

# Specific issues related to the use of satellite data in limited area models

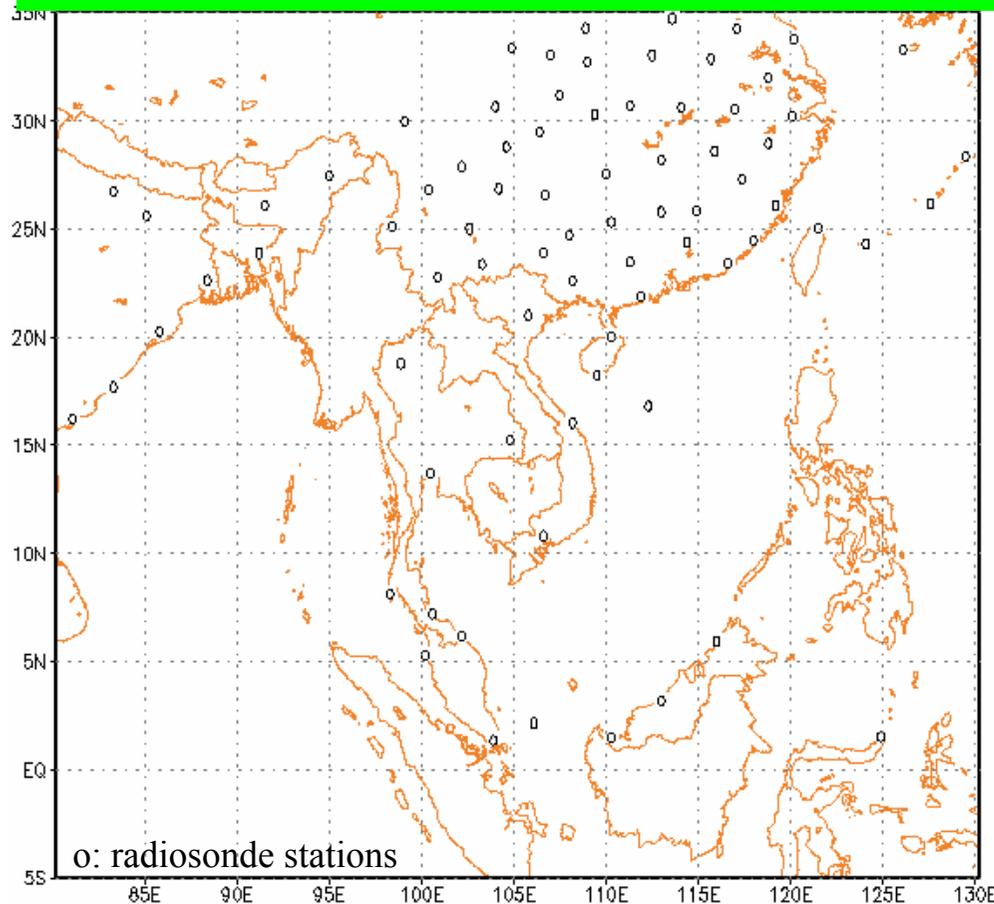
Reinhold Hess, Francesca Di Giuseppe,  
Christoph Schraff, Blazej Krzeminski

ECMWF Seminar

3-7 September 2007, Reading

## Lokal Model HRM (in Vietnam)

### Domain



**HRM:** Hydrostatic Regional Model,  
resolution 28 km (14 km)  
boundary and **initial** values from GME  
provides operational forecasts in **Vietnam**

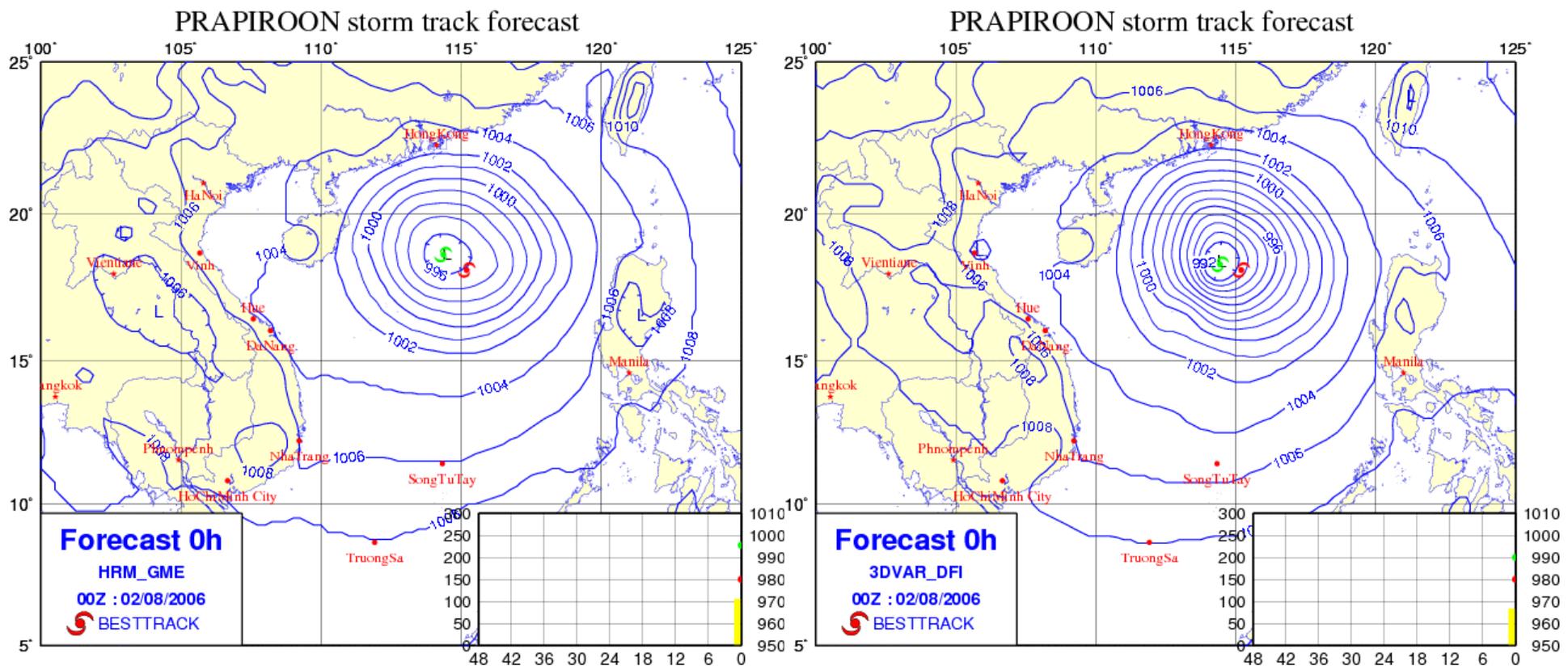
**Project:** Improvement of quantitative  
precipitation forecasts over Vietnam

- application of **3d-Var**
- use of satellite radiances (**ATOVS**)
- adaptation and improvement of physics  
(convection)
- verification
- ...

Do we need own analysis for HRM?

- all data (almost) already in GME

## Forecast quality of regional models depend on ...



initial state interpolated from GME

initial state from 3D-VAR (no satellites)

Source: Le Duc, Vietnam National University of Hanoi

## Forecast quality of regional models depend on ...

- initial state
  - ◆ fit to observations (truth)
  - ◆ consistency with numerical model, small scale features in initial state (resolution, orography, vertical distribution of humidity, etc.)
- numerical model, resolution, approximations (e.g. hydrostasy), physics (e.g. convection), parameterisations
- quality of boundary values
- timeliness of forecast, (short range forecasting)

## Limited area models ...

- high resolution (2-14 km)
- small scale structures in space and time (e.g. convection)
- delicate physics (e.g. steep orography, discontinuous solutions, bifurcations)
- limited predictability of small scale phenomena (computationally and physically)
- fewer constraints (e.g. hydrostasy, geostrophy)
- need to use asynchronous and high frequent observations (e.g. SEVIRI/MSG, radar)
- limited area (driven by boundary values of embedding models)
- over land (use of radiances over land)

## Specific issues of limited area models for the use of satellite data (radiances)

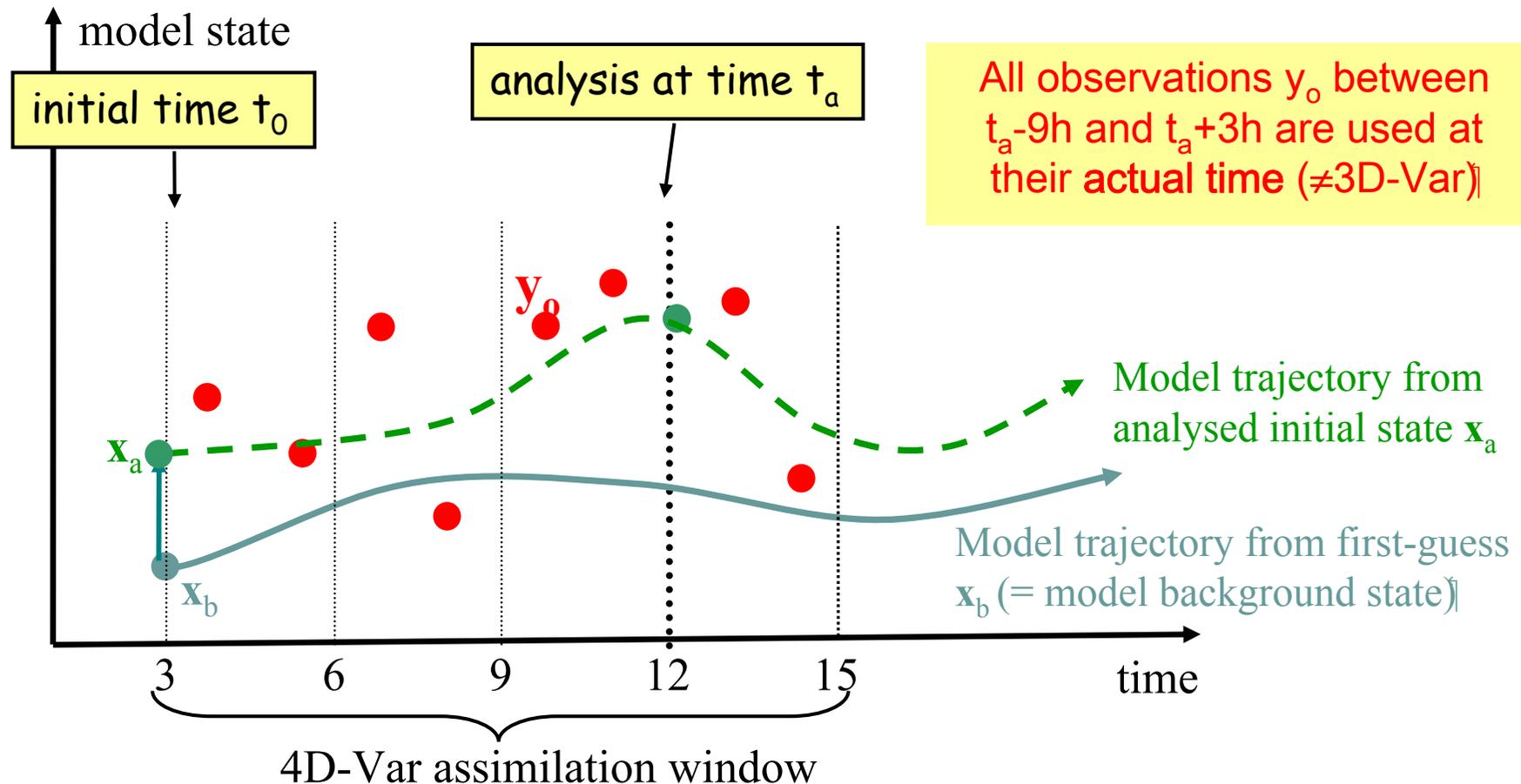
- **assimilation scheme**
  - ◆ provide initial value fit and consistent with the limited area model
  - ◆ use temporally and spatially highly resolved observations (background and observations errors and correlations)
  - ◆ complex and situation dependent statistics for background and observation errors, flow dependence, more critical vertical structure (temp/hum/wind)
- **bias correction**
  - sample size, representativity of samples, overfitting, choice of predictors
- **first guess above model top**
- **tuning** of observations (thinning)
  
- **use of data over land** (surface emissivity, higher resolution of surface conditions)
- **verification**
  - statistical representativity of results, influence of boundary values

