# Diagnosing the impact of satellite observations within data assimilation

#### **Peter Bauer**

#### Acknowledgements:

William Bell, Niels Bormann, Carla Cardinali, Andrew Collard, Mohamed Dahoui, Ron Gelaro, Sean Healy, Graeme Kelly, Philippe Lopez, Tony McNally, Carole Peubey, Mark Rodwell, Graeme Stephens, David Tan, Jean-Noël Thépaut

European Centre for Medium-Range Weather Forecasts Reading, UK



## Satellite data usage at ECMWF

Impact of existing satellite data

- Data monitoring
- Observing system experiments
- Advanced diagnostic tools

Impact of future satellite data

Summary



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#### Data sources: Conventional

#### SYNOP/SHIP/METAR:

- Meteorological/aeronautical land surface weather stations (2m-temperature, dew-point temperature, 10m-wind)
- Ships
- → temperature, dew-point temperature, wind (land: 2m, ships: 25m)

#### **BUOYS:**

- Moored buoys (TAO, PIRATA)
- Drifters
- → temperature, pressure, wind

#### TEMP/TEMPSHIP/DROPSONDES:

- Radiosondes
- ASAPs (commercial ships replacing stationary weather ships)
- Dropsondes released from aircrafts (NOAA, Met Office, tropical cyclones, experimental field campaigns, e.g., FASTEX, NORPEX)
- → temperature, humidity, pressure, wind *profiles*

#### **PROFILERS:**

- UHF/VHF Doppler radars (Europe, US, Japan)
- → wind *profiles*

#### Aircraft:

- AIREPS (manual reports from pilots)
- AMDARs, ACARs, etc. (automated readings)
- → temperature, pressure, wind *profiles*



#### Data sources: Satellites

#### **Radiances** ( $\rightarrow$ brightness temperature = level 1):

- AMSU-A on NOAA-15/18/19, AQUA, Metop
- AMSU-B/MHS on NOAA-17/18, Metop
- SSM/I on F-13/15, AMSR-E on Aqua
- HIRS on NOAA-17/19, Metop
- AIRS on AQUA, IASI on Metop
- MVIRI on Meteosat-7, SEVIRI on Meteosat-9, GOES-11/12, MTSAT-1R imagers

#### **Ozone** ( $\rightarrow$ total column ozone = level 2):

Total column ozone from SBUV on NOAA-17/18, OMI on Aura

#### **Bending angles** ( $\rightarrow$ bending angle = level 1):

COSMIC (6 satellites), GRAS on Metop

#### **Atmospheric Motion Vectors** (→ wind speed = level 2):

Meteosat-7/9, GOES-11/12, MTSAT-1R, MODIS on Terra/Aqua

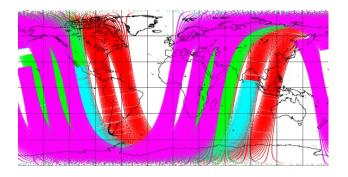
#### **Sea surface parameters** ( $\rightarrow$ wind speed and wave height = level 2):

- Significant wave height from Seawinds on QuikSCAT, AMI on ERS-2, ASCAT on Metop
- Near-surface wind speed from RA-2/ASAR on Envisat, Jason altimeter

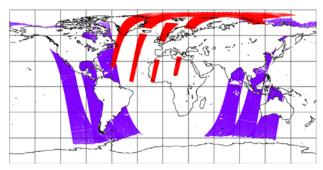


## Example of satellite data coverage

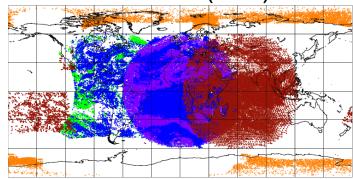
**LEO Sounders** 

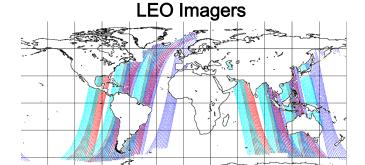


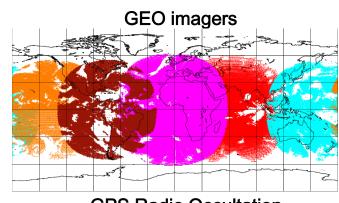
**Scatterometers** 

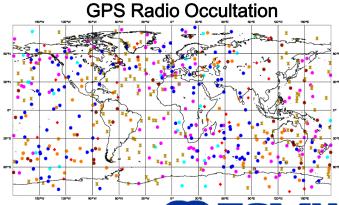


Satellite Winds (AMVs)



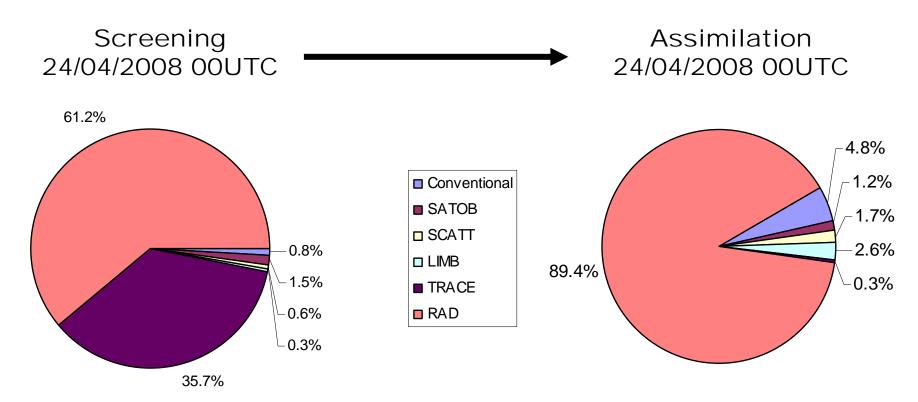








## Data types



- Satellite data amounts to 99% in screening and 95% in assimilation.
- Radiance data dominates assimilation with 90%.
- Relative GPSRO (limb) data amount strongly increases between screening and assimilation while ozone data is largely reduced.



#### Incremental 4D-Var

#### **Initial model state:**

$$\mathbf{x}_o = \mathbf{x}_{o,b} + \delta \mathbf{x}_o$$

Observation operator at time t<sub>i</sub>:

$$H_i(\mathbf{x}_i) = H_i(\mathbf{x}_{i,b}) + \mathbf{H}_i \delta \mathbf{x}_i, \qquad \delta \mathbf{x}_i = \mathbf{M}(\mathbf{x}_o, \mathbf{x}_i) \delta \mathbf{x}_o$$

#### **Cost-function gradient:**

$$\nabla_{\delta \mathbf{x}_o} J = \mathbf{B}^{-1} \delta \mathbf{x}_o + \sum_{i=0}^n \mathbf{M}^T (t_i, t_o) \mathbf{H}_i^T \mathbf{R}_i^{-1} (\mathbf{d}_i - \mathbf{H}_i \delta \mathbf{x}_i)$$

$$\mathbf{d}_{i} = \mathbf{y}_{o,i} - H(\mathbf{x}_{i,b})$$

All terms determine analysis impact of observations
⇒ Ideally, impact diagnostics perform evaluation for all terms

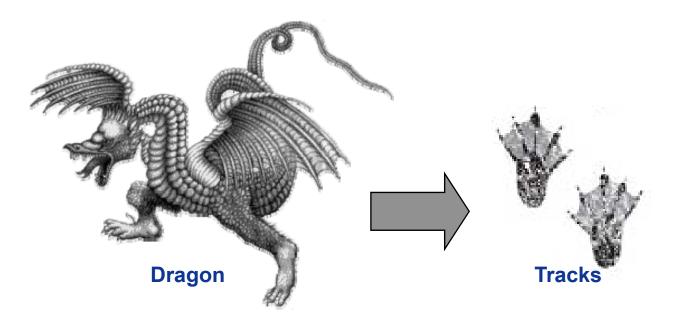
(P. Lopez)



## How is <u>radiance</u> information used?

Forward problem:

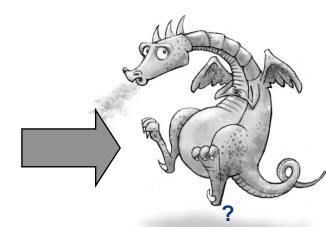
$$y = H(x_b)$$



Inverse problem:

$$\mathbf{x}_{\mathbf{a}} = H^{1} (\mathbf{y}_{\mathbf{o}})$$





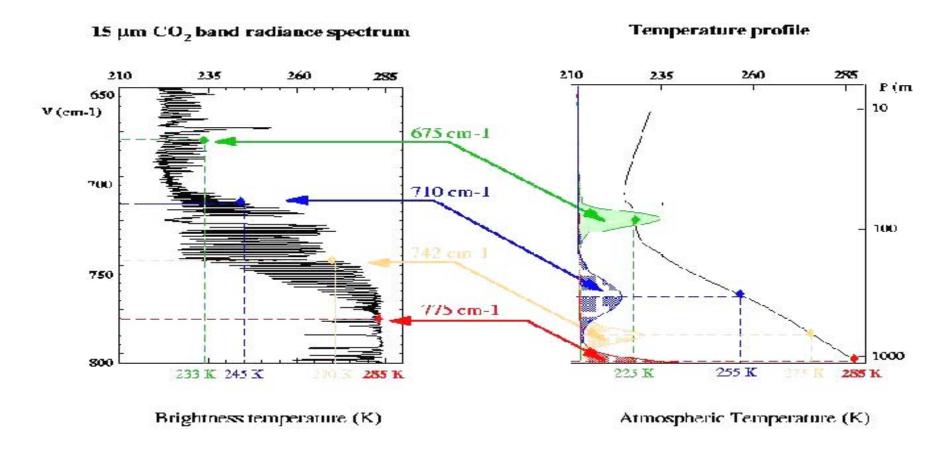
(G. Stephens)



## Forward/inverse modeling

 $y = H(x_b)$ : Forward (interpolation + radiative transfer)

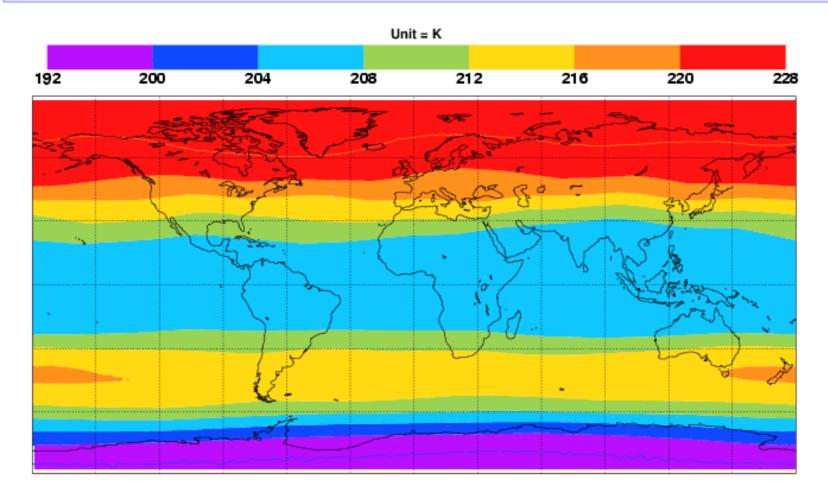
 $x_a = H^{-1}(y_o)$ : Inverse (interpolation + radiative transfer + a priori information)





