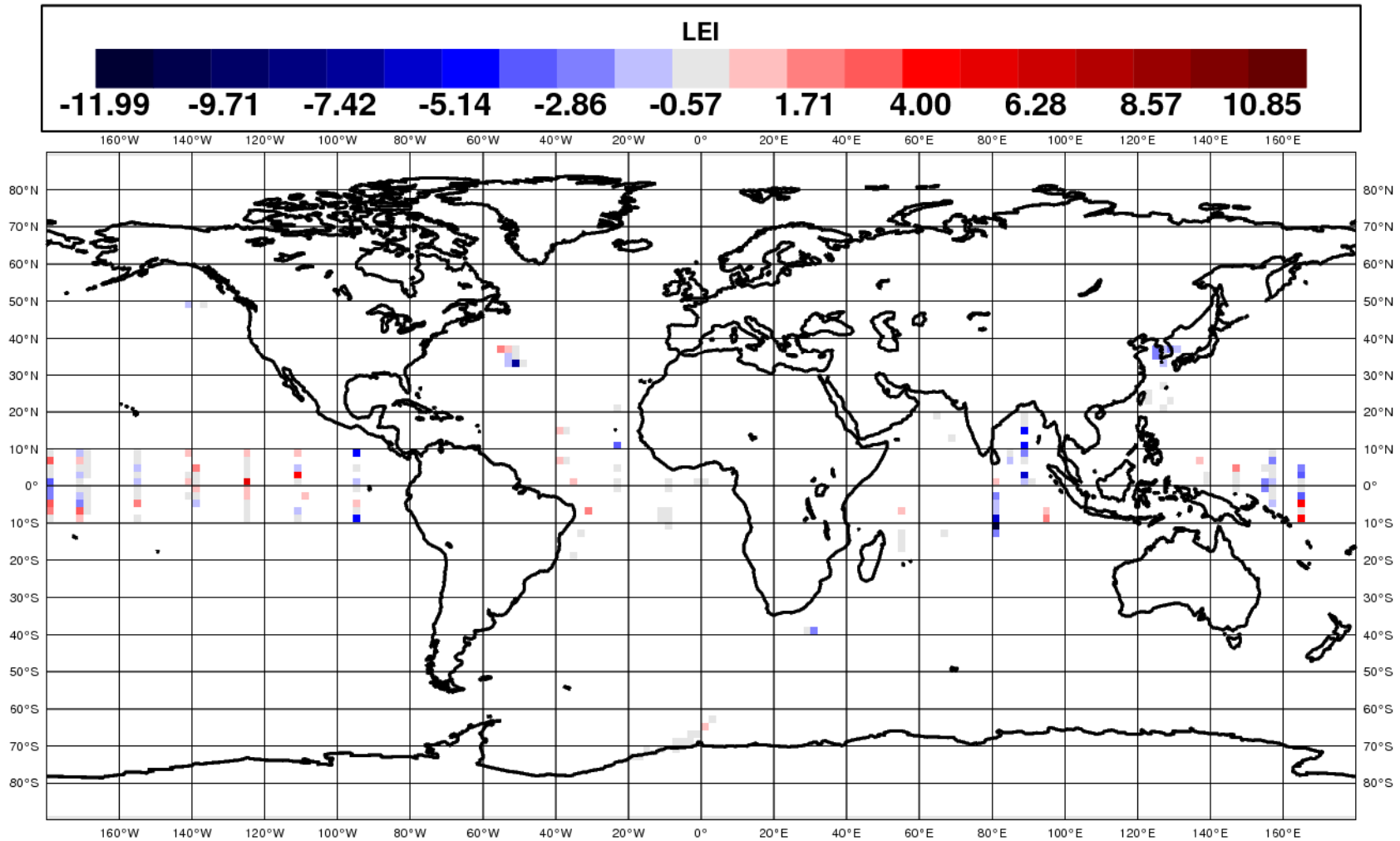
Buoy v wind

Averaged Linear Estimate of Impact (LEI): -49.4005 J/kg

Experiment: B281 / From 2010/12/02 at 00UTC to 2011/01/15 at 18UTC / H UTC cycles / Grid: 2° × 2°.