**4D-Var** approach: initial state minimises misfits of model trajectory to observations and deviation from first guess.

**3D-Var**: as 4D-Var, but all observations valid for time of analysis, no computation of trajectory during minimisation

**3D-Var-FGAT**: use trajectory at observation time for first guess, but keep innovations constant

*All information has to be reflected in initial state (analysis)*

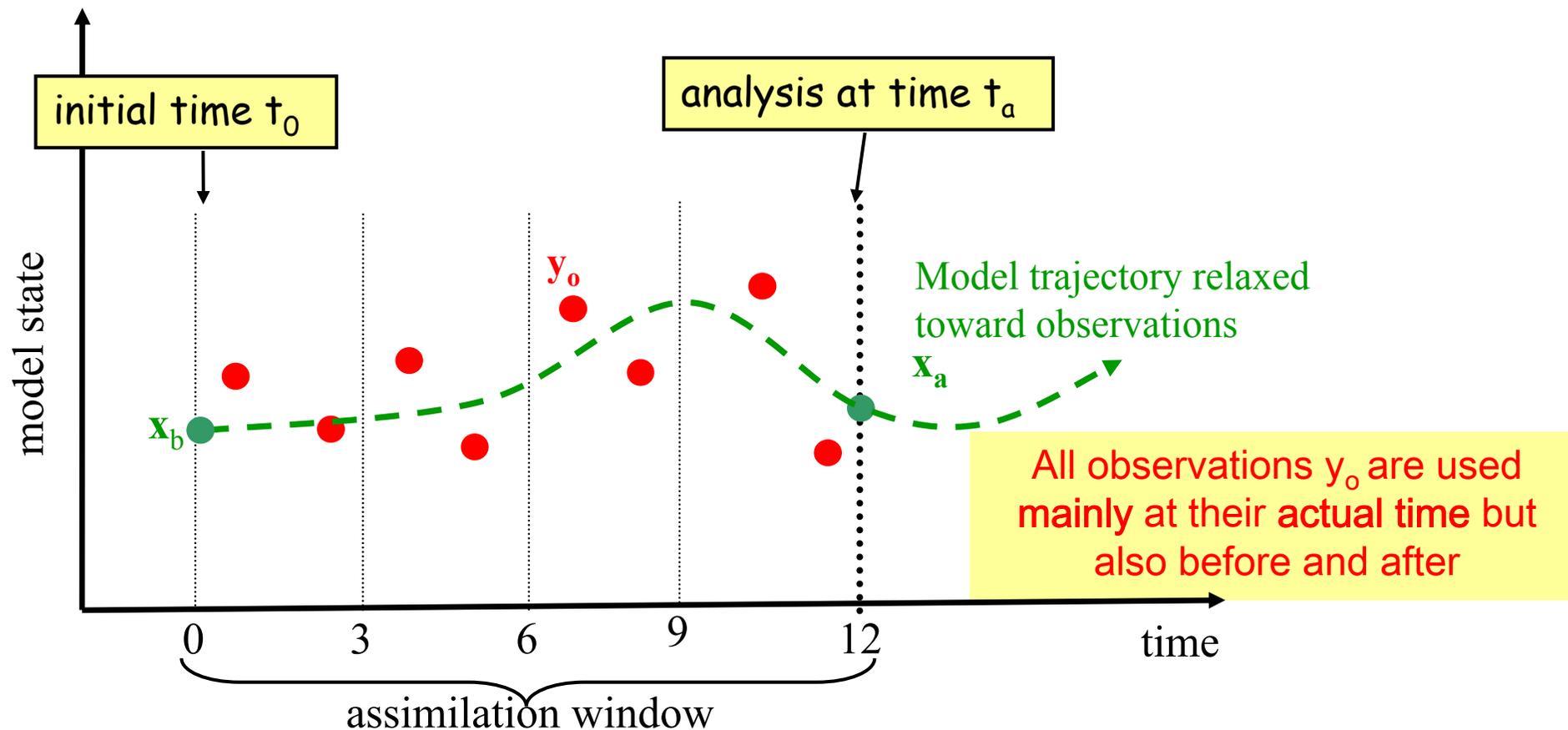


## Nudging approach (Newtonian Relaxation Scheme):

The model trajectory is nudged in every time step towards the observations with special terms additional to the model dynamics (nudging towards observations during forecast).

The sizes of the terms depend on the distance to the observations and on the time difference between observation and current model time.

*Difficult to use nonlinear observations*



# Pros and Cons for limited area models

## 4D-Var:

- + use of asynchronous observations
- + nonlinear observations can be used
- + consistent mathematical framework (obs and fg errors)
- solutions are less smooth and predictable (physically)
- physics are more complex (tangent linear and adjoint)
- specification of background errors is more difficult (boundary, fewer constraints)
- time consuming

## 3D-Var (EnKF):

- + consistent mathematical framework
- + combination with ensemble methods
- use of observations at time of analysis
- requires initialisation

## Nudging:

- + unsteady solutions, complicated physics
- + use of asynchronous **and** high frequent observations
- + fast, provide timely forecast
- + no initialisation required
- + combination with ensemble methods
- use of nonlinear observation operators
- no consistent mathematical framework, lots of tuning required

# COSMO-Project:

*Assimilation of satellite radiances with 1D-Var and Nudging*

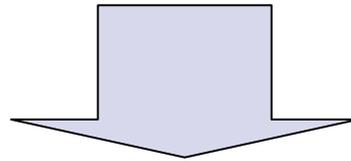
COSMO: Consortium for Small-scale Modeling

(Germany\*, Switzerland, Italy\*, Greece, Poland\*, Romania)

\* Member of Project

Goals of Project:

- Assimilate radiances (SEVIRI, ATOVS, AIRS/IASI) in COSMO-EU
- Explore the use of nonlinear observation operators with **Nudging**



1DVAR + Nudging

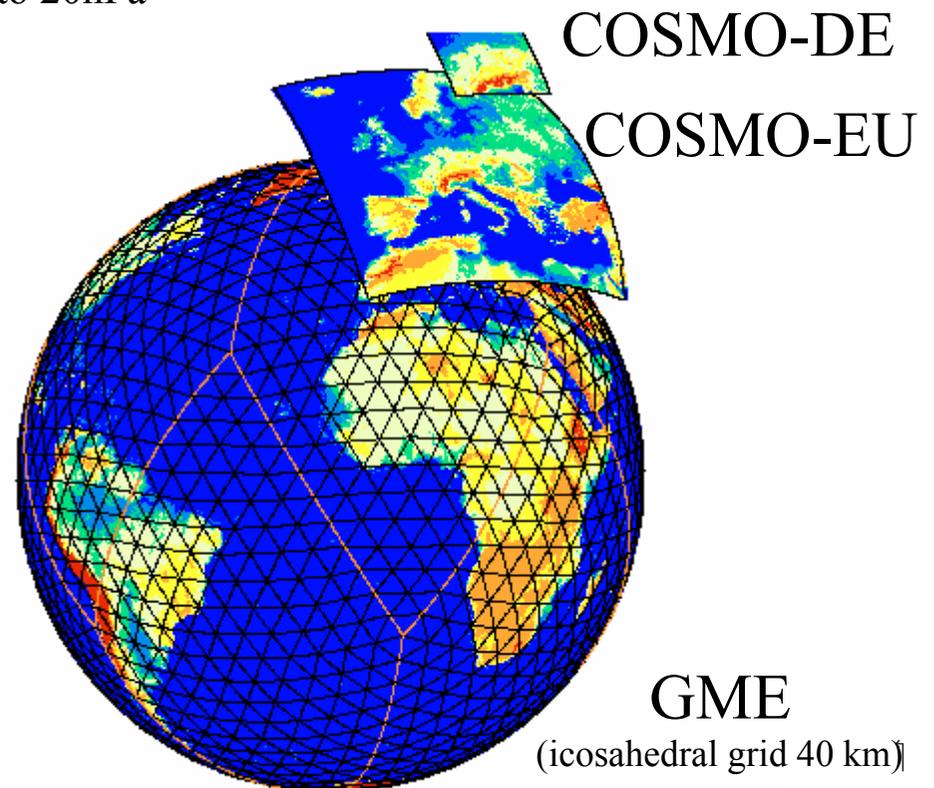
**i.e.** RETRIEVE temperature and humidity profiles and then nudge them as “pseudo”-observations

## Lokal Model COSMO-EU (LME)

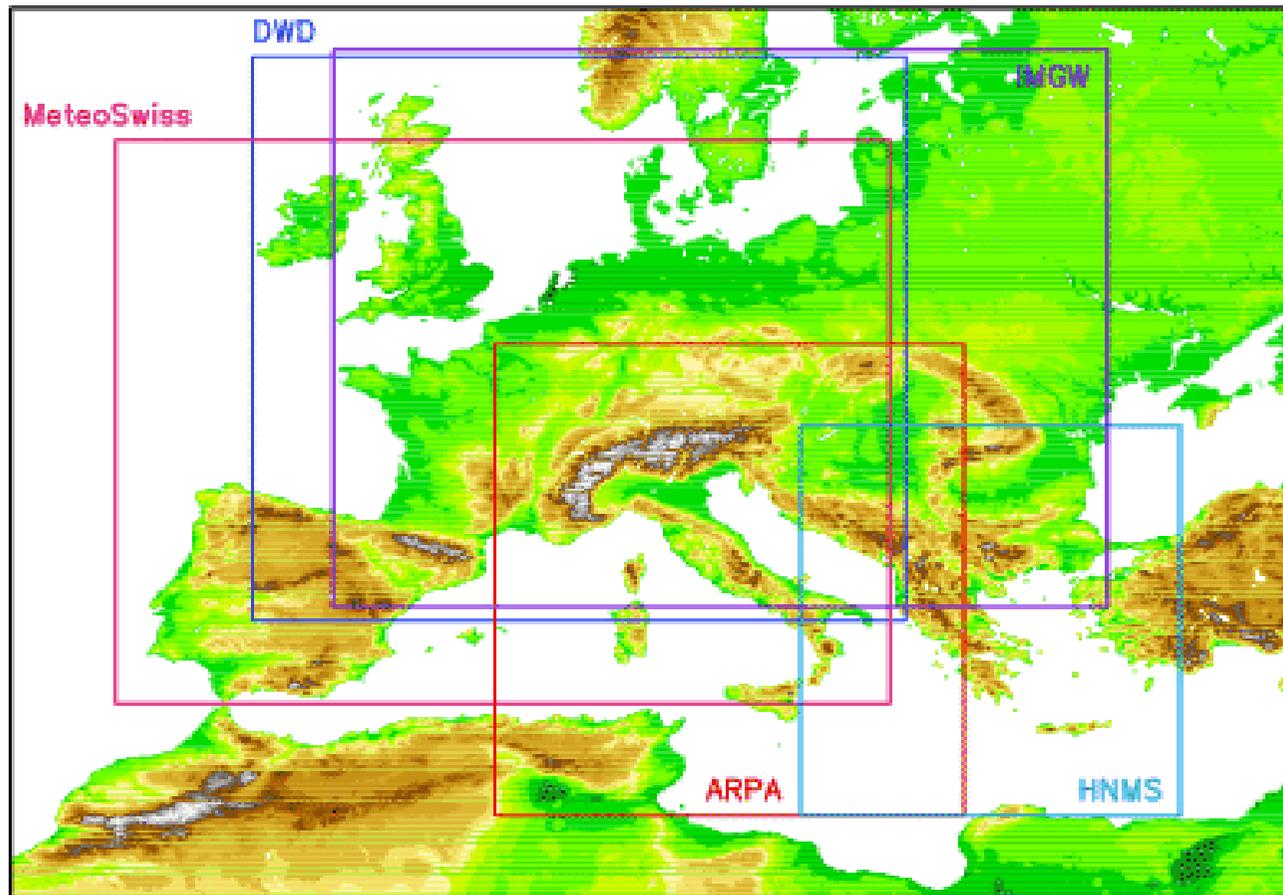
**COSMO-EU** (regional model): non-hydrostatic, rotated latitude-longitude grid, mesh size 7km  
terrain following hyb. coordinate, 40 layers up to 20hPa  
forecast range: 78 for initial dates 00, 12  
prognostic cloud ice, prognostic rain  
boundary values from GME

Analysis: continuous nudging  
observations: conv., AIREP, AMDAR, ACARS  
cutoff: 2h30  
variational soil moisture analysis

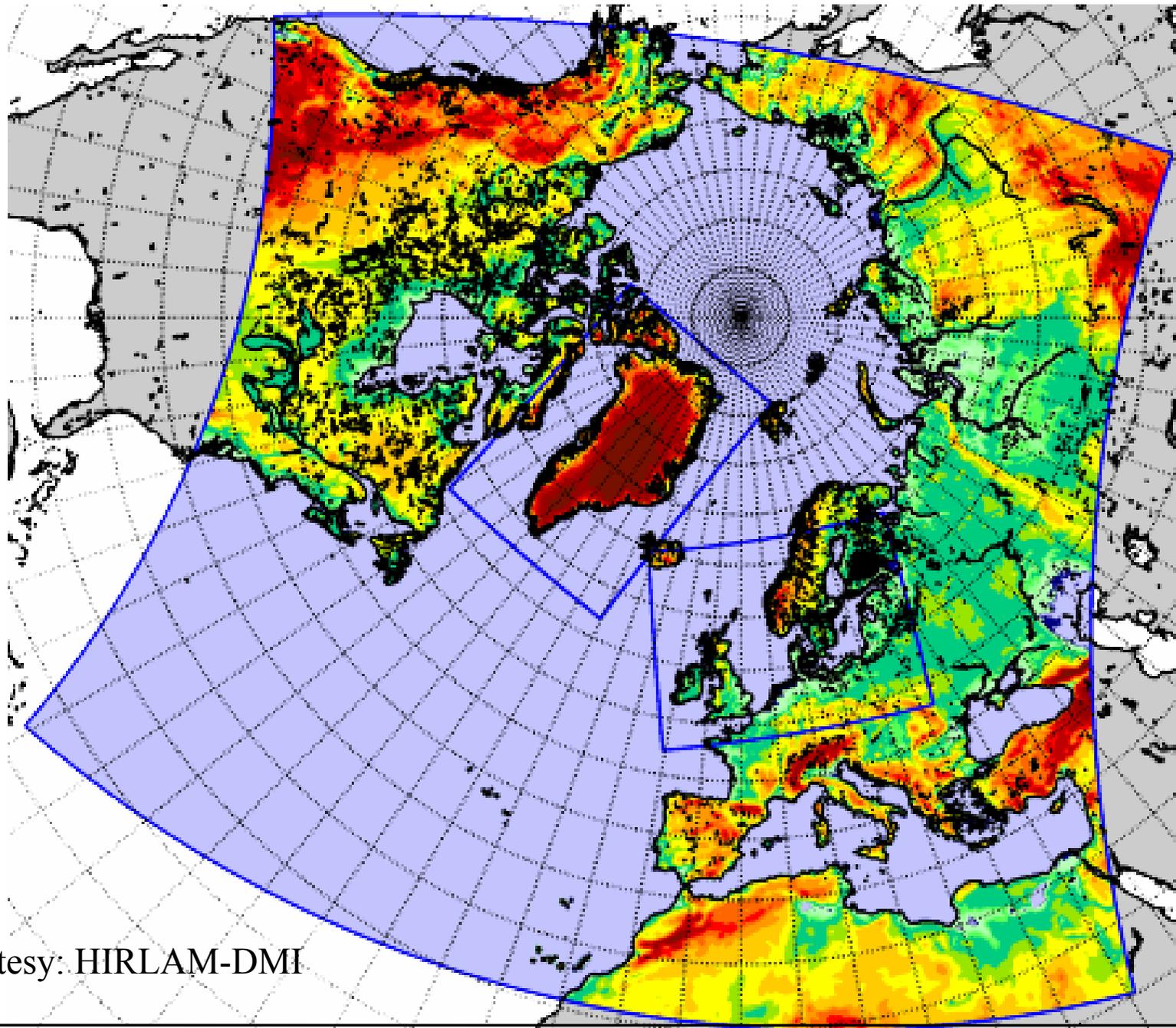
**COSMO-DE** (local model): similar to COSMO-EU  
mesh size 2.8km, explicit convection  
latent heat nudging of radar reflectivities  
boundary values from COSMO-EU