### **AMSU-A observations**

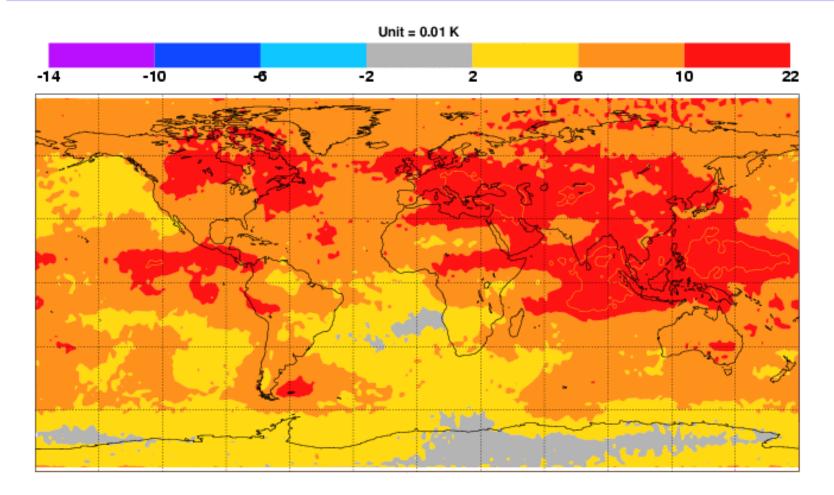
AMSUA ch 9 (~T100) OBS\_VAL, 20090401-20090815, Expver=43, Analysis=DCDA, Time=[00,12] Microwave brightness temperature, weighting function: 150 to 30 hPa





## AMSU-A observation-model

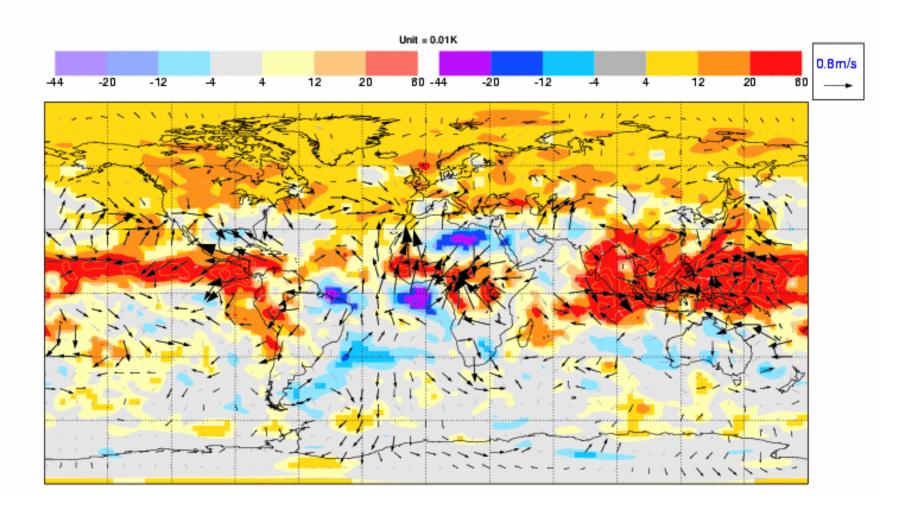
AMSUA ch 9 (~T100) FG\_DEP, 20090401-20090815, Expver=43, Analysis=DCDA, Time=[00,12] Microwave brightness temperature, weighting function: 150 to 30 hPa





## Mean T-increment

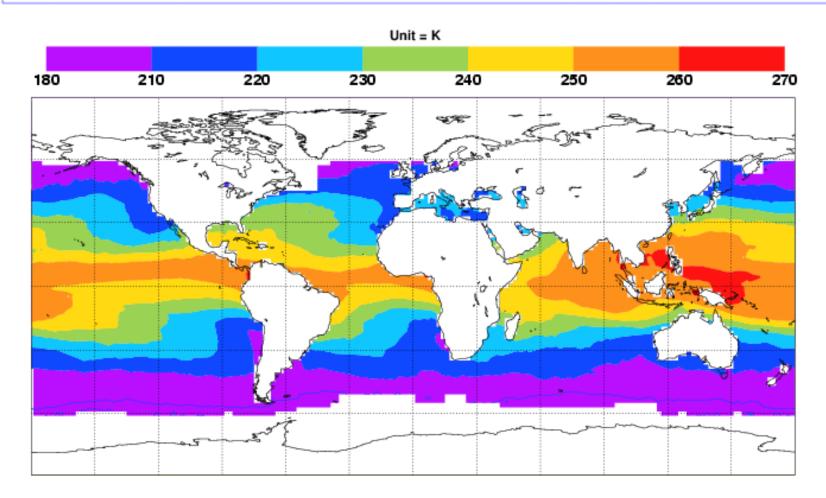
Analysis increments: Mean T100, 20090402-20090815, Expver=35R3 43, Analysis=dcda. Deep colours = 5% significance





## SSM/I observations

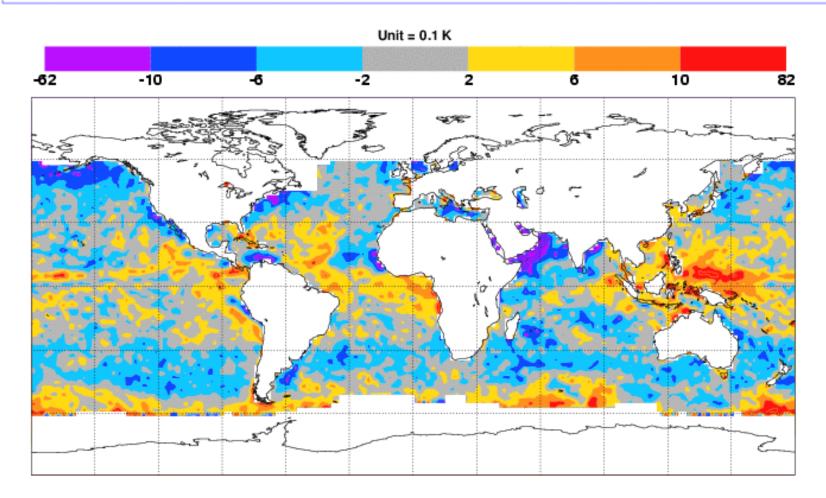
SSMI ch 3 (~q850) OBS\_VAL, 20090401-20090815, Expver=43, Analysis=DCDA, Time=[00,12]
All-sky microwave brightness temperature, +ve correlation with humidity, weighting: surface to 500 hPa





## SSM/I observation-model

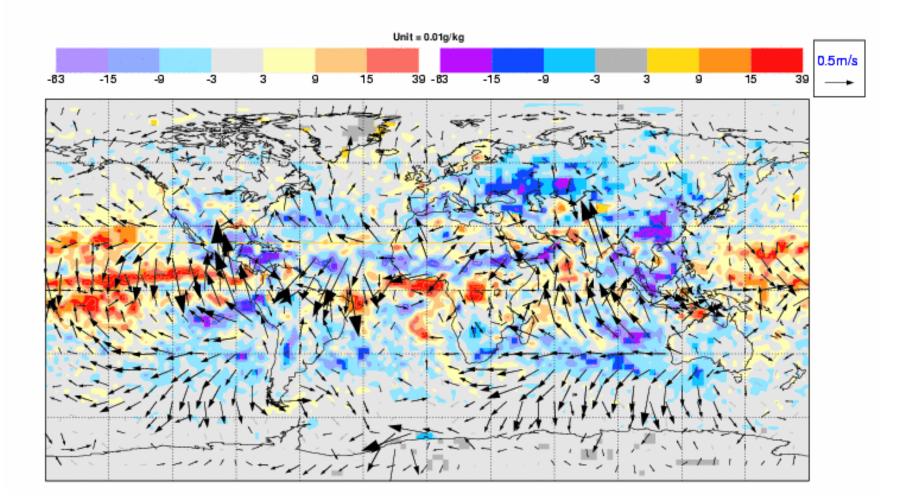
SSMI ch 3 (~q850) FG\_DEP, 20090401-20090815, Expver=43, Analysis=DCDA, Time=[00,12] All-sky microwave brightness temperature, +ve correlation with humidity, weighting: surface to 500 hPa





## Mean q-increment

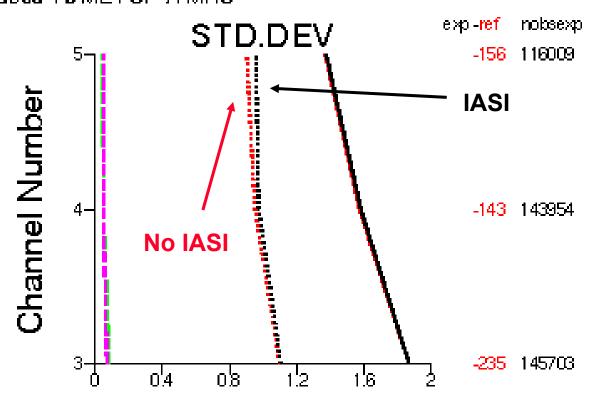
Analysis increments: Mean q850, 20090402-20090815, Expver=35R3 43, Analysis=dcda. Deep colours = 5% significance