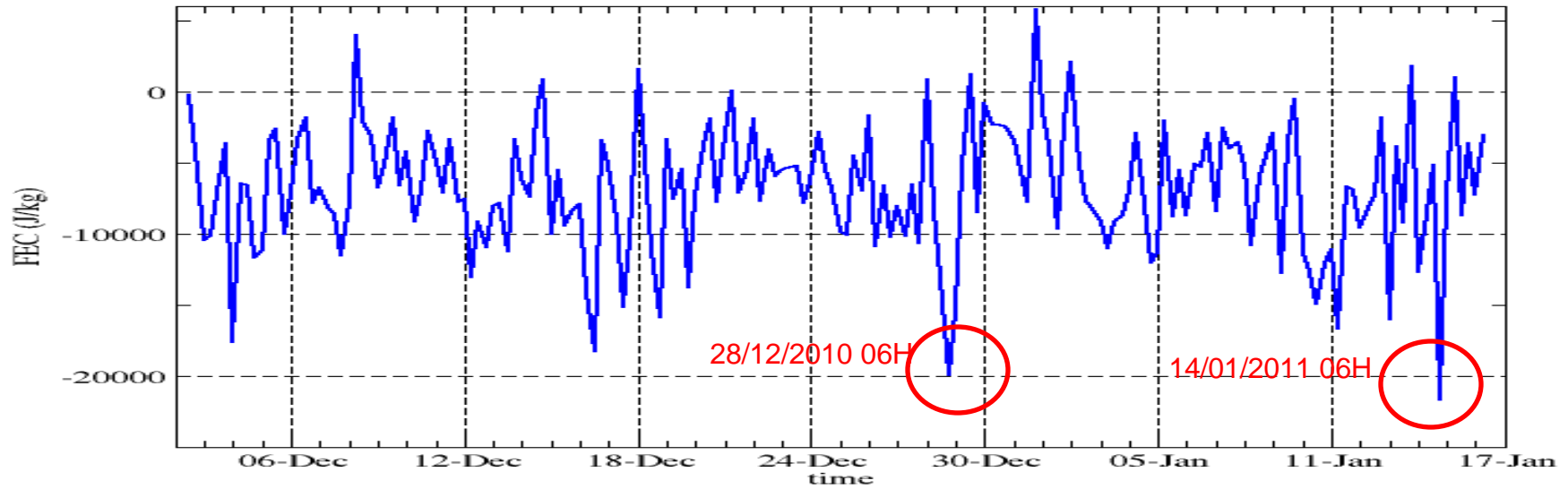
Objective function : Total energy of the 24h forecast error.