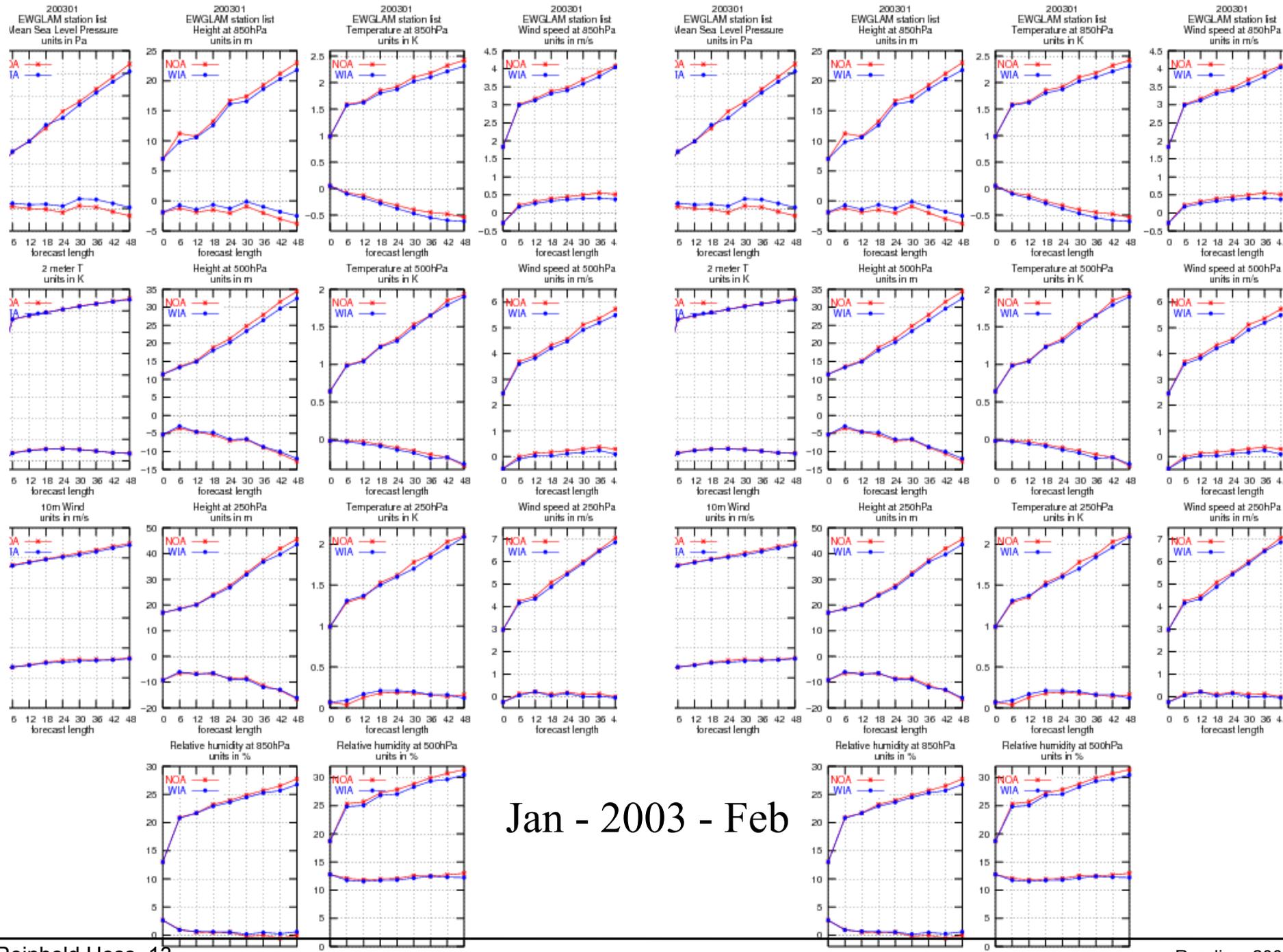
## Applications of COSMO-model within COSMO-group



→ ensembles



courtesy: HIRLAM-DMI



Jan - 2003 - Feb

## Outline of the talk

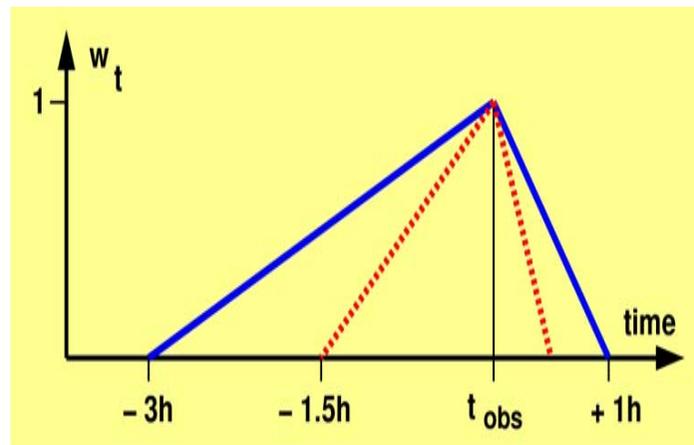
- ✓ Introduction: Forecast quality of limited area models
- ✓ COSMO-Project: assimilation of radiances for COSMO-EU (ATOVS/AMSU-A, SEVIRI, IASI)
  - 1D-Var and nudging
  - background errors
  - bias correction
  - first guess above model top (stratosphere)
  
  - preliminary results for AMSU-A
  - preliminary results for SEVIRI/MSG
  
  - conclusions

## Use of nonlinear operators with Nudging at appropriate time

conventional observations:

nudge observation 1.5 h before (and 30 min after) observation time with temporal weighting depending on time difference to observation

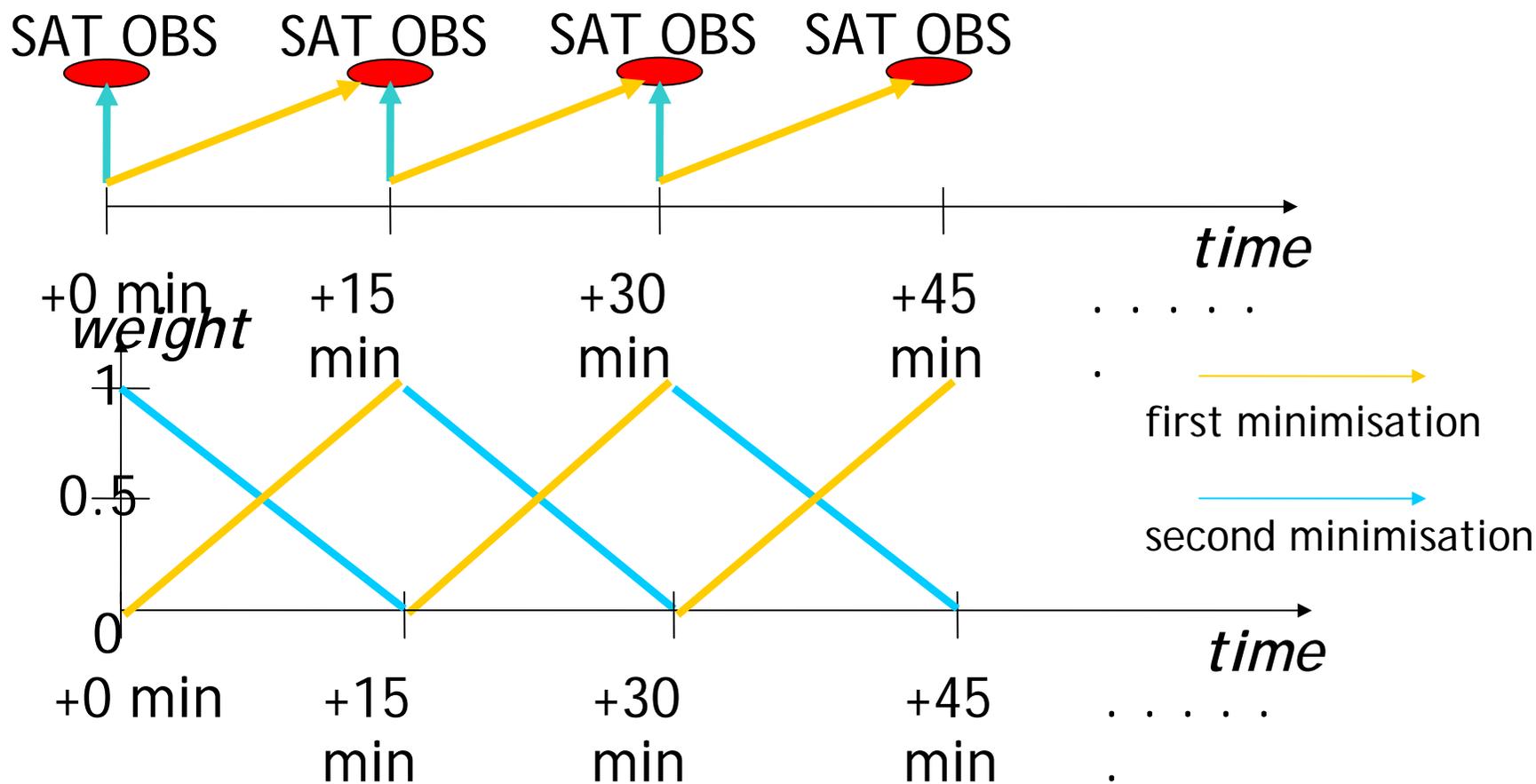
### Temporal Weighting



**preliminary retrievals** have to be computed for nonlinear observations  
use first guess available -1.5 h before observation time  
repeat retrieval every 30 min until nudging analysis reach observation time  
**attention:** first guess and observation become correlated!!!

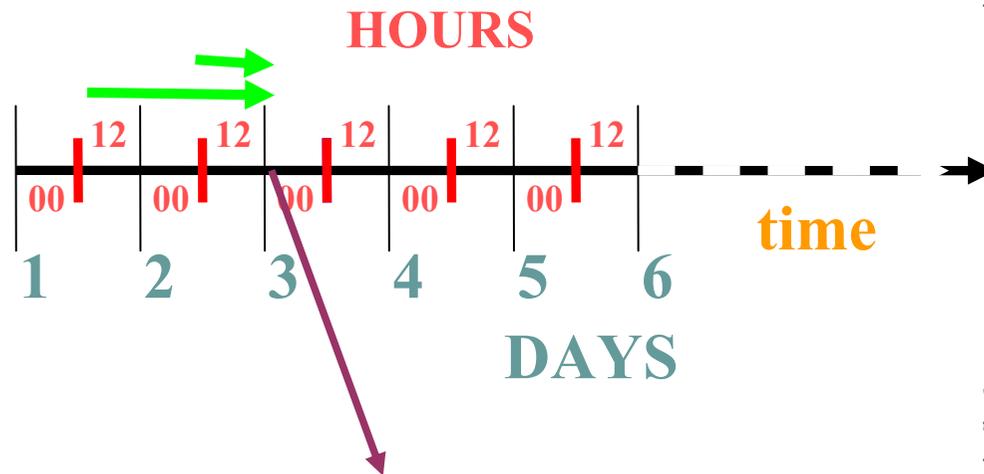
# Use of nonlinear operators with Nudging for observations with high temporal resolution