## The curious case of degrading MHS channel 5 by adding IASI

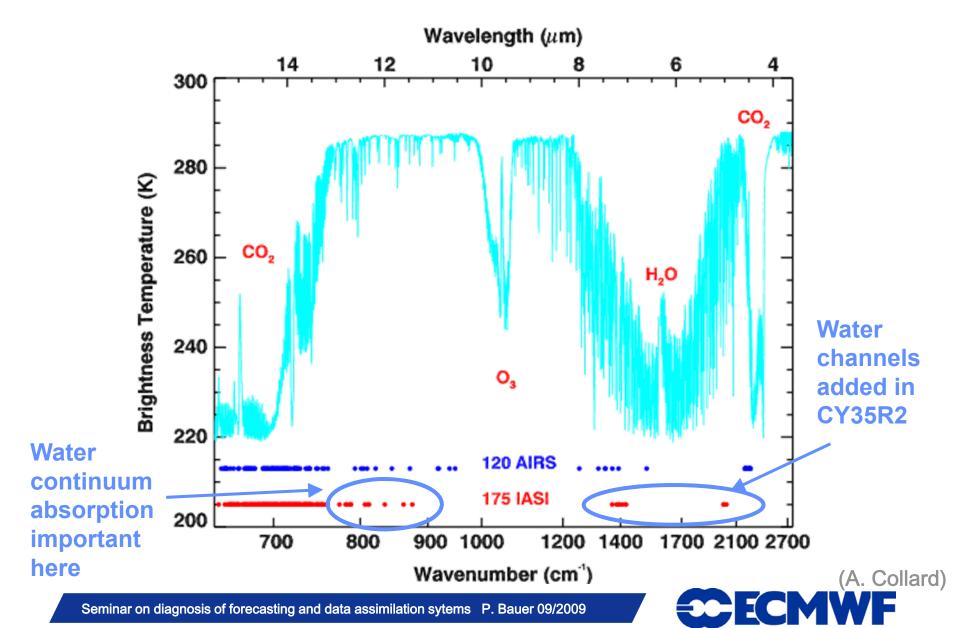
exp:ezep /DA (black) v. ezeq/DA 2007080100-2007081612(12) EUMETSAT TOVS-1C metop-a MHS Tb Tropics used Tb METOP-A MHS



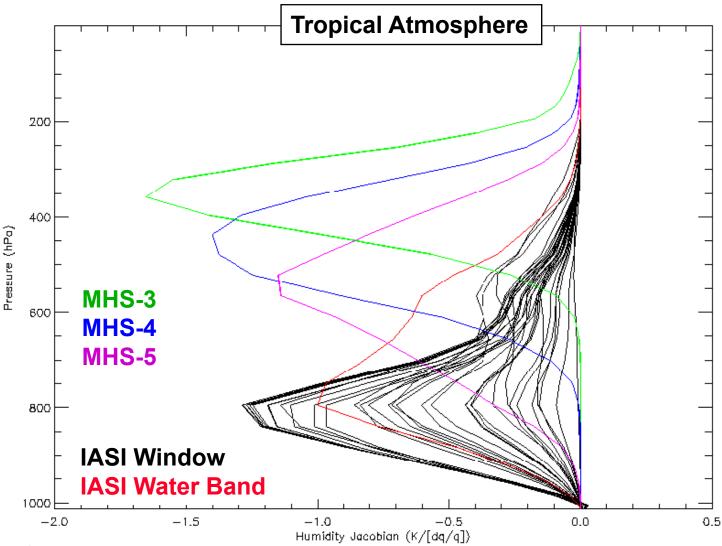
(A. Collard)



## ... without assimilating IASI water vapour channels



## IASI window (water vapour) Jacobians

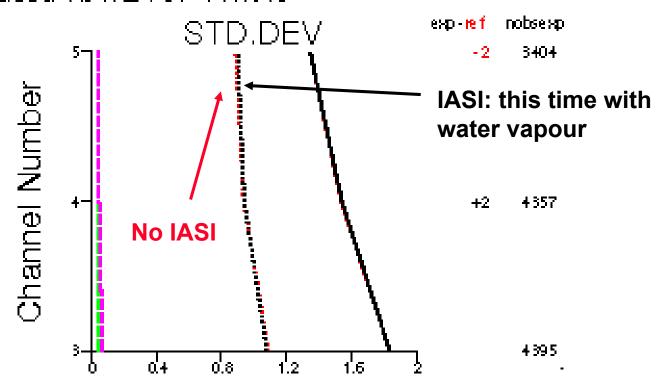


(A. Collard)



## ... and with assimilating IASI water vapour channels

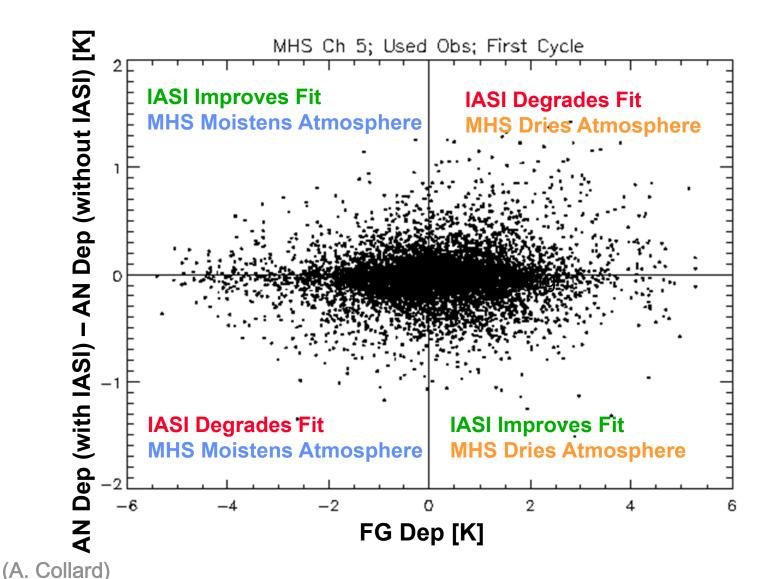
exp:f010 /DA (black) v. ezeq/DA 2007080100 EUMETSAT TOVS-1C metop-a MHS Tb Tropics used Tb METOP-A MHS



(A. Collard)

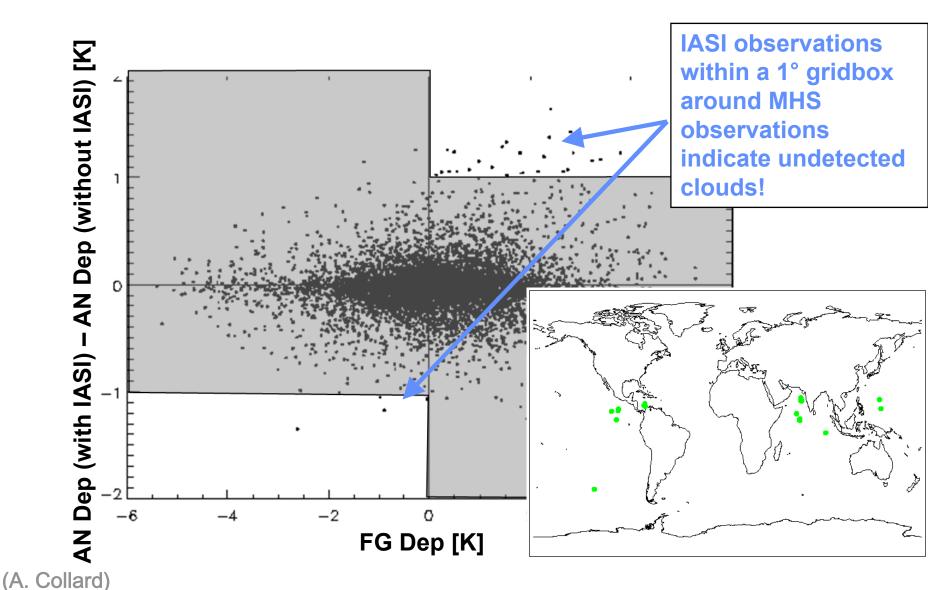


## Why is MHS-5 Analysis Degraded?





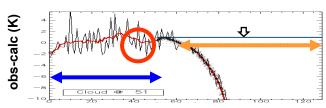
## Why is MHS-5 Analysis Degraded?





## AIRS/IASI cloud detection

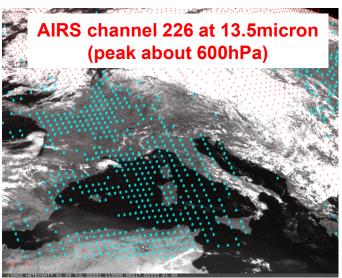
A non-linear pattern recognition algorithm is applied to departures of the observed radiance spectra from a computed clear-sky background spectra.

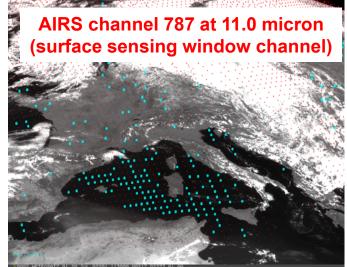


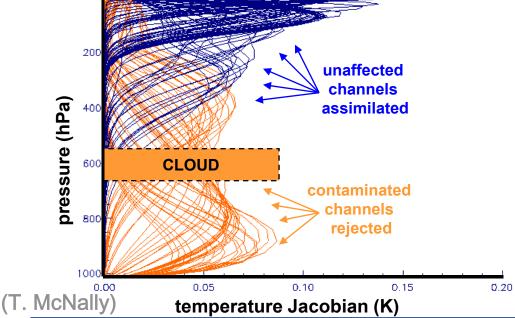
Vertically ranked channel index

This identifies the characteristic signal of cloud in the data and allows contaminated channels to be rejected



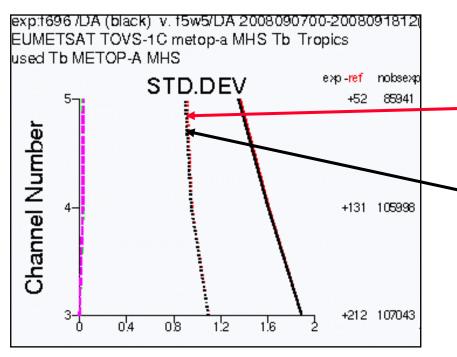








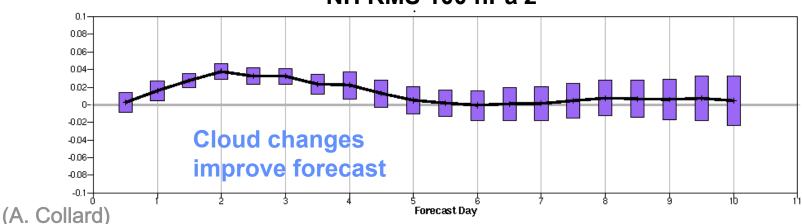
#### MHS observation statistics/scores



IASI without water vapour and unsafe cloud detection

IASI: this time with water vapour & cloud fix

#### NH RMS 100 hPa z





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## Impact of existing satellite data

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Impact of future satellite data