Observation type: buoy, parameter: Surface V wind, level : all

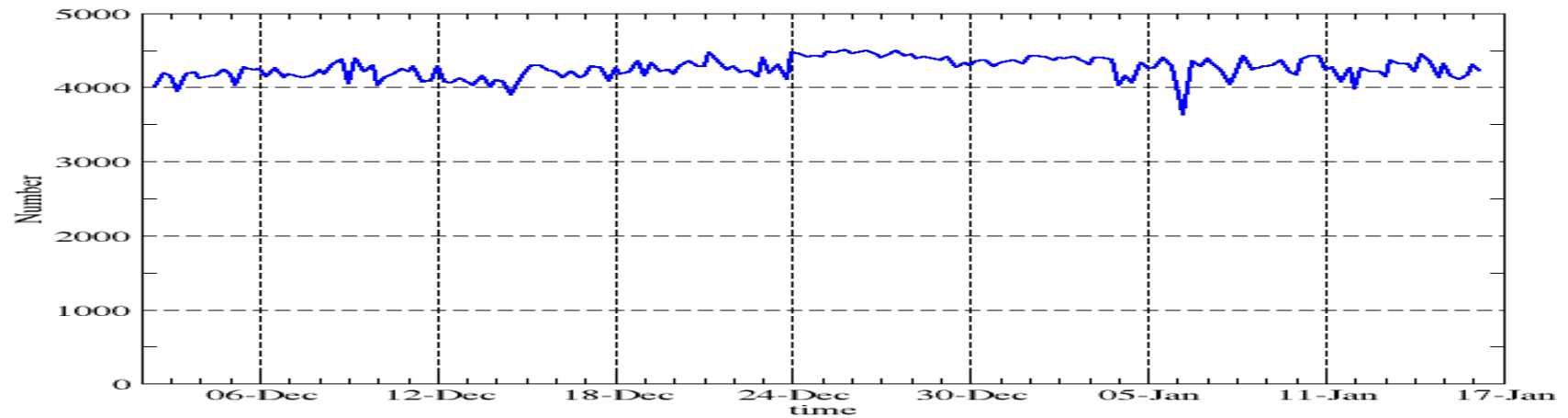


Buoy time series

FEC timeserie for Buoys



Number of obervations timeserie for Buoys



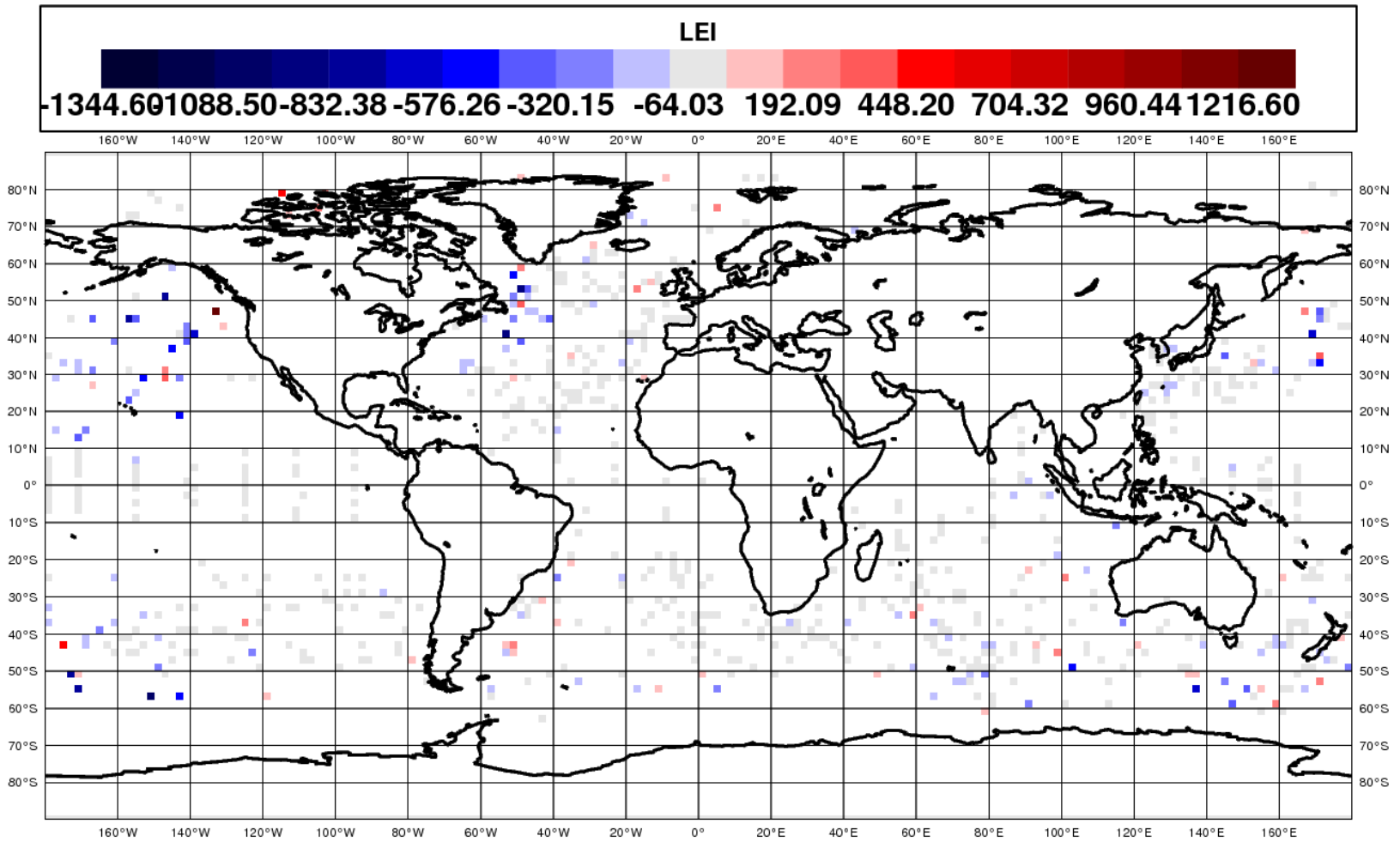
Buoy on the 28/12/2010 06H UTC

Averaged Linear Estimate of Impact (LEI): -19926.1 J/kg

Experiment: B281 / From 2010/12/28 at 06UTC to 2010/12/28 at 06UTC / H UTC cycles / Grid: 2° × 2°.

Objective function : Total energy of the 24h forecast error.