e.g. MSG/SEVIRI with time resolution of 15 min

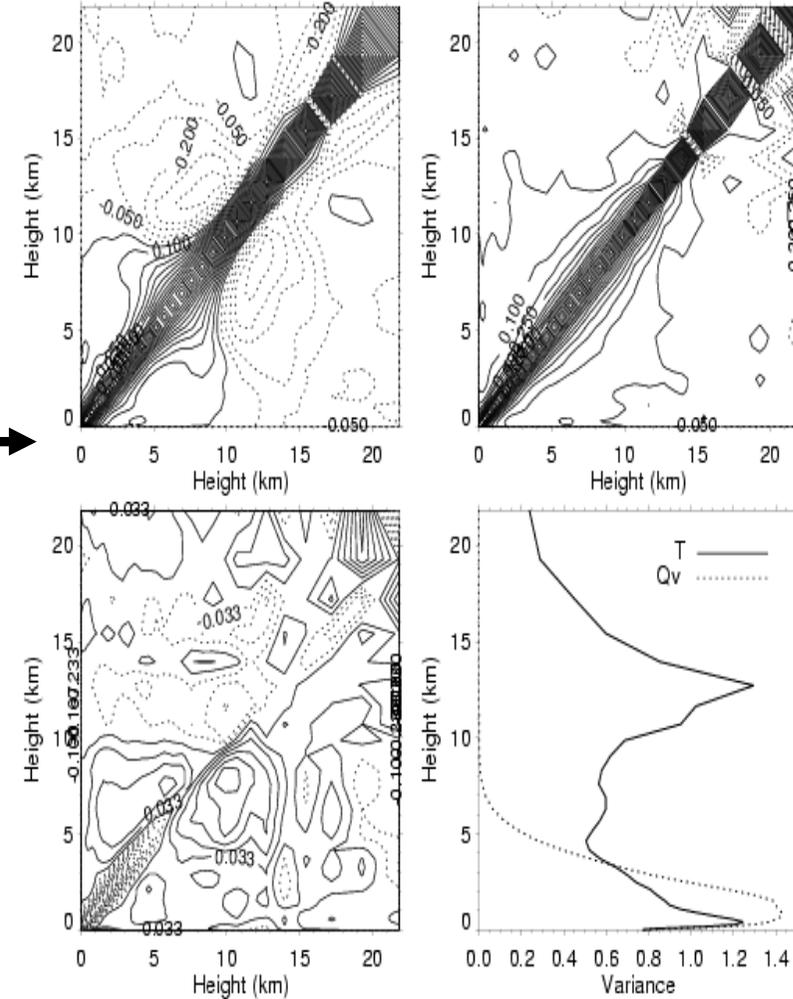


# Error covariance matrix B

The B matrix is calculated using forecast comparisons at +12h and +36h averaged over three months worth of data.



Hour of comparison of the two forecasts



## Background error covariance matrix $B$

complication: error structures from boundary values

- standard NMC-method  
(large error structures in statistics do not reflect small scale errors)
- lagged NMC-method (ALADIN): use identical boundary values  
(use boundary values from the same run of the embedding model)  
no boundary errors lead to error statistics of smaller scale
- ensemble  $B$ : pertubated observations or physis  
error statistics somehow in between standard and lagged NMC-method

However:

- less constraints on  $B$  (geostrophy, hydrostacy)
- high variation in error structures, more motivation for  
situation dependent, flow dependent or adaptive error structures



## Bias Correction for ATOVS

scanline and airmass dependent correction (Eyre, Harris & Kelly)

scanline correction: what is reasonable sampling size?

→ variance of a mean variable  $\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$  is  $\sigma_{\bar{x}}^2 = \frac{1}{n^2} \sum_{i=1}^n \sigma_i^2 = \frac{\sigma^2}{n}$

for given error variance the required sample size is  $n \geq \frac{\sigma^2}{\sigma_{\bar{x}}^2}$

→ application to obs – fg brightness temperatures:  $\Delta y_i = (y_i^o - y_i^f)$

(statistics are for each individual fov)

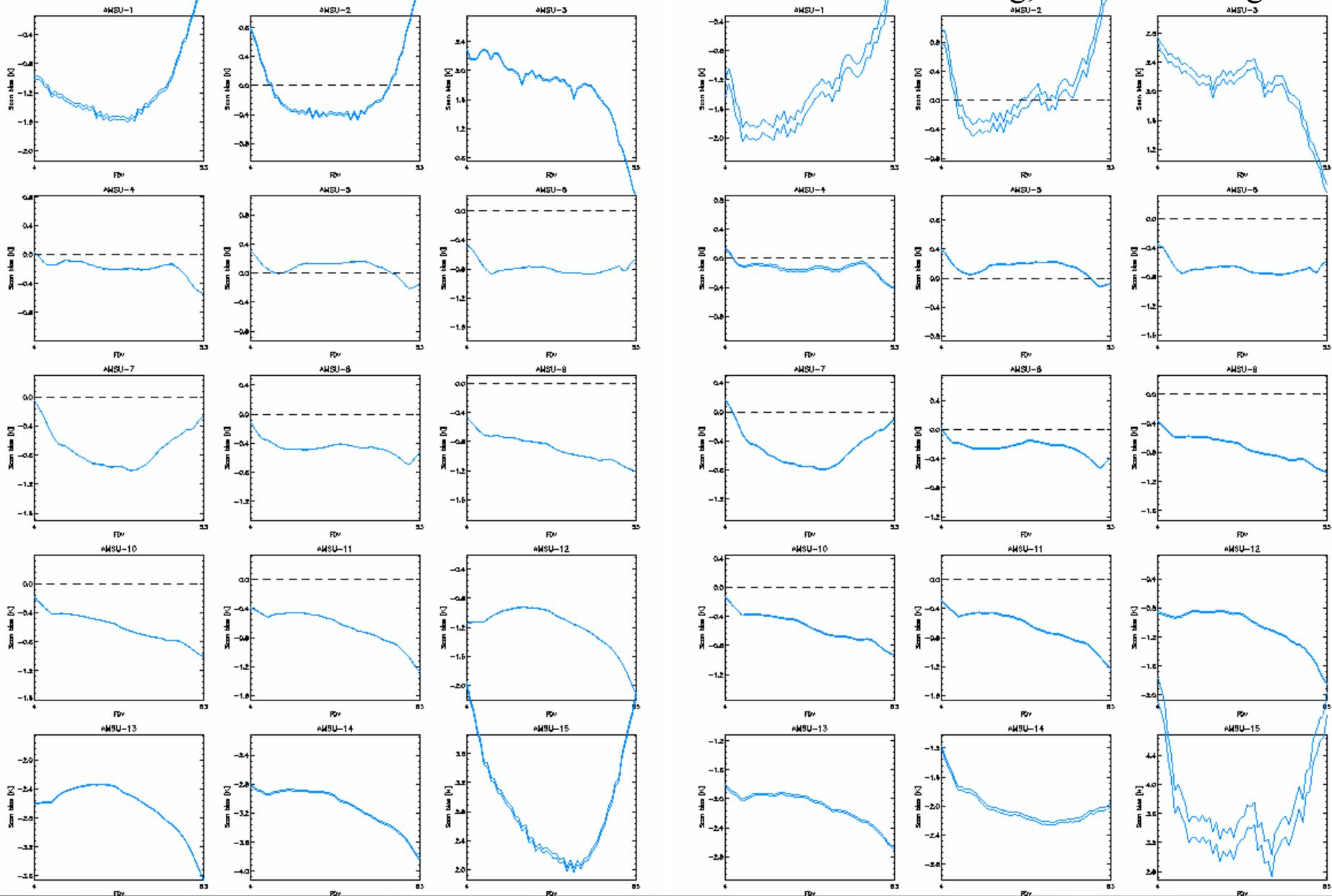
<b>examples:</b>	channel	AMSU-A 6	AMSU-A 4	AMSU-A 3	for $\sigma_{\bar{x}} \leq 0.01 \text{ K}$
	$\sigma_x =$	0.2 K	0.4 K	2.1 K	
	n =	400	1600	21000	

- time to obtain required sample sizes for individual fovs depend on model area (size over sea)
- for COSMO-EU two weeks for most relevant temperature sounding channels
- what about **representativity** (synoptic scenarios, seasonal changes) ?