Summary



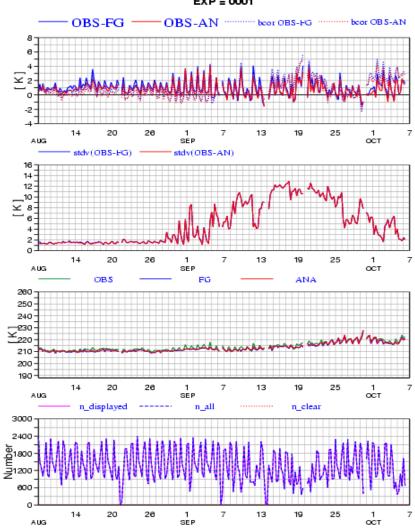
## Data monitoring - time series

#### Statistics for Radiances from Aqua / AIRS

Channel = 2104, All Data

Area: lon\_w= 0.0, lon\_e= 360.0, lat\_n= -70.0, lat\_s= -90.0 (over sea)

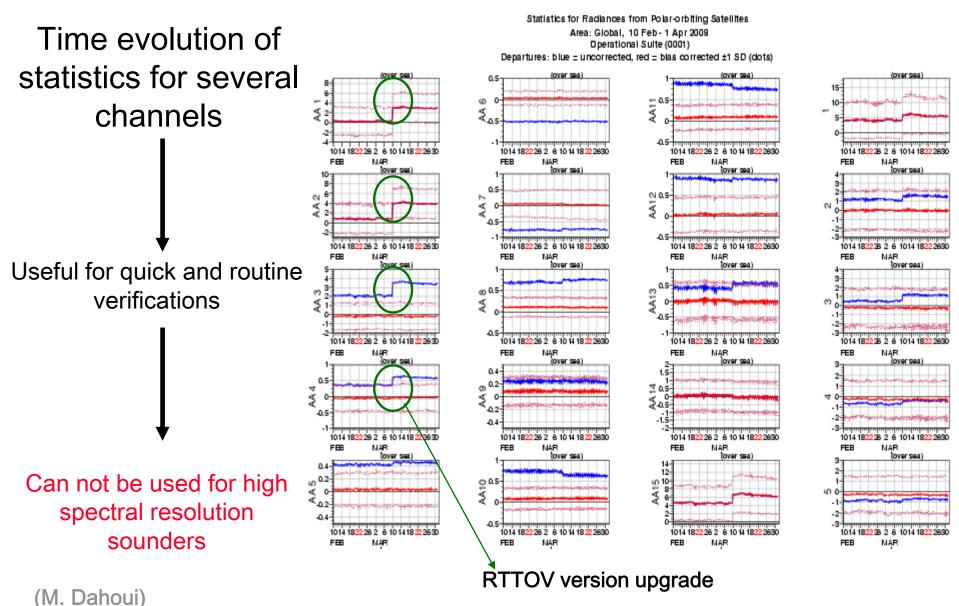
Time evolution of statistics over predefined areas/surfaces/flags



(M. Dahoui)



## Data monitoring - overview plots



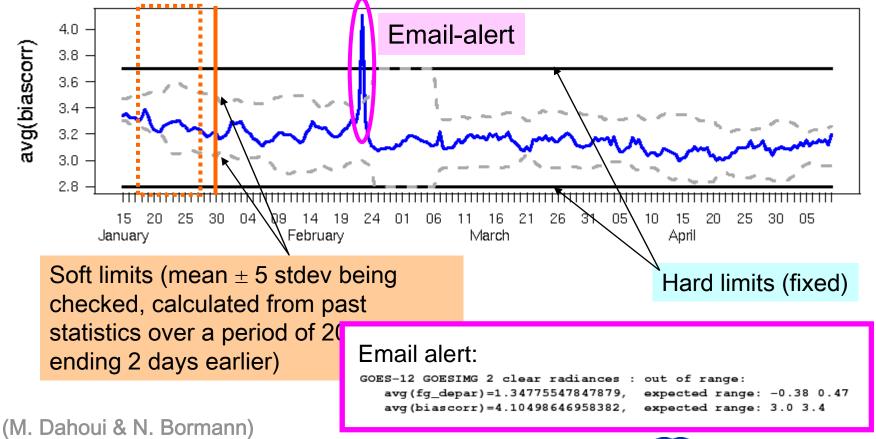


## Data monitoring - automated warnings

http://www.ecmwf.int/products/forecasts/satellite\_check/

Selected statistics are checked against an expected range.

E.g., global mean bias correction for GOES-12 (in blue):





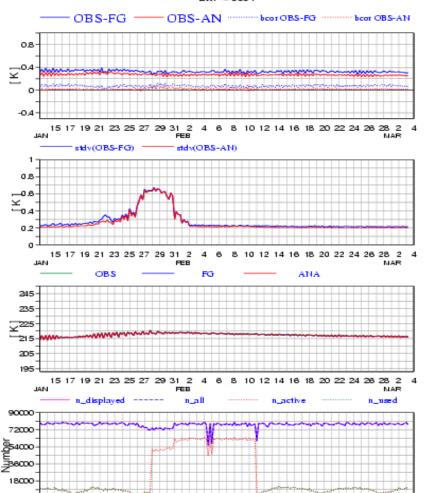
## Data monitoring - automated warnings

#### Statistics for Radiances from NOAA-16 / AMSU-A

Channel = 10, Se lected data: c lear

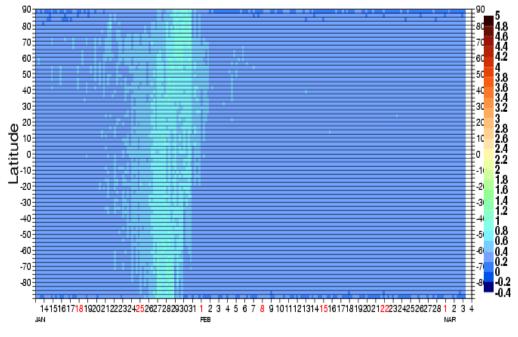
Area: lon\_w= 0.0, lon\_e=360.0, lat\_n= 90.0, lat\_s=-90.0 (all surface types)

EXP = 0001



STATISTICS FOR RADIANCES FROM NOAA-16 / AMSU-A
STDV OF FIRST GUESS DEPARTURES (OBS-FG) [K] (CLEAR)
CHANNEL = 10

EXP = 0001, DATA PERIOD = 2009011300 - 2009030400 Min: 0.136874 Max: 1.0036 Mean: 0.362573





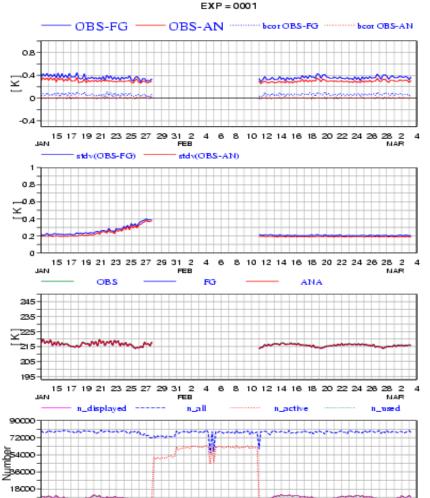
## Data monitoring - automated warnings

#### Statistics for Radiances from NOAA-16 / AMSU-A

Channel = 10, Used Data

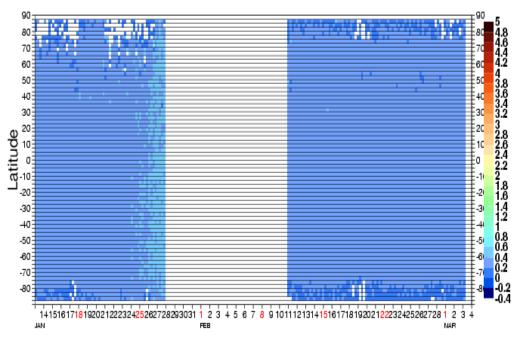
Area: lon\_w= 0.0, lon\_e= 360.0, lat\_n= 90.0, lat\_s=-90.0 (all surface types)

EXP = 0001



STATISTICS FOR RADIANCES FROM NOAA-16 / AMSU-A
STDV OF FIRST GUESS DEPARTURES (OBS-FG) [ K ] (USED)
CHANNEL = 10

EXP = 0001, DATA PERIOD = 2009011300 - 2009030400 Min: 0 Max: 0.657172 Mean: 0.286458





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## Observing System Experiments

#### 'Continuous' observation impact assessment:

- Assessment of all individual and combined components of observing system
- → OSEs denying types from operational model/observing system, adding types to baseline system.

#### **Investigating fundamental observation impact:**

- Comparison between instruments that constrain similar variables.
- Evaluation of specific operator sensitivity & 4D-Var mechanisms.
- → OSEs with single observation type in addition to poor observing system (e.g. conventional + AMVs + 1 sounder) and operational model version

#### Adding (improving) a new observation type:

- Introduction of new observation types.
- Improvement of assimilation of existing observations.
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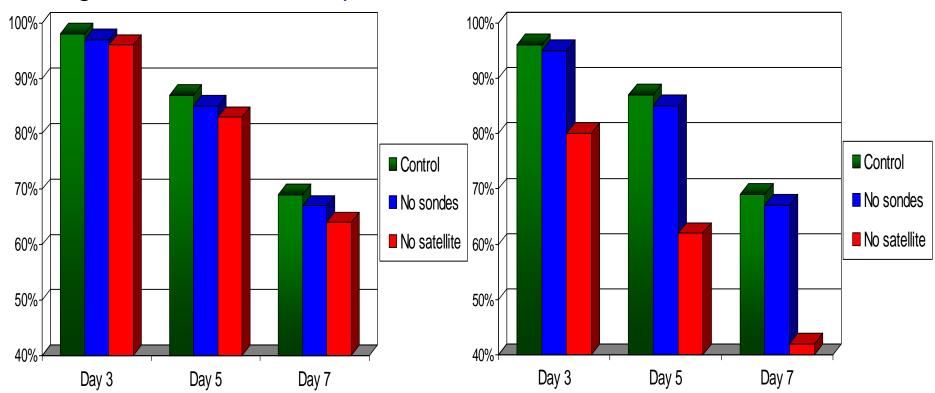


## Impact of conventional vs satellite data

## **Observing System Experiments**

Anomaly correlation of 500hPa height for Northern hemisphere

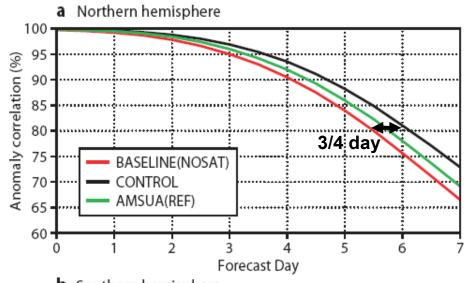
Anomaly correlation of 500hPa height for Southern hemisphere