Observation type: buoy, parameter: All, level : all



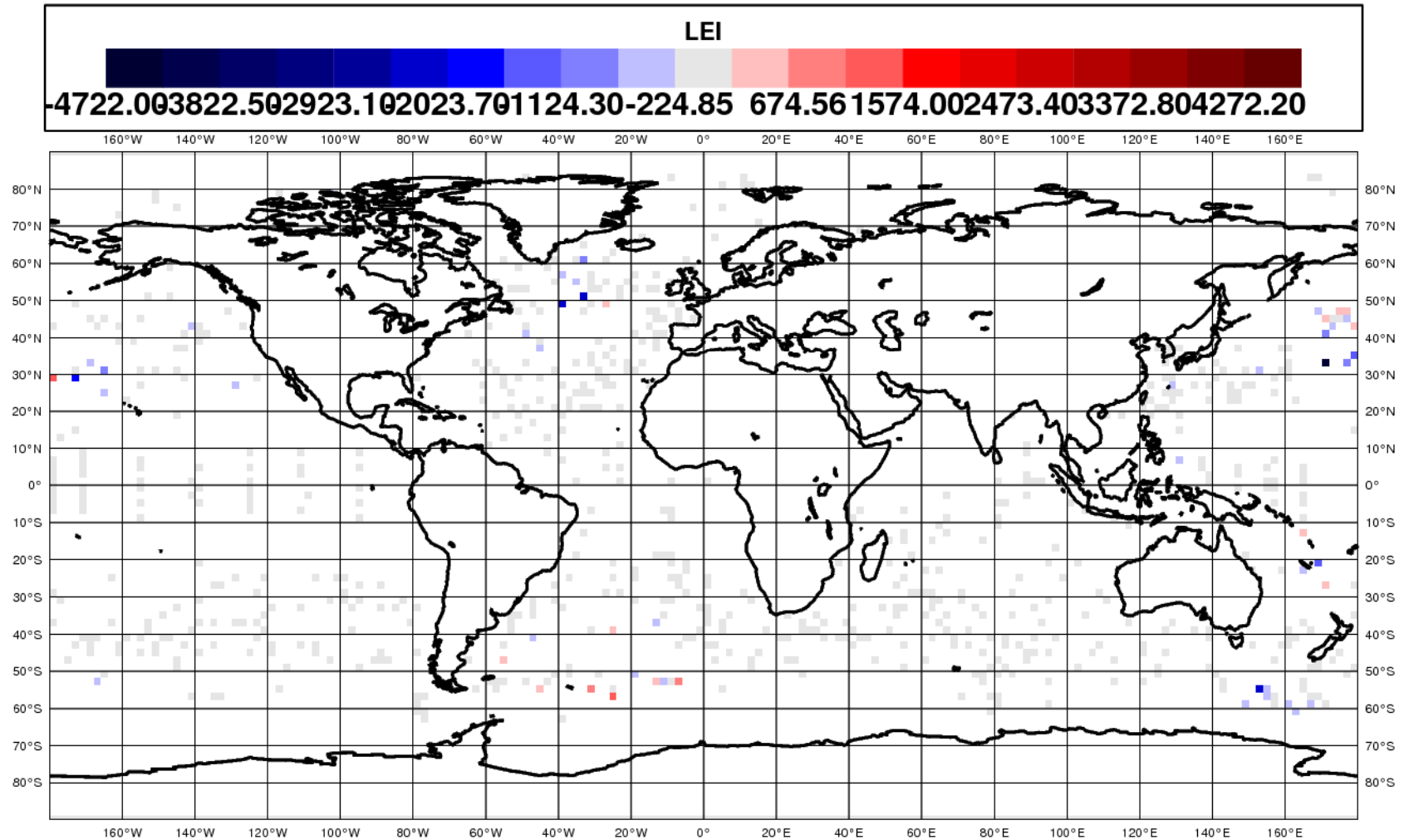
Buoy on the 14/01/2011 06H UTC

Averaged Linear Estimate of Impact (LEI): -21635.7 J/kg

Experiment: B281 / From 2011/01/14 at 06UTC to 2011/01/14 at 06UTC / H UTC cycles / Grid: 2° × 2°.

Objective function : Total energy of the 24h forecast error.

Observation type: buoy, parameter: All, level : all





Conclusions

- BUOY observations are used operationally at Météo-France for upper air and surface analyses (global and LAMs)
- Small fraction of observations assimilated in Météo-France models
- Significant fraction of conventional observations over oceans (for surface pressure)
- Wind measurements from BUOY are more accurate than SHIP but less accurate than ASCAT
- Relative information content (DFS) larger than the percentage of observations (lower for most satellite data)
- Largest forecast sensitivity to observations (when normalised by the amount of data) => each individual observation is very valuable to reduce short range forecast errors (surface pressure)
- Timeliness of BUOY data could be improved (one hour lag compared to other surface data)



Thank you for
your attention