# scanline biases AMSU/NOAA 18 (15 to 25 June 2007)

GME all areas

GME lat 30 to 60 deg, lon: -30 to 0 deg



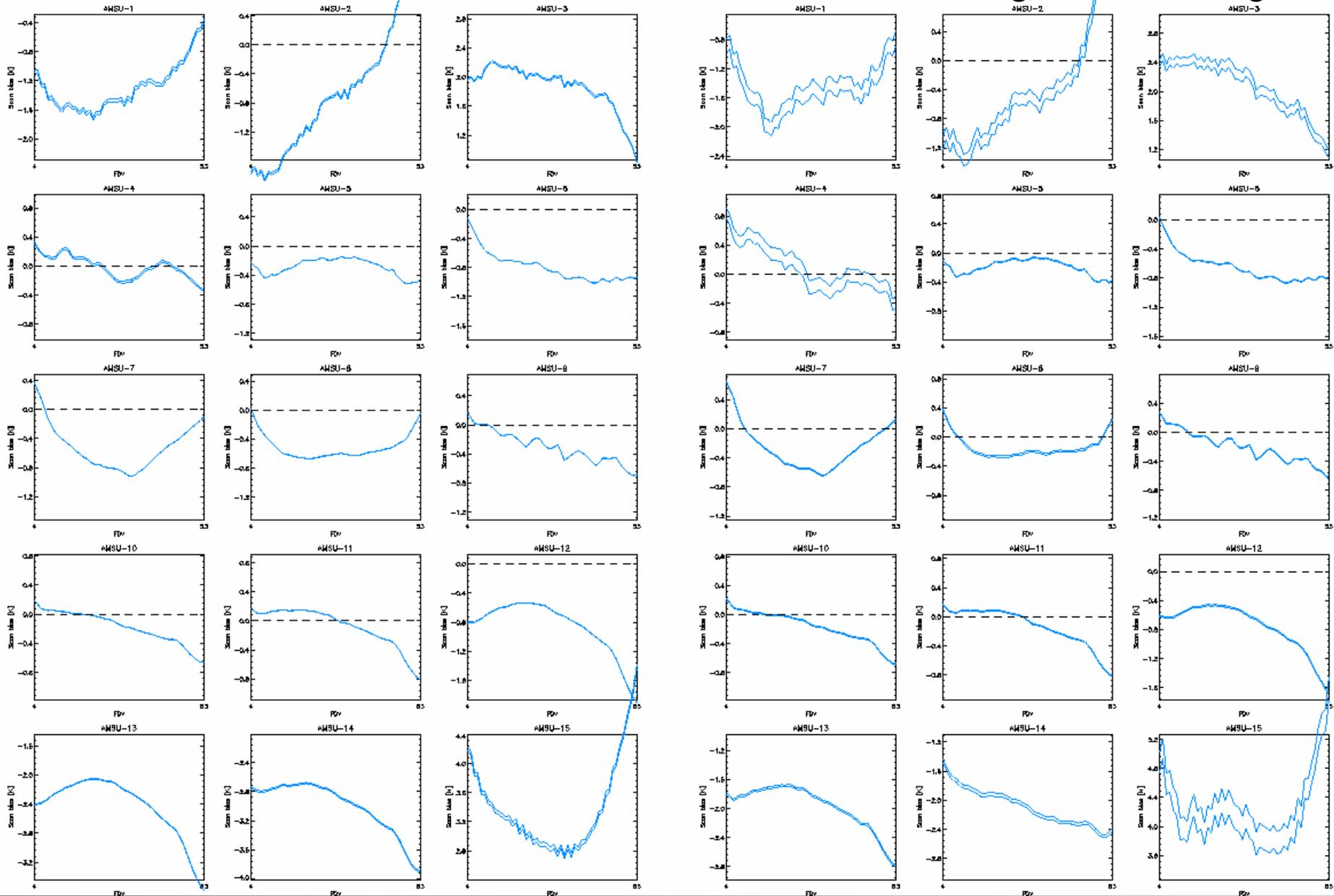
Reinhold Hess, 20 approx 50000 obs/fov

approx 1500 obs/fov Reading, 2007

# scanline biases AMSU/NOAA 16 (15 to 25 June 2007)

GME all areas

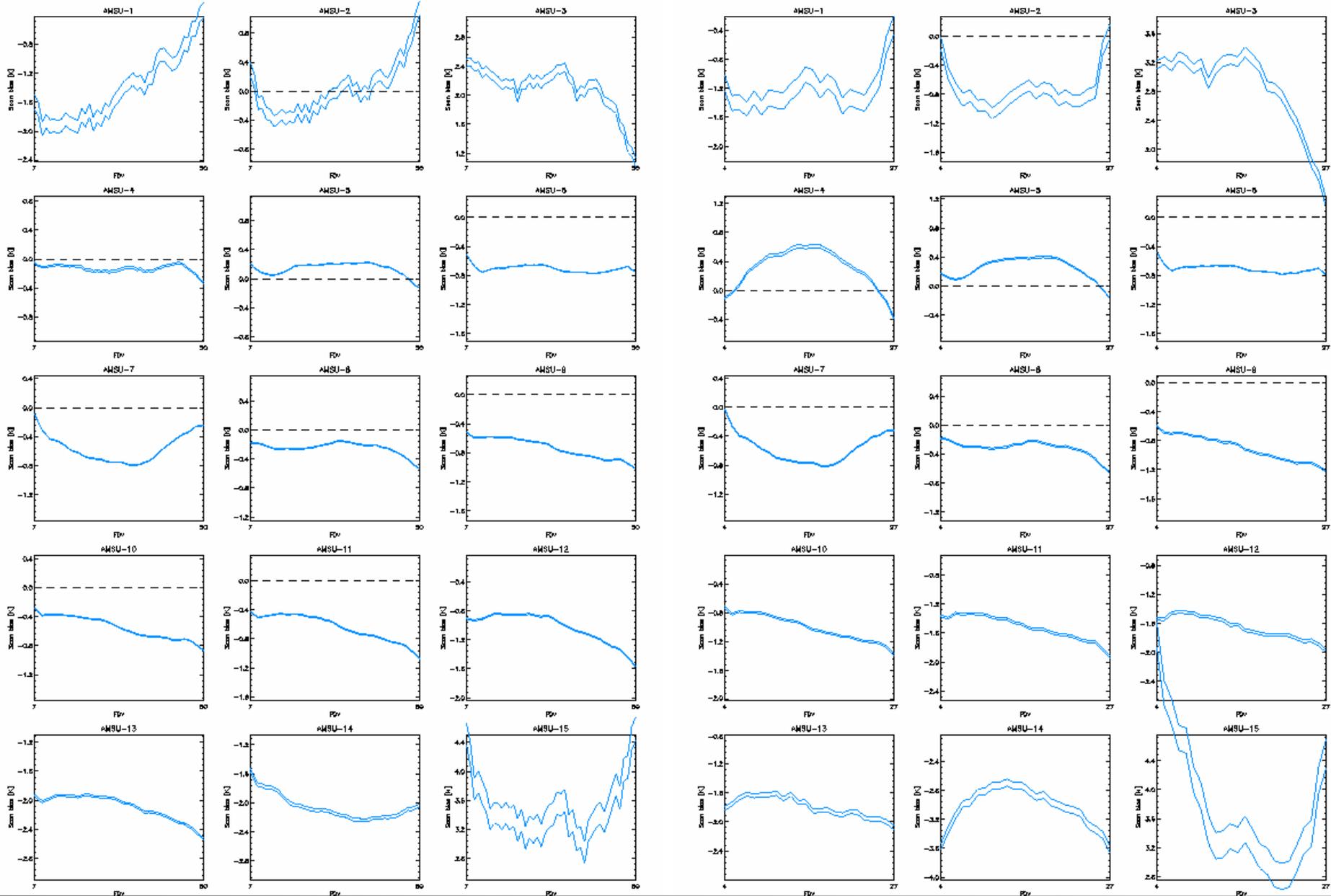
GME lat 30 to 60 deg, lon -30 to 0 deg



# scanline biases AMSU/NOAA 18 (15 to 25 June 2007)

GME lat 30 to 60 deg, lon:-30 to 0 deg

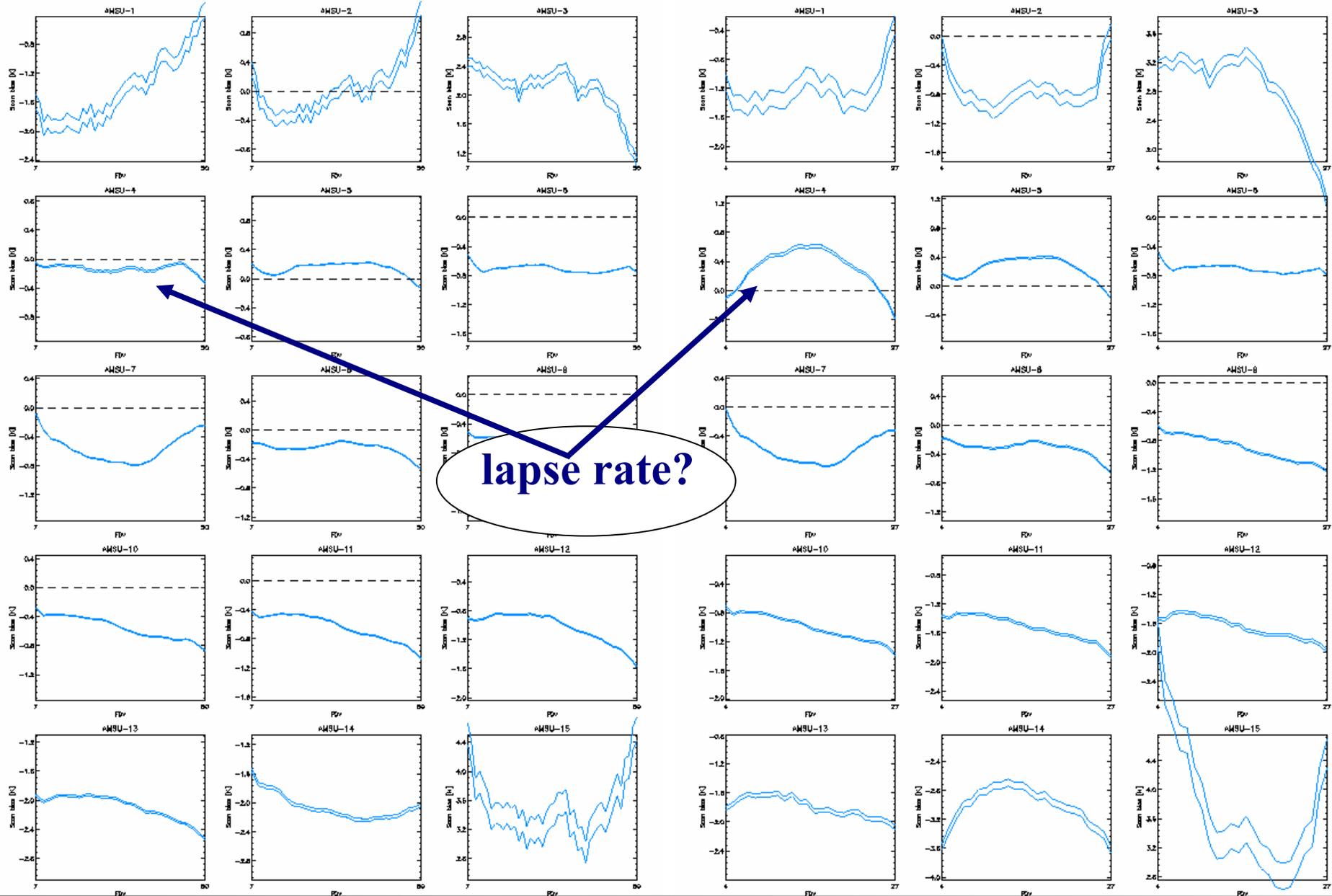
COSMO-EU: approx 1200-1500 fovs



# scanline biases AMSU/NOAA 18 (15 to 25 June 2007)

GME lat 30 to 60 deg, lon:-30 to 0 deg

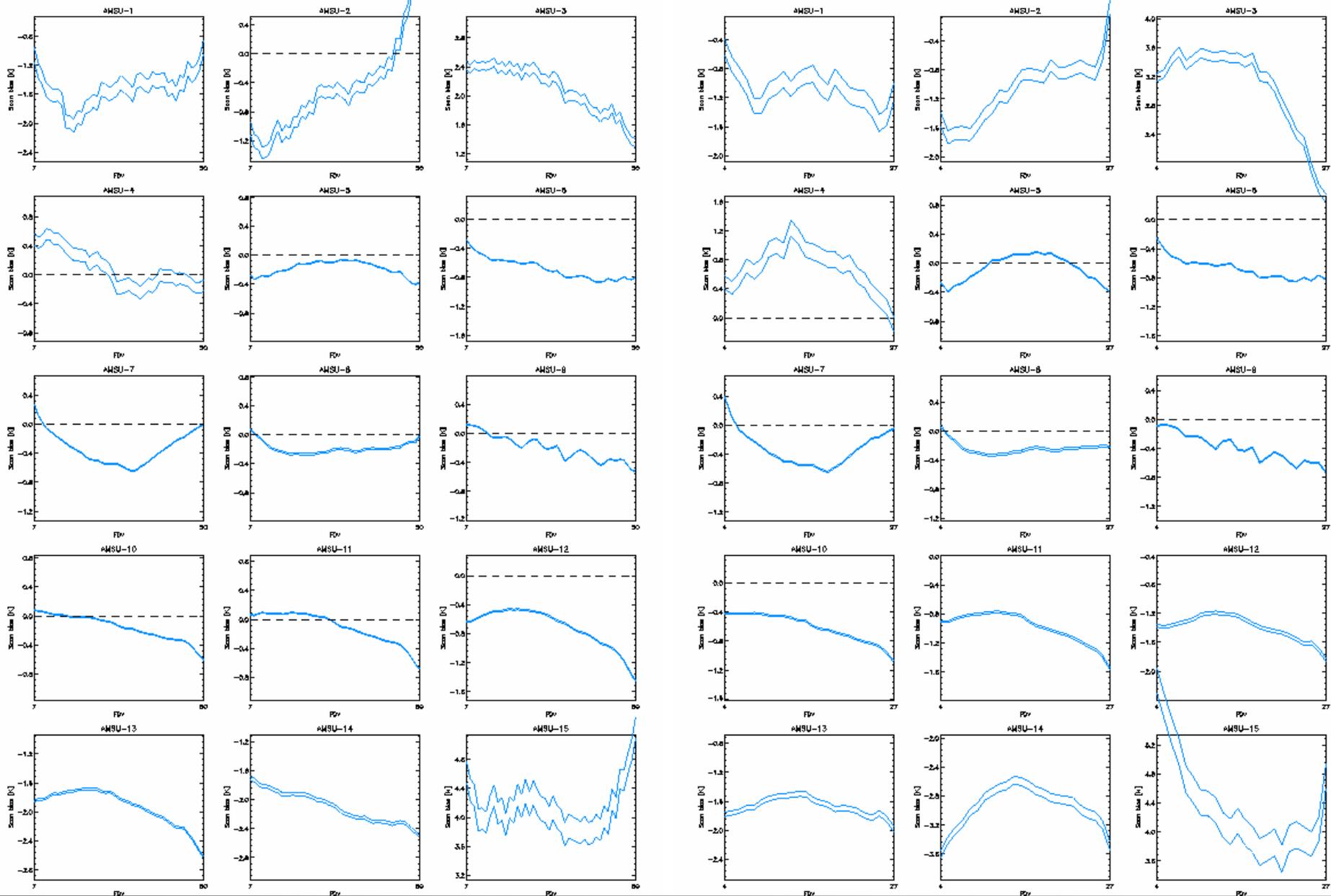
COSMO-EU: approx 1200-1500 fovs



# scanline biases AMSU/NOAA 16 (15 to 25 June 2007)

GME lat 30 to 60 deg, lon:-30 to 0 deg

COSMO-EU: approx 1200-1500 fovs



Reinhold Hess, 24 approx 1200 obs/fov

approx 1200-1500 obs/fov Reading, 2007

## Bias Correction for ATOVS

### scanline correction

- irregular shape
- constant with area (for most channels, no latitude dependency)
- sample size not too critical
- representativity seems no issue
- GME and COSMO-EU show significant differences only for surface (and humidity channels)
- no significant influence of interpolation with ATOVPP/AAPP (6 resp. 4 side fovs removed)

### air mass correction

- idea: air mass bias is situation dependent, model these biases using meteorological predictors
- choices of predictors:
  - ◆ AMSU-A 5 and 9 (observed or simulated)
  - ◆ mean temperatures (50-200 hPa and 300-1000 hPa), SST, IWV
  - ◆ predictors for scanline correction: zenith angle, square of zenith angle (or remove scanline correction before air mass correction)
  - ◆ latitude as predictor/coefficients variable with latitude band

**What is a good choice of predictors?**

# air mass correction, choice of predictors

**example:** using two predictors p1 and p2 and a constant fit parameters c1, c2 and c3 to obs-fg differences:

$$\text{minimise: } F(c_1, c_2, c_3) = \frac{1}{2} \sum_{i=1, \dots, n} (y_i^o - y_i^{fs} - c_1 p_{1,i} - c_2 p_{2,i} - c_3)^2$$

$$\text{solution of } \frac{\partial F}{\partial c_1} = \frac{\partial F}{\partial c_2} = \frac{\partial F}{\partial c_3} = 0 \text{ provides } c_1, c_2, c_3 (\Delta y_{i=1, \dots, n}) \text{ with } \Delta y_i = (y_i^o - y_i^{fs})$$

## Gaussian error analysis:

What are the errors of c1, c2 and c3 for given  $\Delta y_i = (y_i^o - y_i^{fs})$  with errors  $\sigma_e$  ?

$$\text{derivatives of } c_1, c_2, c_3 (\Delta y_{i=1, \dots, n}) \text{ are } \begin{pmatrix} \frac{\partial c_1}{\partial \Delta y_i} \\ \frac{\partial c_2}{\partial \Delta y_i} \\ \frac{\partial c_3}{\partial \Delta y_i} \end{pmatrix} = -(\text{Hess } F)^{-1} \begin{pmatrix} \frac{\partial^2 F}{\partial c_1 \partial \Delta y_i} \\ \frac{\partial^2 F}{\partial c_2 \partial \Delta y_i} \\ \frac{\partial^2 F}{\partial c_3 \partial \Delta y_i} \end{pmatrix}$$

(chain rule for implicit functions)

$$\text{approximated errors: } \begin{pmatrix} \Delta c_1 \\ \Delta c_2 \\ \Delta c_3 \end{pmatrix} = \begin{pmatrix} \sum_{i=1, \dots, n} \frac{\partial c_1}{\partial \Delta y_i} \Delta y_i \\ \sum_{i=1, \dots, n} \frac{\partial c_2}{\partial \Delta y_i} \Delta y_i \\ \sum_{i=1, \dots, n} \frac{\partial c_3}{\partial \Delta y_i} \Delta y_i \end{pmatrix}$$

$$\text{covariance: } \text{Kov} \begin{pmatrix} \Delta c_1 \\ \Delta c_2 \\ \Delta c_3 \end{pmatrix} = -(\text{Hess } F)^{-1} \sigma_e^2 \quad \text{Hess } F = \begin{pmatrix} \sum p_{1,i}^2 & \sum p_{1,i} p_{2,i} & \sum p_{1,i} \\ \sum p_{1,i} p_{2,i} & \sum p_{2,i}^2 & \sum p_{2,i} \\ \sum p_{1,i} & \sum p_{2,i} & n \end{pmatrix}$$