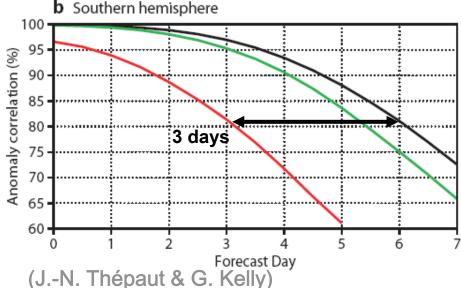


(J.-N. Thépaut & G. Kelly)



## Combined impact of all satellite data





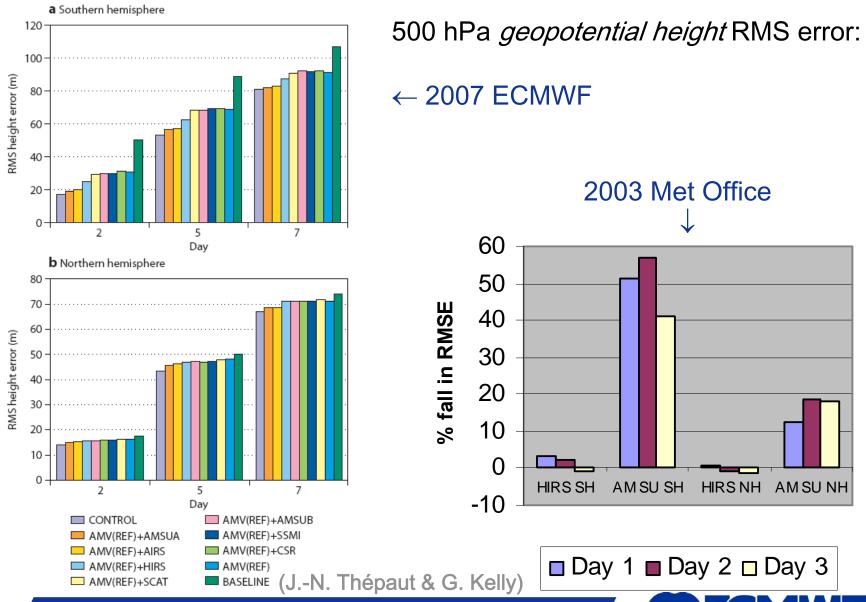
## EUCOS Observing System Experiments (OSEs):

- 2007 ECMWF forecasting system,
- winter & summer season,
- different baseline systems:
  - no satellite data (NOSAT),
  - NOSAT + AMVs,
  - NOSAT + 1 AMSU-A,
- general impact of satellites,
- impact of individual systems,
- all conventional observations.

← 500 hPa *geopotential height* anomaly correlation

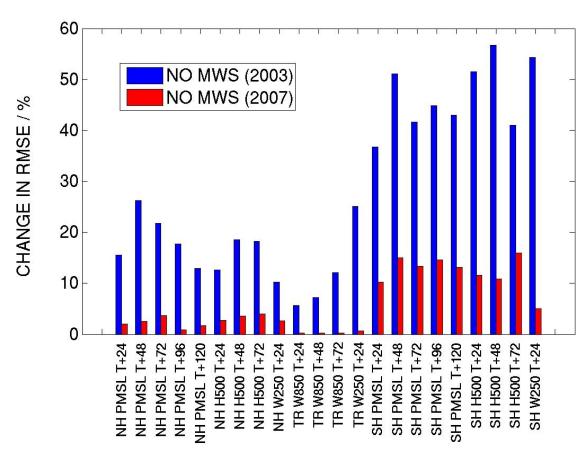


## Individual impact of satellite data





# Impact of microwave sounder data in NWP: Met Office OSEs



#### 2003 OSEs:

- N-15,-16 and -17 AMSU
- N-16 & N-17 HIRS
- AMVs
- Scatterometer winds
- SSM/I ocean surface wind speed
- Conventional observations

#### 2007 OSEs:

- N-16, N-18, MetOp-2 AMSU
- SSMIS
- AIRS & IASI
- Scatterometer winds
- AMVs
- SSM/I ocean surface wind speed
- Conventional observations

(W. Bell)



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Investigating fundamental observation impact:

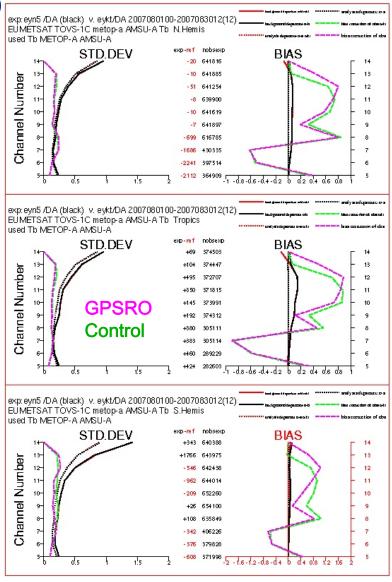
**GPSRO** 

Example: How do GPSRO data (unbiased – not bias corrected) affect variational bias correction of AMSU-A radiance data (at levels where model temperature biases are significant)?

- OSE: only conventional + Metop AMSU-A, MHS, initialized with operational analysis:
  - control
  - control + COSMIC GPSRO
- Variational bias correction active.
- → AMSU-A channel 8-11 bias correction smaller when GPSRO data present (better constraint)
- → AMSU-A channel 12-13 bias correction larger when GPSRO data present (model bias too large?)

(AMSU-A channel 14 bias frozen)

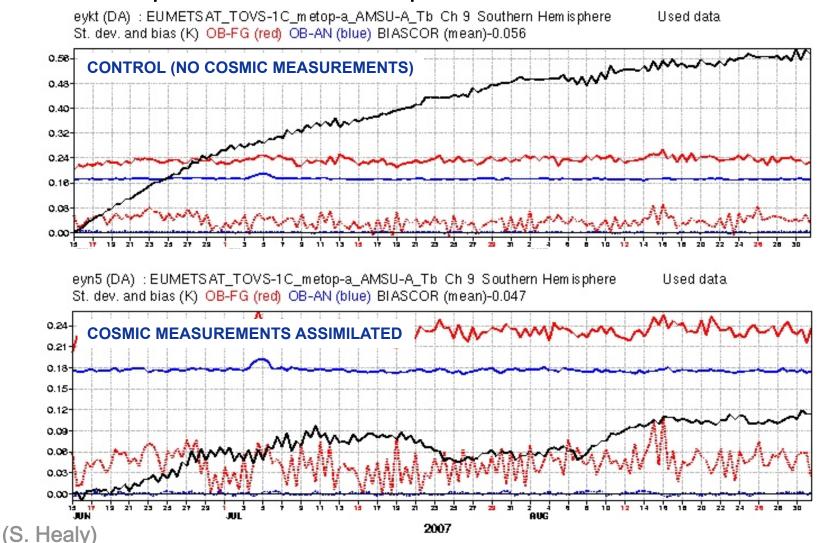
(S. Healy)





# Investigating fundamental observation impact: GPSRO

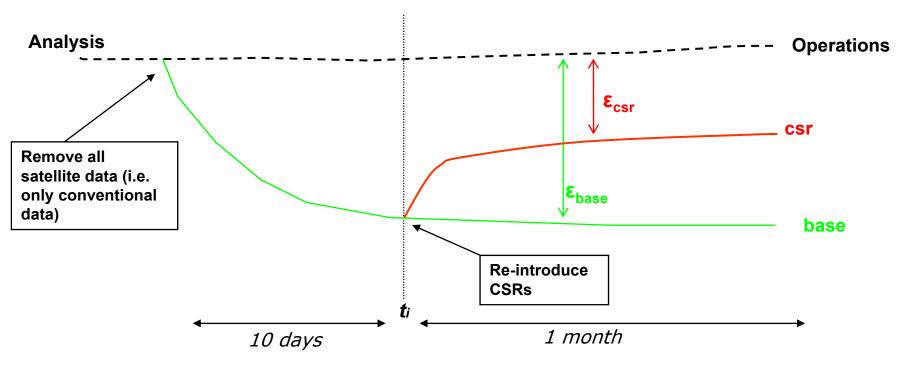
#### Metop AMSU-A channel 9 departure & bias correction evolution



# Investigating fundamental observation impact: CSRs from geostationary satellite radiometers

Objective: What is impact of CSRs on wind analysis?

- CSRs: Meteosat-9 only, 2 water vapour channels (300 and 500 hPa)
- AMVs: Meteosat-9 only, all AMVs assimilated in operations (infrared, water vapour and visible)



(c<sub>base</sub>

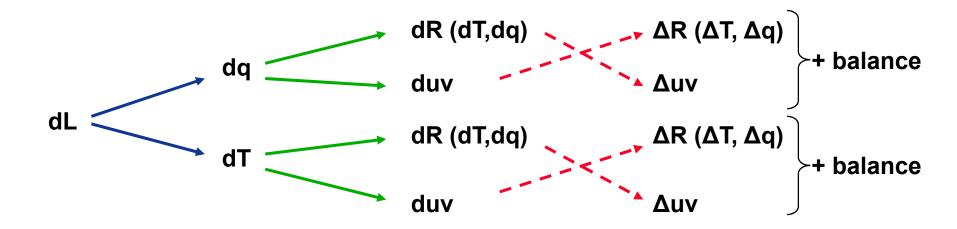
 $(\varepsilon_{\text{base}} - \varepsilon_{\text{csr}}) > 0 \Rightarrow \text{positive impact}$ 

(C. Peubey & T. McNally)



# Investigating fundamental observation impact: CSRs from geostationary satellite radiometers

$$\nabla_{\delta x_o} J = \mathbf{B}^{-1} \delta \mathbf{x}_o + \sum_{i=0}^{n} \mathbf{M}^{T} (t_i, t_o) \mathbf{H}_i^{T} \mathbf{R}_i^{-1} (\mathbf{d}_i - \mathbf{H}_i \delta \mathbf{x}_i)$$



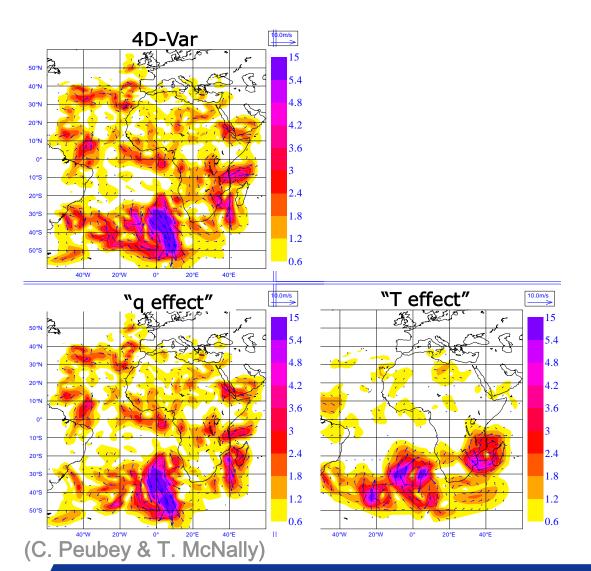
**Experiments:** 

- Baseline + "4D-Var" CSRs
- Baseline + "T tracer effect" CSRs (i.e. no q sensitivity)
- Baseline + "q tracer effect" CSRs (i.e. no T sensitivity)

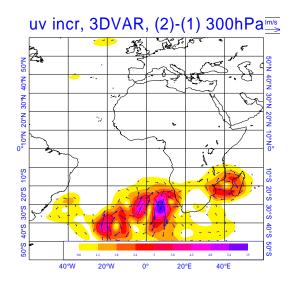
(C. Peubey & T. McNally)



## 300 hPa wind increments 1st analysis cycle



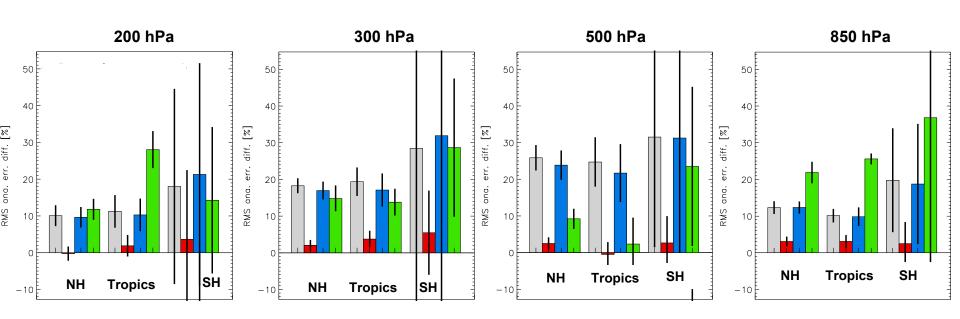
## No q-tracing due to missing model integration





## Analysis error reduction

WIND SPEED: Base + ■4D-Var CSRs + ■"no 4D-Var" CSRs + ■"q eff." CSRs + ■no 4D-Var CSRs+AMVs



- → With cycling alone (no 4D-Var), CSRs do not have a significant impact on winds.
- → Most of the CSR impact on wind seem to come from the 4D-Var tracer effect of humidity.
- → Not shown: humidity and temperature also better with 4D-Var wind increment, so there is a feedback between wind and temperature/moisture:

better wind 

→ better q and T

(C. Peubey & T. McNally)



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## Adding a new observation type

#### **Technical implementation:**

- BUFR conversion of received format (if necessary)
- BUFR conversion to observational database (ODB) that is used in analysis system.
- Management of satellite/instrument IDs in system
- Generation of radiative transfer model coefficients
- Screening (q/c for data problems, clouds, surfaces)
- Management of satellite/instrument in variational bias correction

#### **Monitoring:**

- Blacklisting of observations (i.e. data active in screening but not in minimization)
- Monitoring experiments to evaluate data quality and spin up biases

**Diagnostics** 

#### **Analysis impact evaluation:**

- Assimilation experiments with data active and evolved biases (plus control)
- Impact on short-range forecast/analysis fit to other observations
- Impact on mean analysis state

#### Forecast impact evaluation:

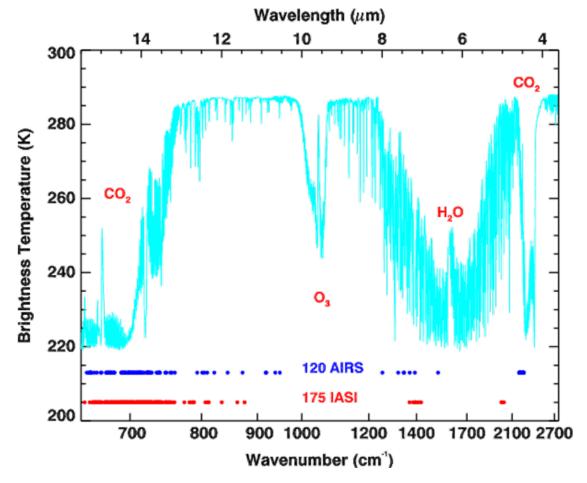
- Assimilation experiments with data active and evolved biases (plus control)
- Impact on short-to-medium-range forecasts (statistical significance)

#### **Evaluation for operational implementation with a new cycle:**

Repeat previous two steps with other modifications



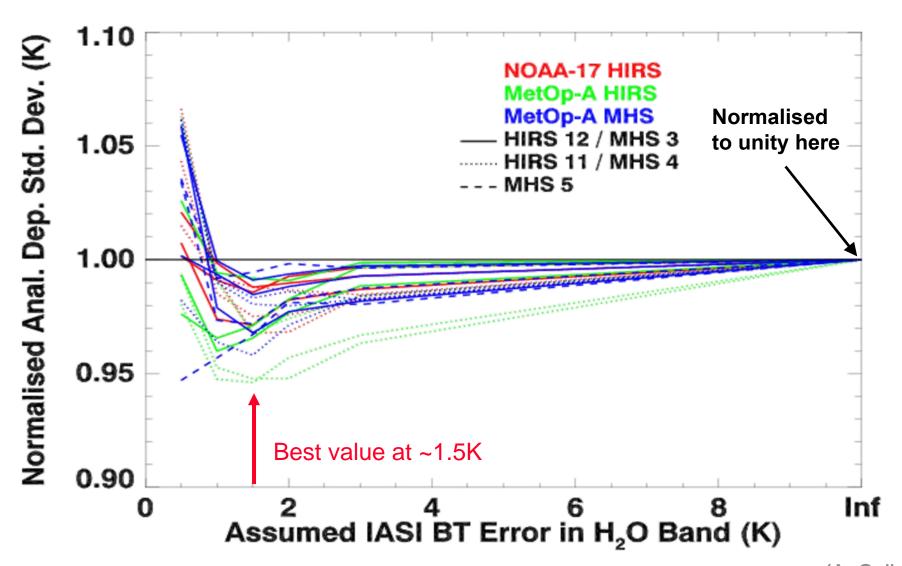
## Example - Advanced IR sounders



- AIRS CO<sub>2</sub> and H<sub>2</sub>O channels assimilated since October 2003.
- IASI CO<sub>2</sub>/H<sub>2</sub>O channels assimilated since June 2007/March 2009.
- Assimilated in clear-sky areas and above clouds; since March 2009 in fully overcast situations, AIRS (not IASI) over land surfaces/sea-ice.
- Continuous revision of channel usage, quality control.



## Adding 10 IASI water vapour channels



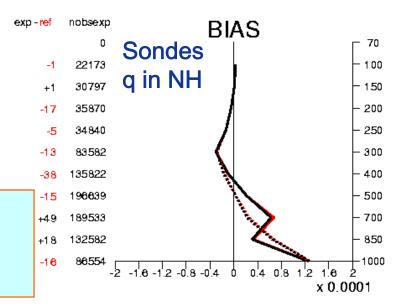


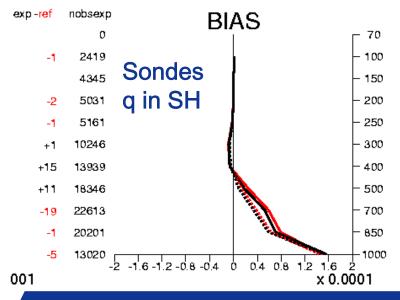
#### Fit to other observations: Radiosondes