## air mass correction, choice of predictors

Hessian and error correlation matrix only depend on predictors  
- same for all channels

using AMSU-A 4, 9 and a constant as predictors:

• COSMO-EU (15-25.6.07): 
$$\text{Cor} = \begin{pmatrix} 1,00 & 0,94 & -0,99 \\ 0,94 & 1,00 & -0,98 \\ -0,99 & -0,98 & 1,00 \end{pmatrix}$$

• GME (15-25.6.07), lim area: 
$$\text{Cor} = \begin{pmatrix} 1,00 & 0,88 & -0,98 \\ 0,88 & 1,00 & -0,95 \\ -0,98 & -0,95 & 1,00 \end{pmatrix}$$

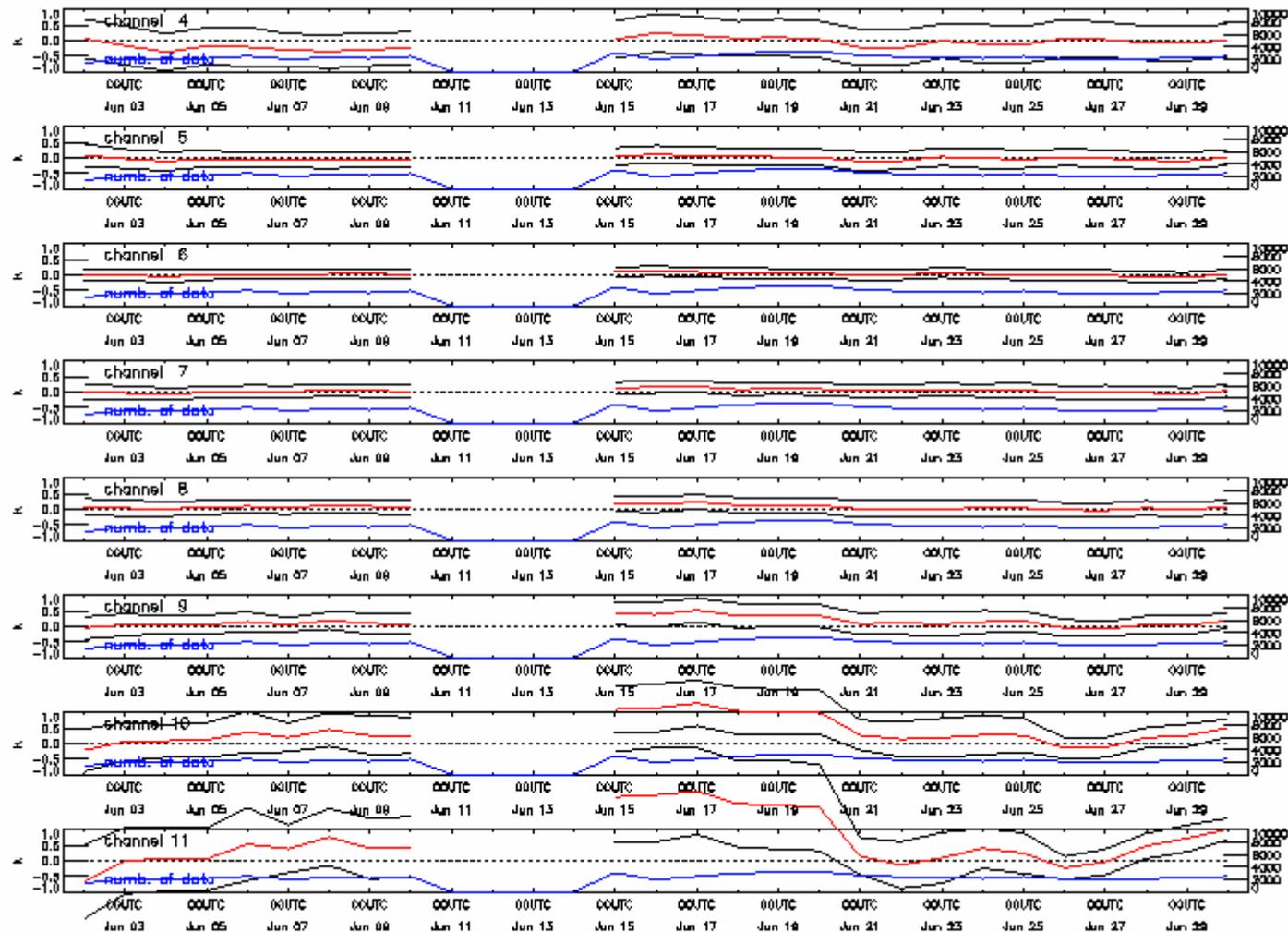
• GME (15-25.6.07), all area: 
$$\text{Cor} = \begin{pmatrix} 1,00 & 0,60 & -0,91 \\ 0,60 & 1,00 & -0,87 \\ -0,91 & -0,87 & 1,00 \end{pmatrix}$$

errors of predictors are highly correlated (in this example)  
overfitting – useless, loose information, by removing signal  
other choice of predictors  
longer time series for higher representativity  
usually: **several months** of data

biases depend on synoptic situation, season

# timeserie of bias corrected observations minus first guess

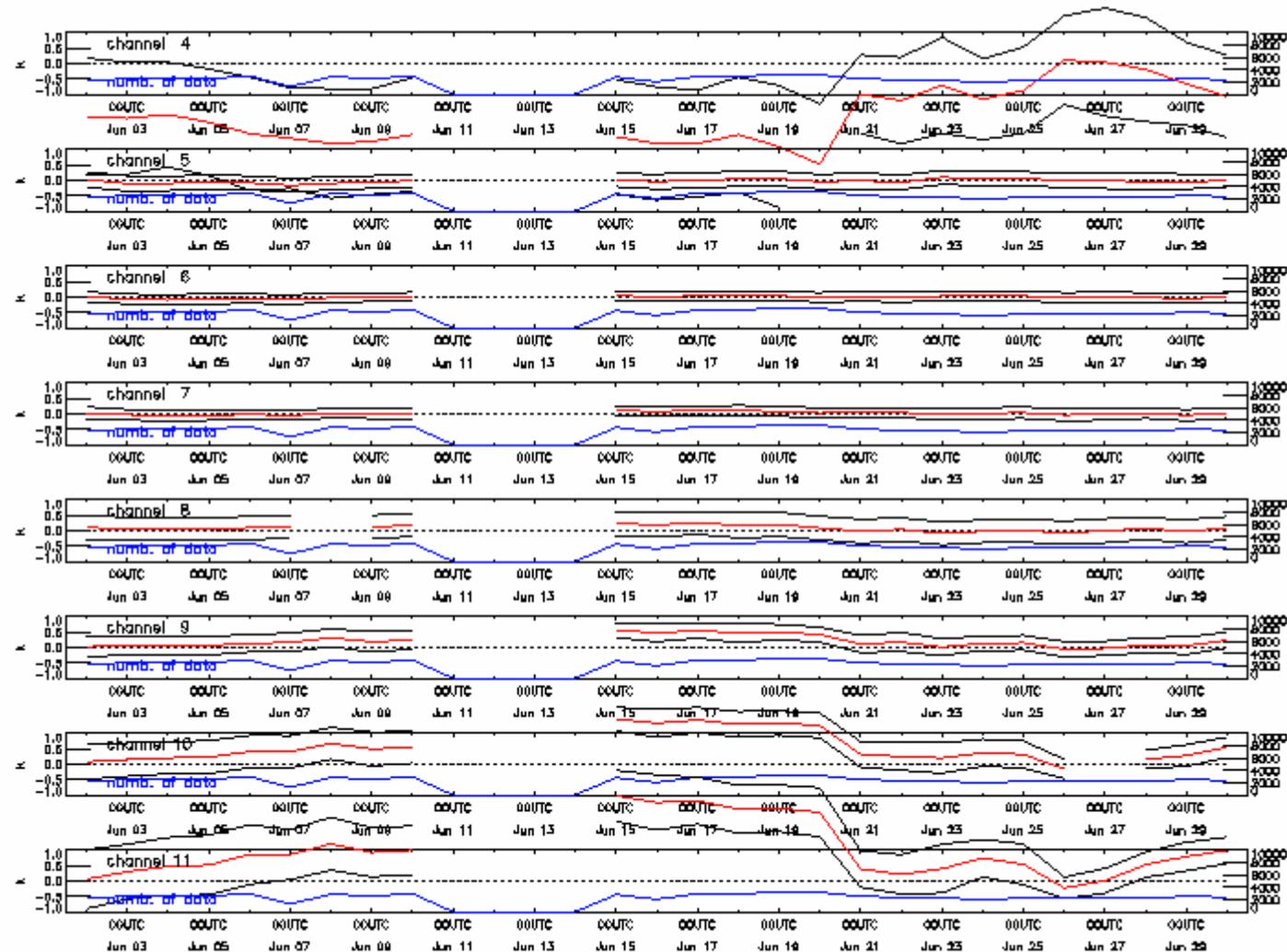
## AMSU-A channels 4-11, NOAA-18



stable in the troposphere, however large variations for high sounding channels

# timeserie of bias corrected observations minus first guess

## AMSU-A channels 4-11, NOAA-16



stable in the stratosphere, however large variations for high sounding channels

## provide first guess values above model top

limited area models usually have lower model top than required by RTTOV (0.1hPa - 0.05hPa)

COSMO-EU: 30hPa

HIRLAM: 10hPa

ALADIN: 1hPa

- increase height of model top
- use climatological values (inaccurate, use only lower peaking channels)
- use forecasts of global model IFS (accurate, but timely receive of IFS forecasts required)
- linear regression of high peaking channels to model levels (Met Office)

$$y = W x$$

x: high peaking channels  
y: temperatures on RTTOV levels  
W: regression matrix

compute  $W$  with training data set

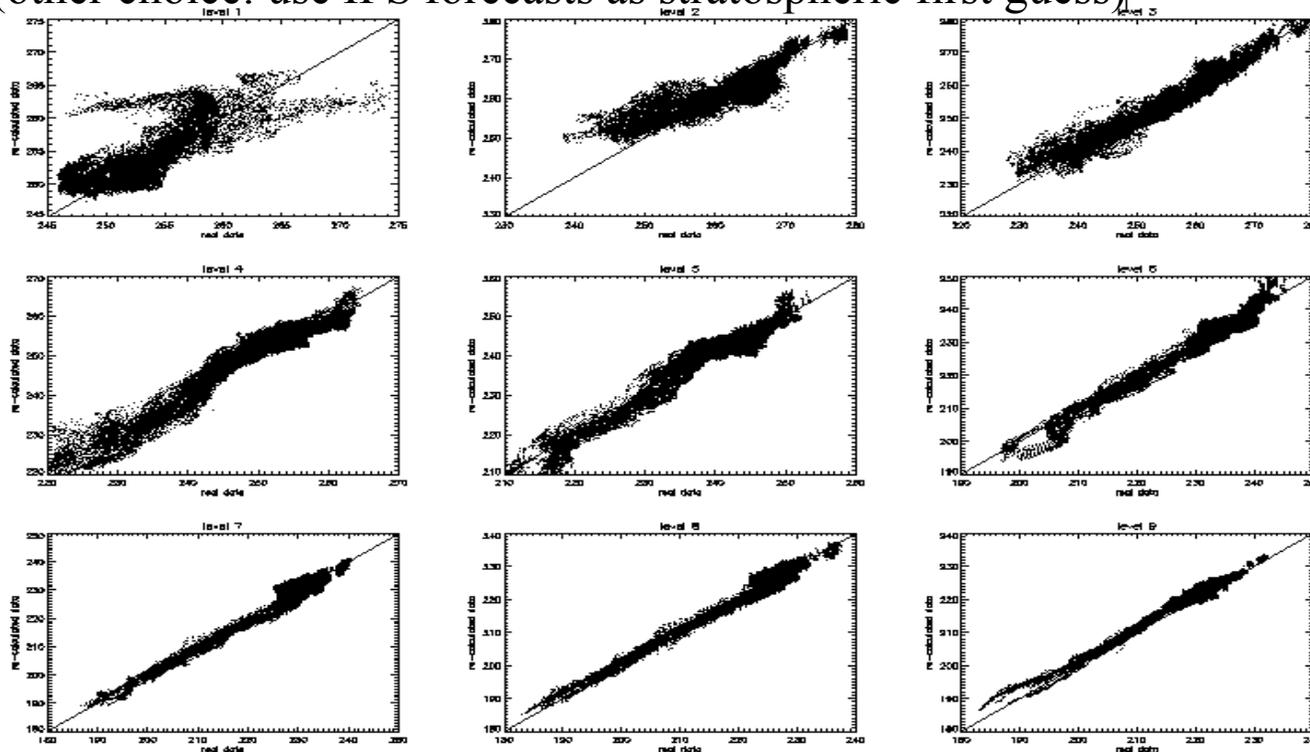
reasonable:

- no humidity
- more or less linear relation between, high peaking channels and level temperature (no clouds)

provide first guess values above model top (COSMO-EU: 30hpa)

→ use of climatological values (ERA40) seems not sufficient

→ linear regression of top RTTOV levels from stratospheric channels  
(other choice: use IFS forecasts as stratospheric first guess)