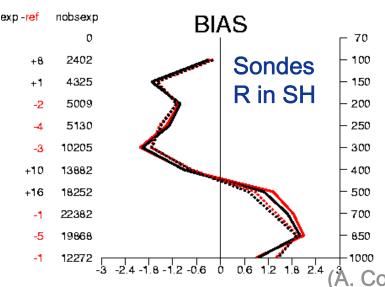
Black: IASI w/ humidity channels

**Red: Control** 

The addition of the IASI water band improves the analysis fit to radiosondes

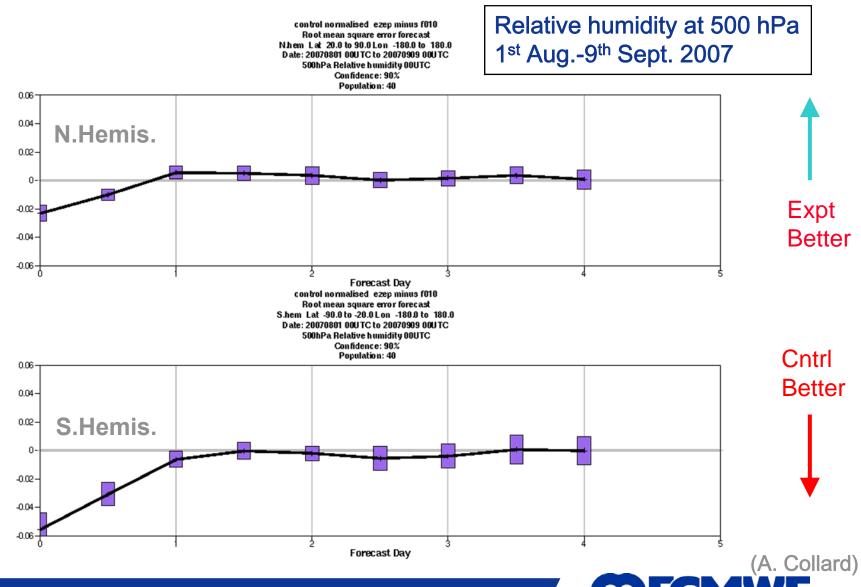




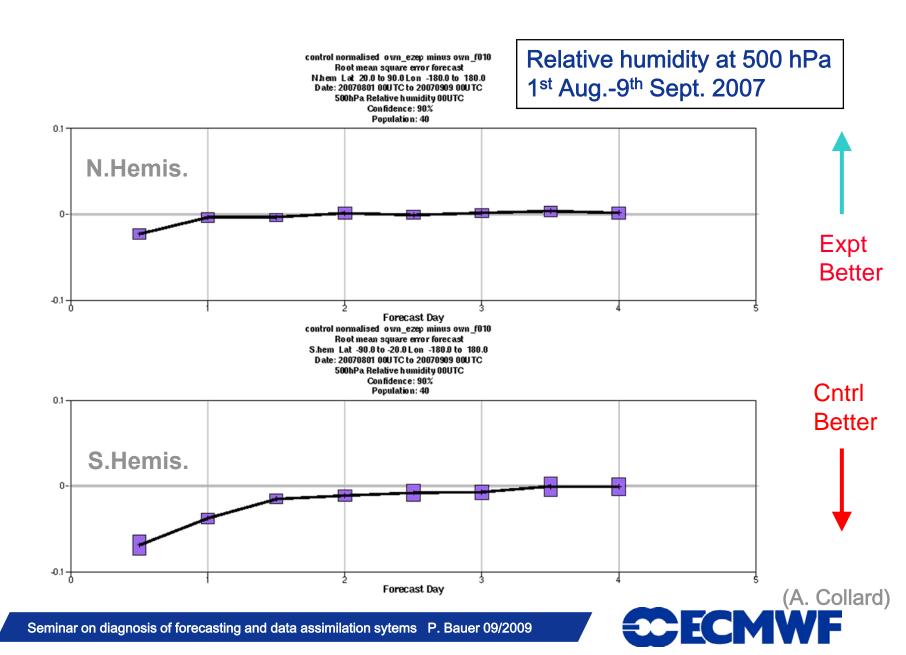




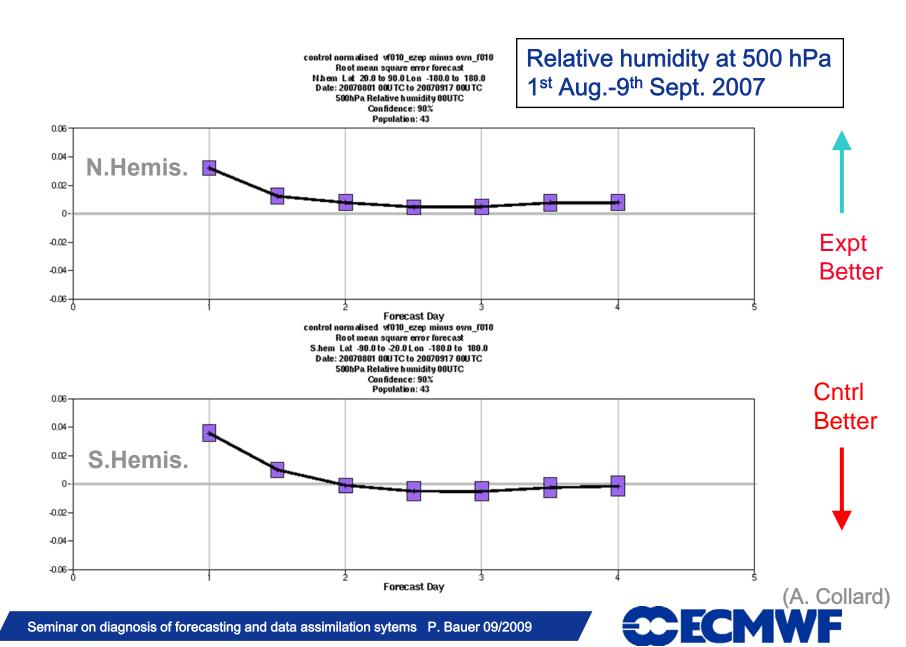
## Forecast skill: Verified with operational analysis



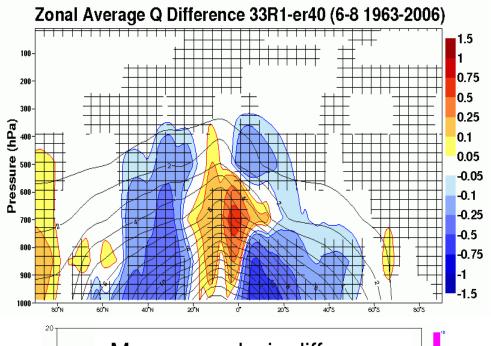
## Forecast skill: Verified with own analysis

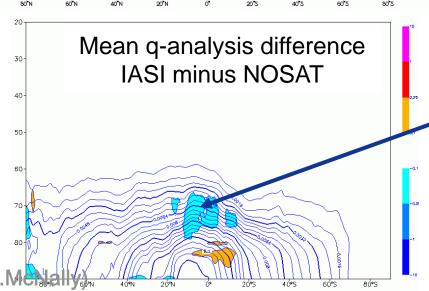


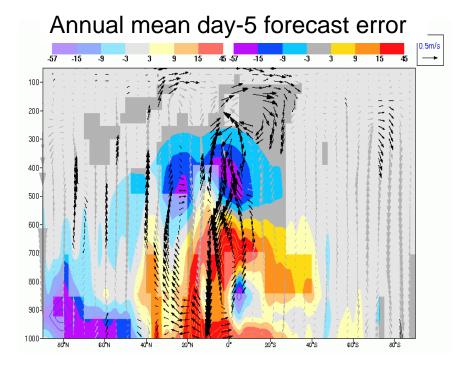
## Forecast skill: Verified with experiment's analysis



#### Forecast skill and model bias







The IASI observations act to dry the NOSAT (and OPS) system which has run to an excessively moist state

- → is consistent with the observed climate bias of the forecast model 700hPa.
- → observations that draw analysis away from model climate will score negatively unless when both experiment and control are verified with improved analysis



## Satellite data usage at ECMWF

## Impact of existing satellite data

- Data monitoring
- Observing system experiments
- Advanced diagnostic tools

Impact of future satellite data

Summary



## Adjoint diagnostics - OSEs

Remarks on ADJ-OSE by Ron Gelaro (NASA GMAO):

**ADJ**: measures the impacts of observations in the context of all other observations present in the assimilation system

**OSE**: removal of observations changes or degrades the system... differs for each member

**ADJ**: measures the impact of observations in each analysis cycle separately and against the control background

**OSE**: measures the impact of removing information from both the background and analysis in a cumulative manner

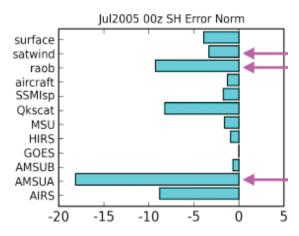
**ADJ**: measures the response of a single forecast metric to all perturbations of the observing system

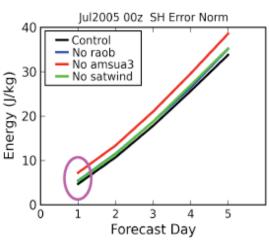
**OSE**: measures the effect of a single perturbation on all forecast metrics

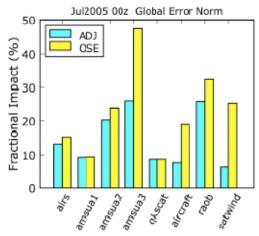


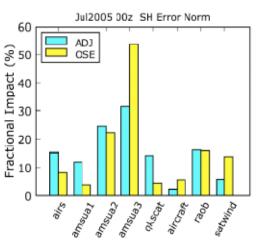
## Adjoint diagnostics - OSEs

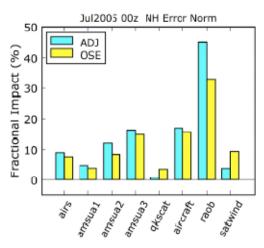
## % Contributions to 24hr Forecast Error Reduction July 2005

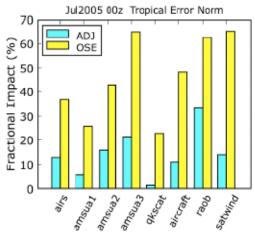












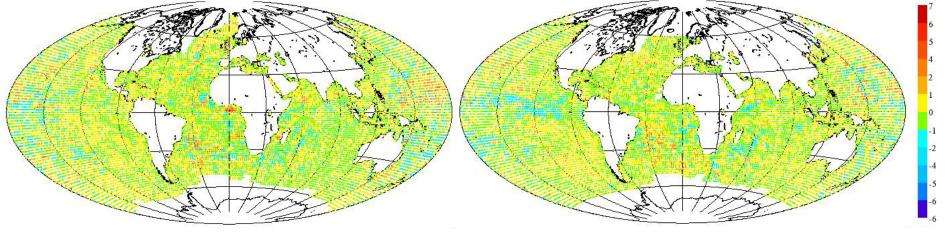
(R. Gelaro)

(Y. Zhu & R. Gelaro 2009)

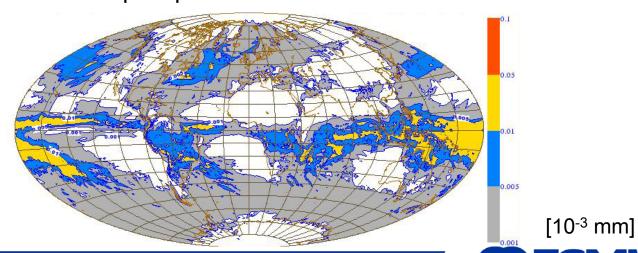
# Investigating fundamental observation impact: TCWV

Forecast sensitivity to observations in analysis





Mean 36-12h precipitation forecast initialized at 12 UTC



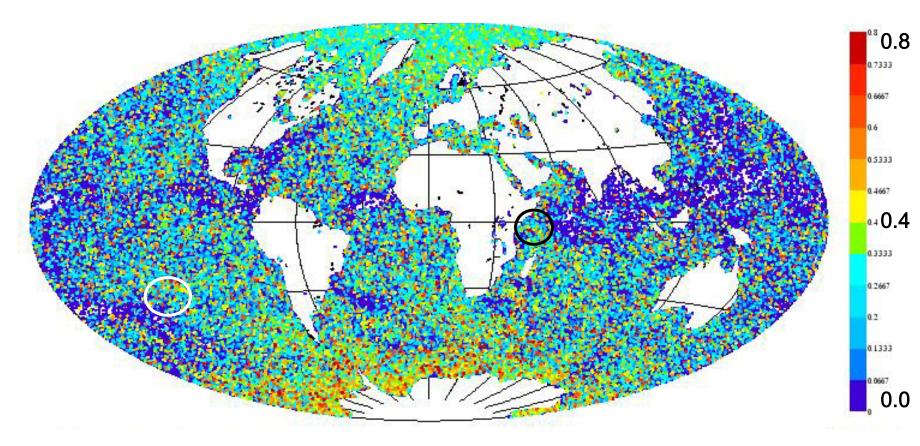
[J/kg]

Seminar on diagnosis of forecasting and data assimilation sytems P. Bauer 09/2009

(C. Cardinali)

#### IASI - channel 212 (250 hPa)

First-guess departure standard deviation (K; 7 days)
Assimilation over sea-ice but not over land