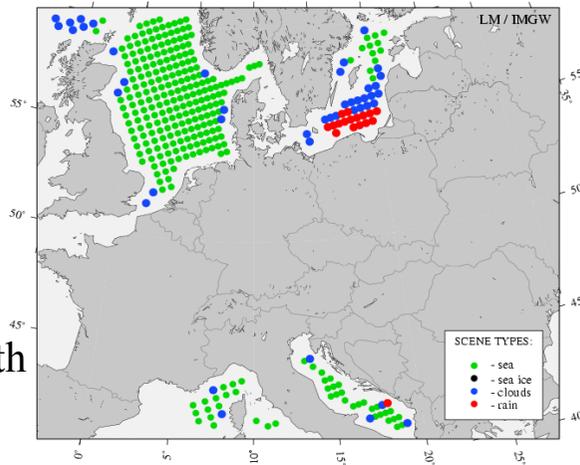


ECMWF profiles versus estimated profiles, top GME levels  
accuracy about 5K for lower levels, but ECMWF may have errors in stratosphere too

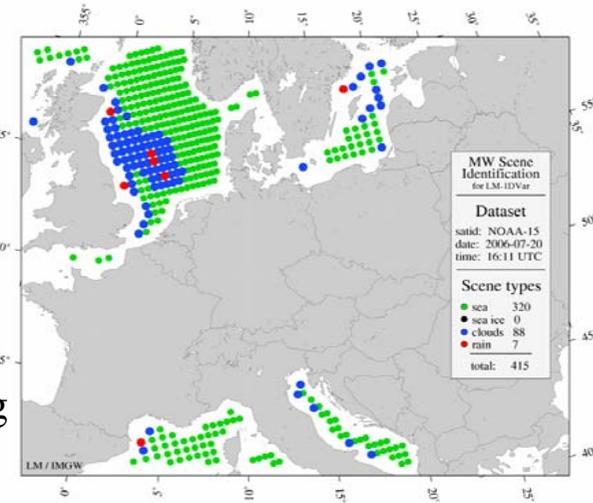
## 1D-Var for LME – Cloud and Rain detection

Microwave surface emissivity model: rain and cloud detection (Kelly & Bauer)

microwave scene identification  
NOAA-18 2006-08-05 11:59 UTC



Validation with  
radar data



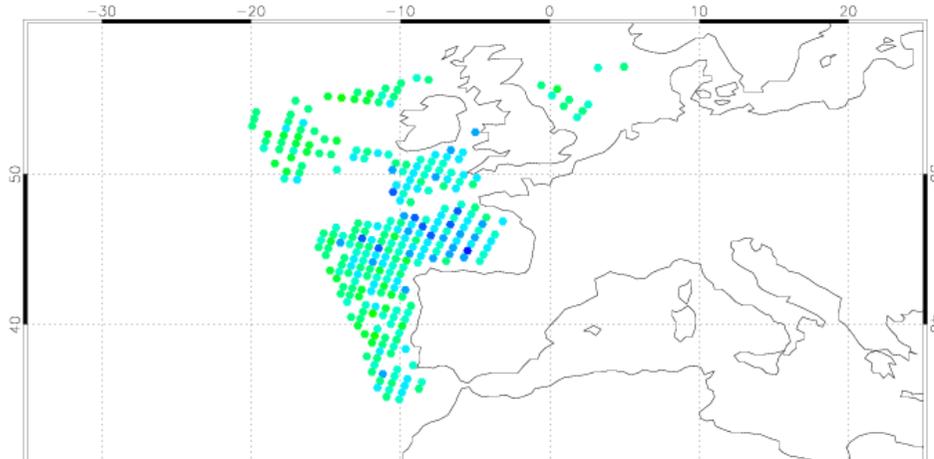
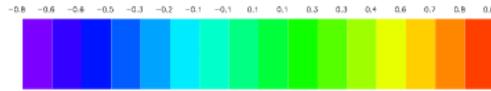
Validation  
with  
MSG imaging



# observation increments and resulting 1D-Var increments

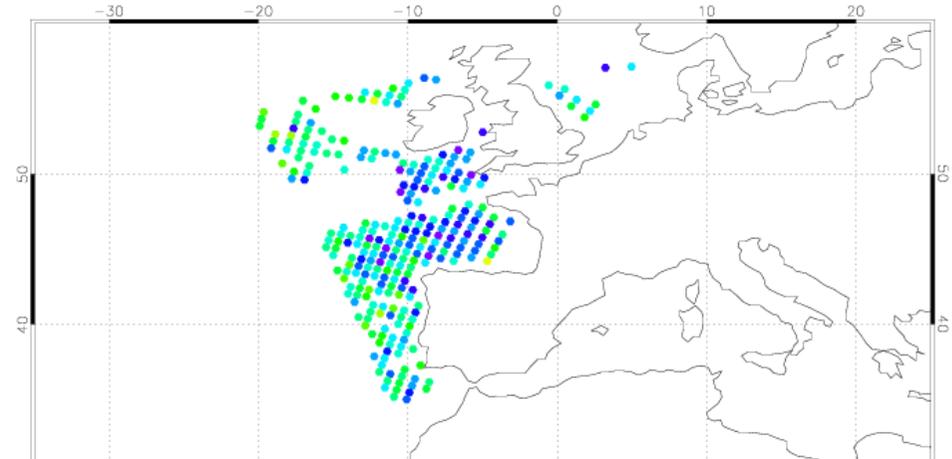
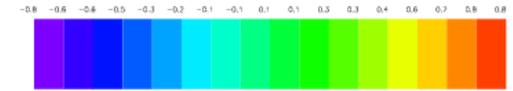
Corrected - Model BT

AMSU-A 6  
Date/Time: 200705100300



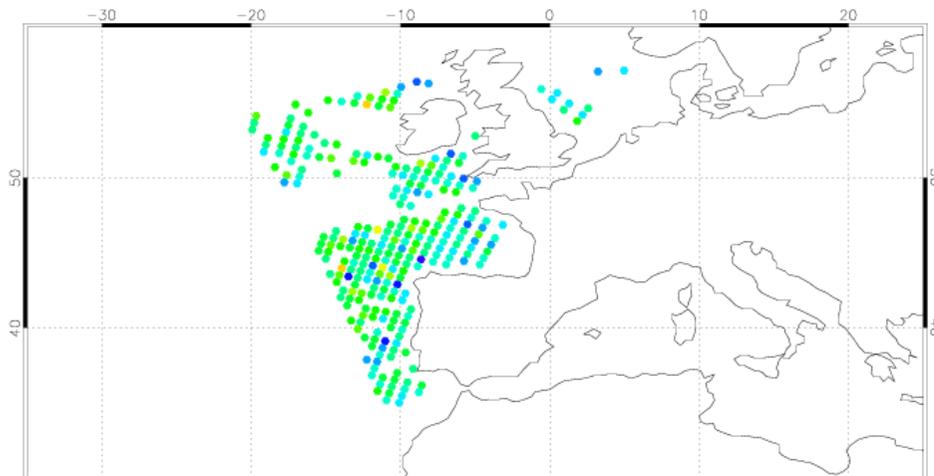
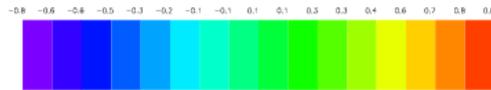
Retrieval - Backgr. T

Level 15  
Date/Time: 200705100300



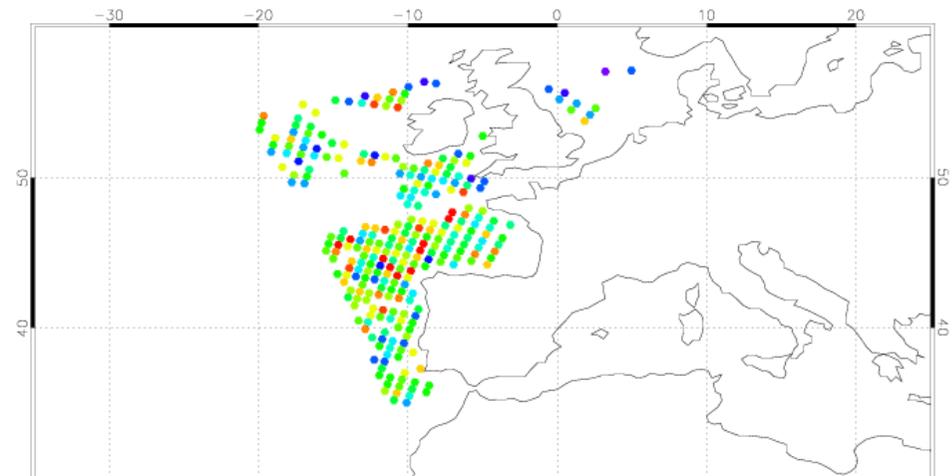
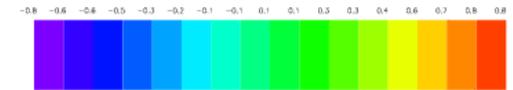
Corrected - Model BT

AMSU-A 5  
Date/Time: 200705100300



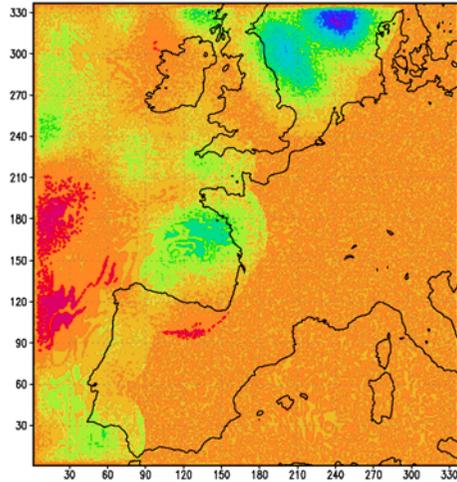
Retrieval - Backgr. T

Level 23  
Date/Time: 200705100300

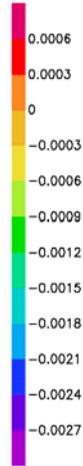
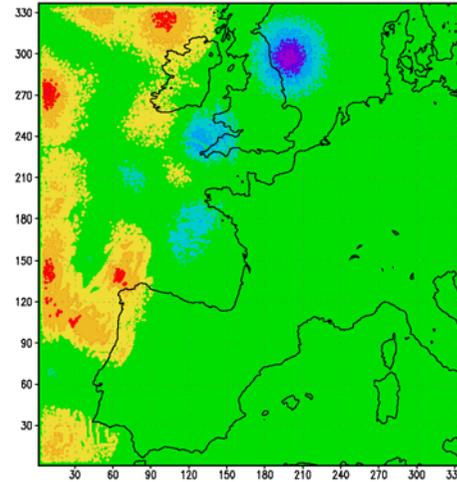


# $\Delta T$ -‘analysis increments’ from ATOVS, after 1 timestep (sat only), $k = 20$

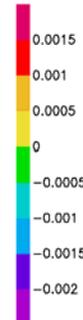
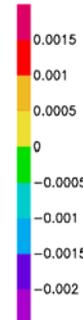
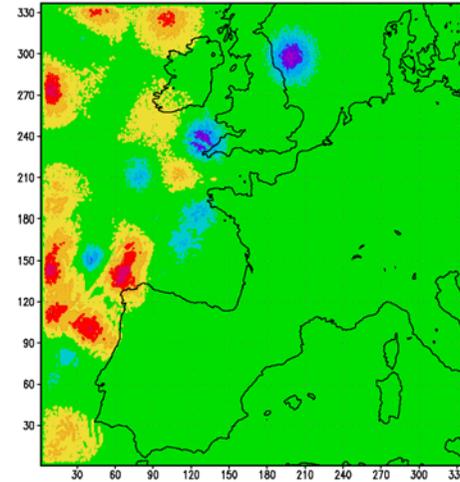
no thinning of 298 ATOVS



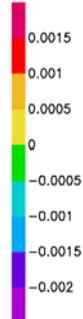
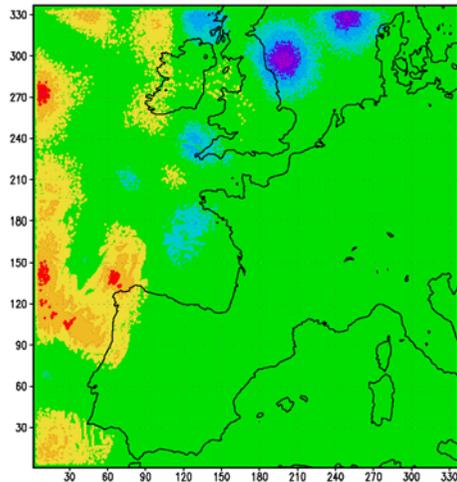
30 ATOVS by old thinning (3)



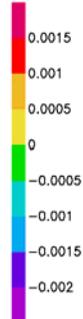
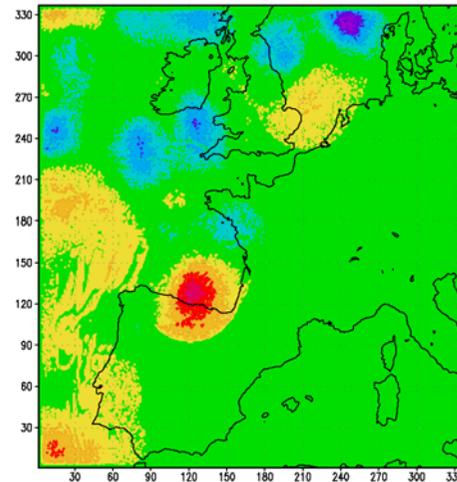
30 ATOVS, correl. scale 70%