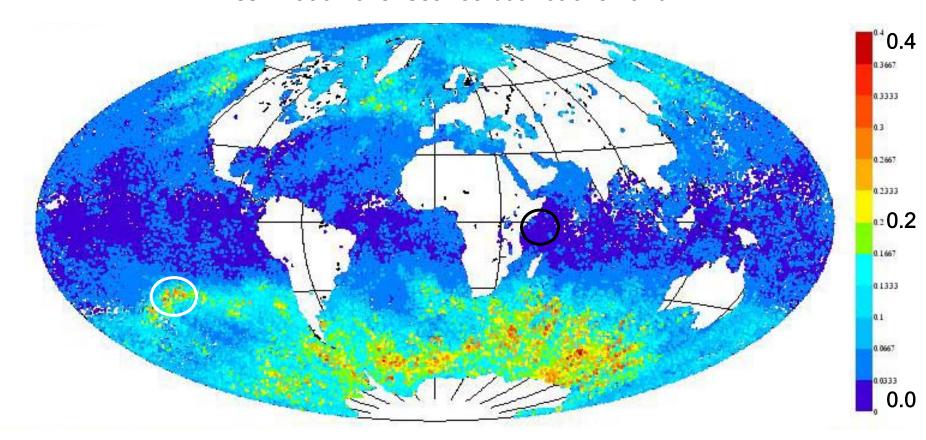


⇒ Information available for analysis from observations (= innovation)



#### IASI - channel 212 (250 hPa)

Mean analysis sensitivity to observations\* (7 days)
Assimilation over sea-ice but not over land



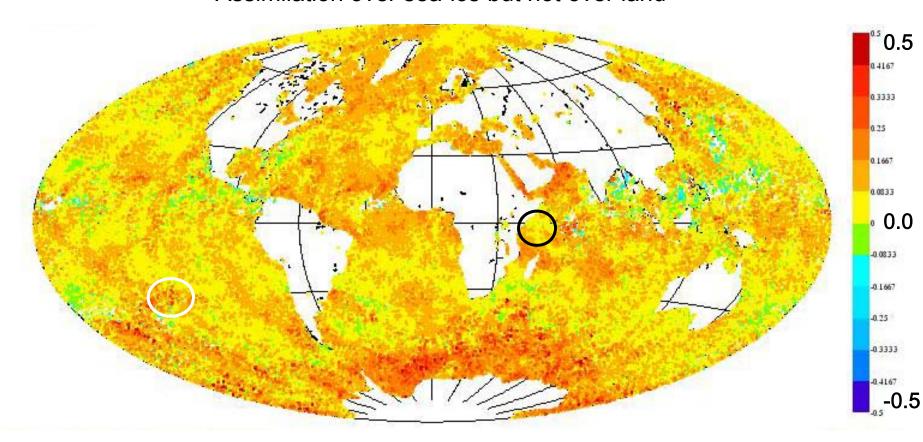
⇒ Sensitivity of the analysis to those observations

(\* or self-sensitivity, see Cardinali et al. (2004))



### IASI - channel 212 (250 hPa)

Mean analysis increment (K; 7 days)
Assimilation over sea-ice but not over land



⇒ Work performed by the analysis in observation space



## Satellite data usage at ECMWF

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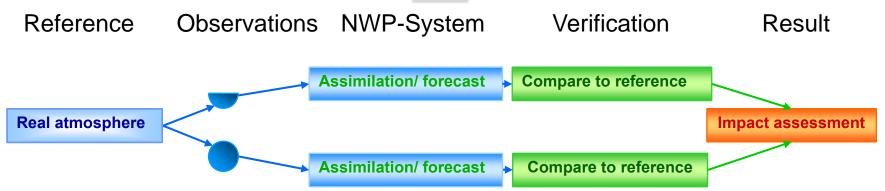
## Impact of future satellite data

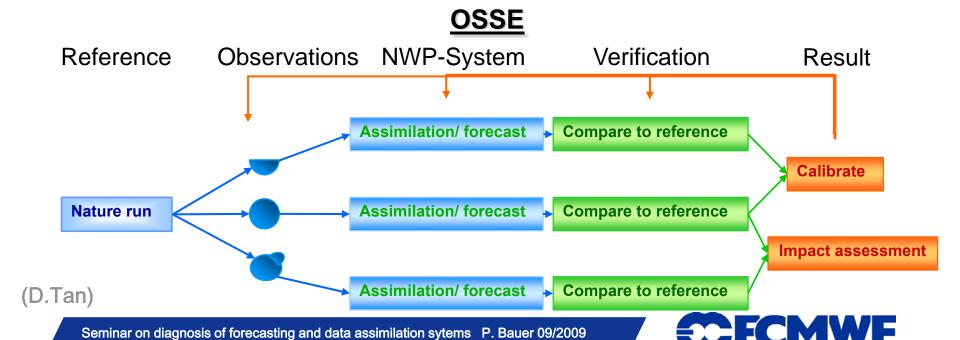
Summary



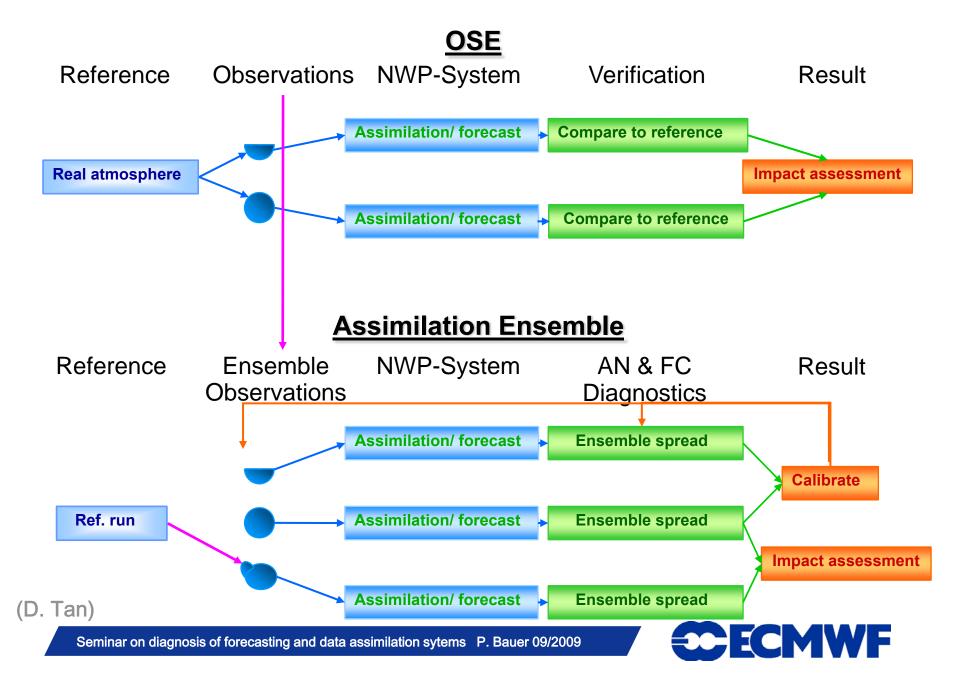
#### OSEs and OSSEs

#### <u>OSE</u>

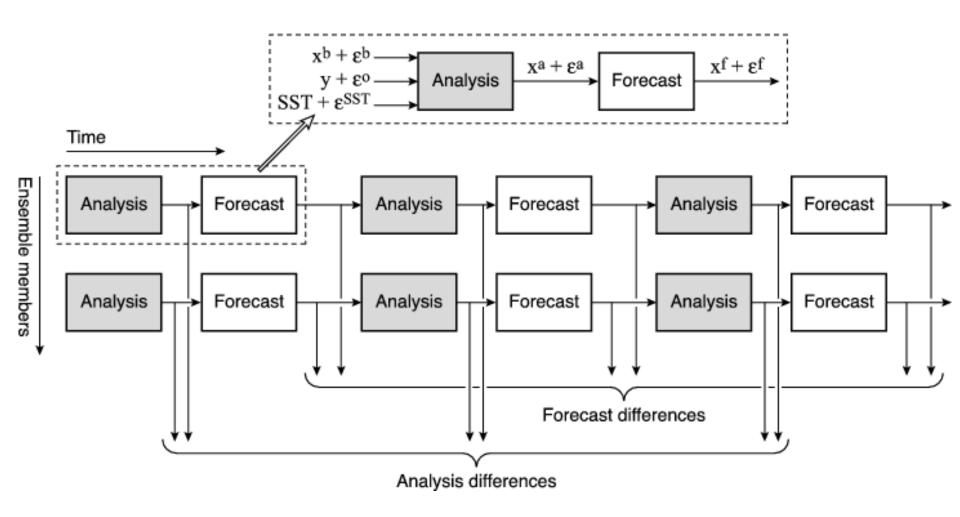




#### **OSEs and Ensembles**



#### Data assimilation ensembles

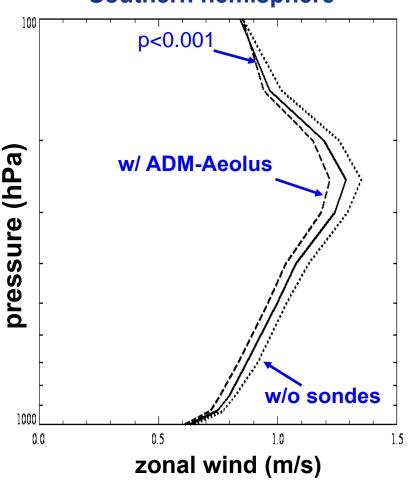


(D. Tan)



#### Data assimilation ensembles

## 12-hour forecast impact Southern hemisphere



Ensemble spread in zonal wind (u, m/s) ≈ forecast error

Scaling factor ~2 for wind error:

- Performance similar in Tropics, Northern and Southern hemisphere
- Simulated DWL adds value at all altitudes and in longer-range forecasts (t+48h, t+120h)
- Differences significant (T-test)
- Supported by information content diagnostics

(D. Tan)



## Satellite data usage at ECMWF

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Impact of future satellite data

Summary



## Concluding remarks

- OSEs are continuously performed for:
  - assessment of new (revised) observation impact along model updates;
  - study of basic impact features (poor observing system);
  - assessment of ensemble of observing system components.
- Impact is currently evaluated using:
  - fit to short-range forecast/analysis model fields (consistency, reference observations);
  - model forecast skill using standard scores.
- Shortcomings of current observation impact assessment:
  - evaluation of individual observation type impact on fit of model fields to other observation types is only available for analyses *not* forecasts;
  - diagnostics for tuning/optimization of observing system is not available (thinning, channel selection, observation errors);
  - overview diagnostics require large and costly set of OSEs, no continuous built-in evaluation yet;
  - standard forecast scores can contradict analysis evaluation (new observations add noise and may increase root-mean-square 'error').