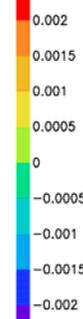
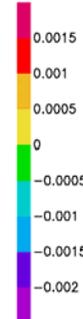
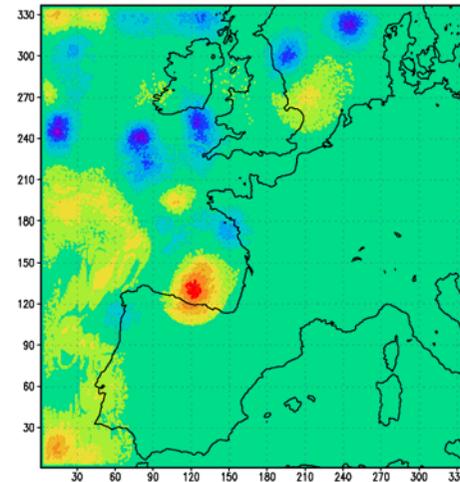
40 ATOVS by thinning (3)



82 ATOVS by thinning (2)



82 ATOVS, correl. scale 70%

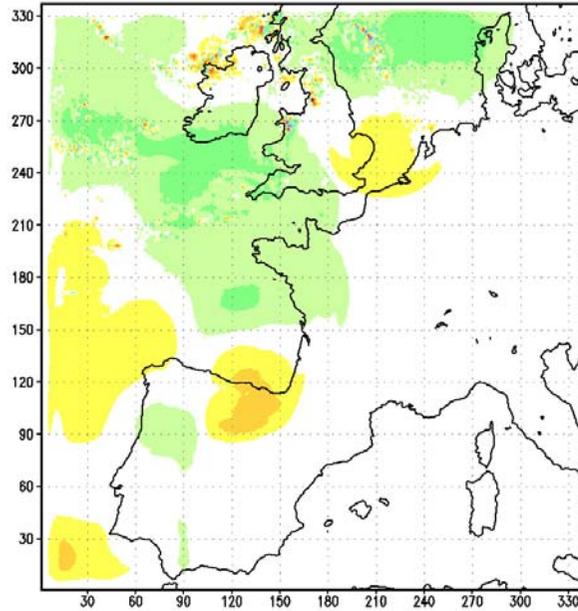


# $\Delta T$ -‘analysis increments’ from ATOVS, after 30 minutes (sat only), $k = 20$

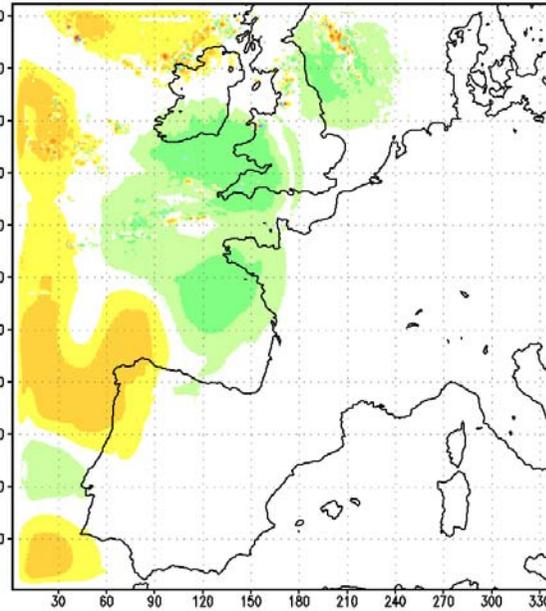
no thinning of 298 ATOVS

30 ATOVS by old thinning (3)

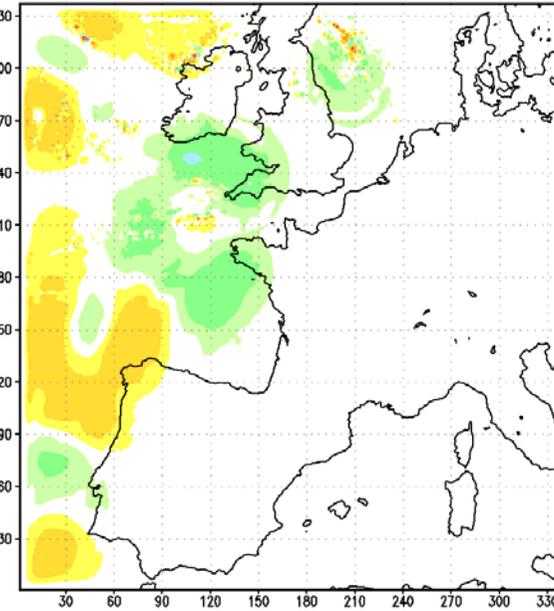
30 ATOVS, correl. scale 70%



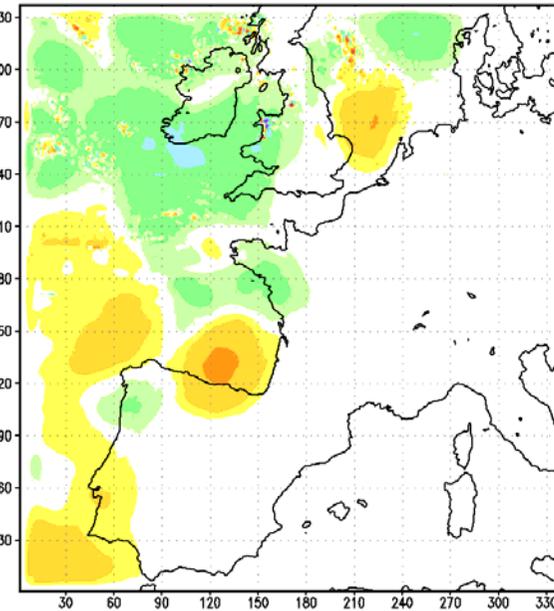
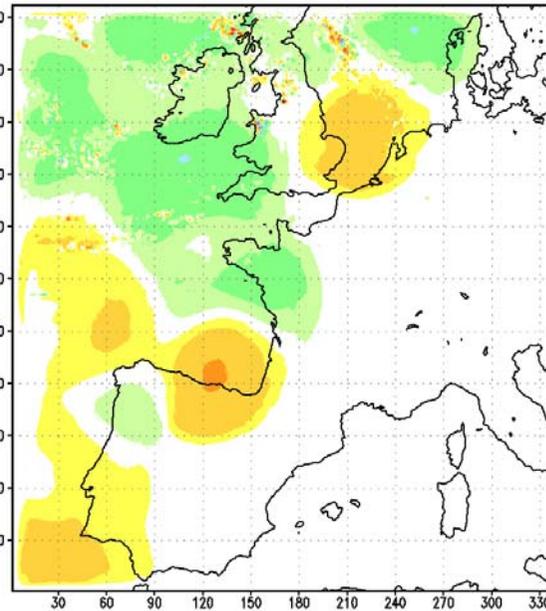
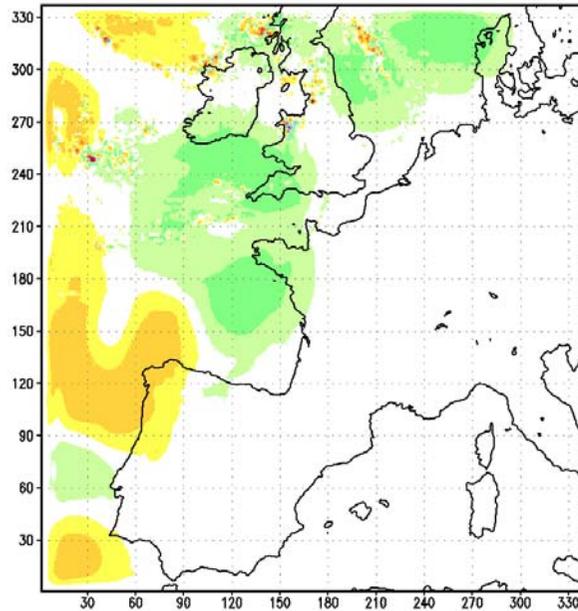
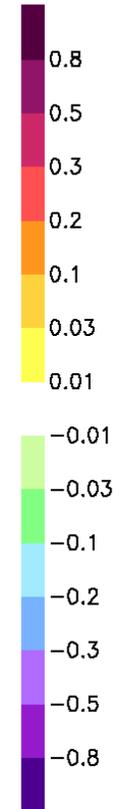
40 ATOVS by thinning (3)



82 ATOVS by thinning (2)



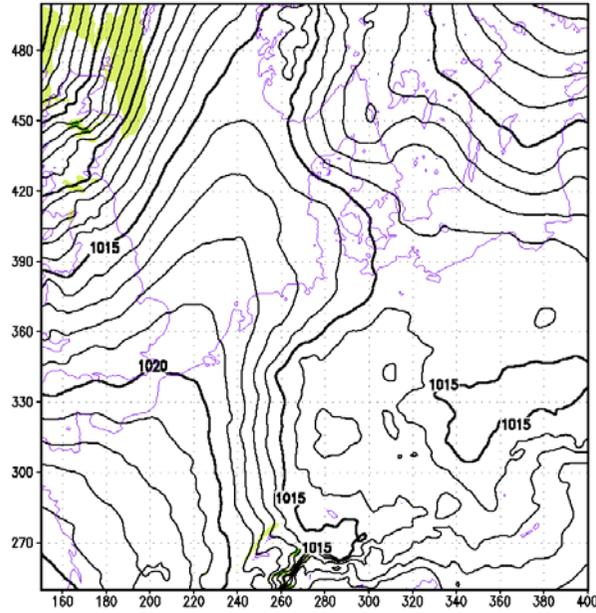
82 ATOVS, correl. scale 70%



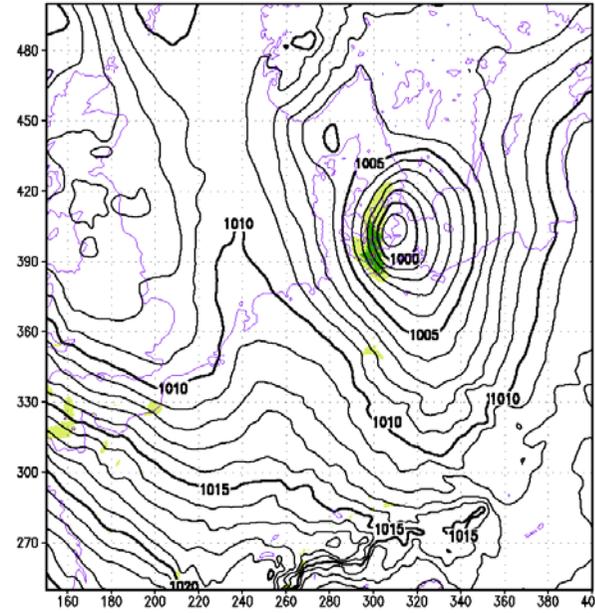
mean sea level pressure & max. 10-m wind gusts

valid for 17 May 2007 , 0 UTC

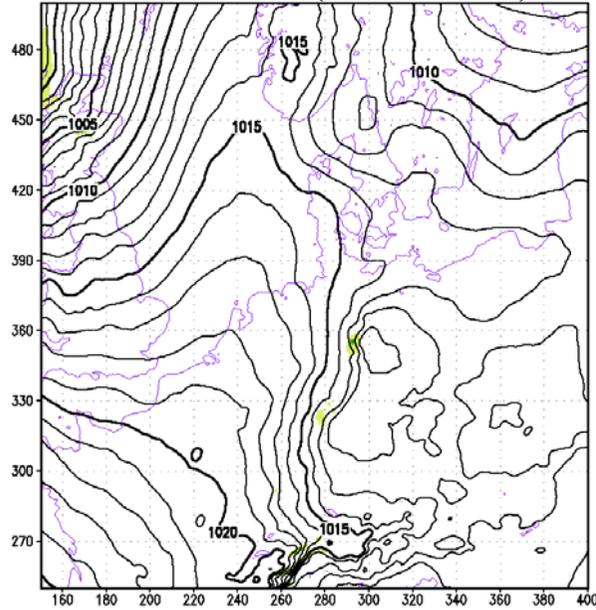
+ 60 h, REF (no 1DVAR)



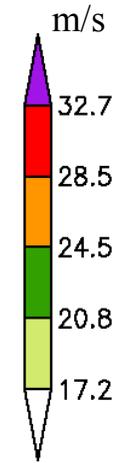
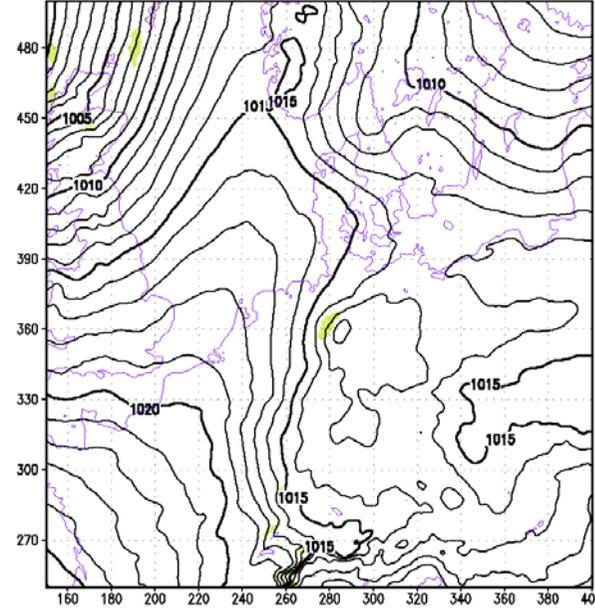
analysis



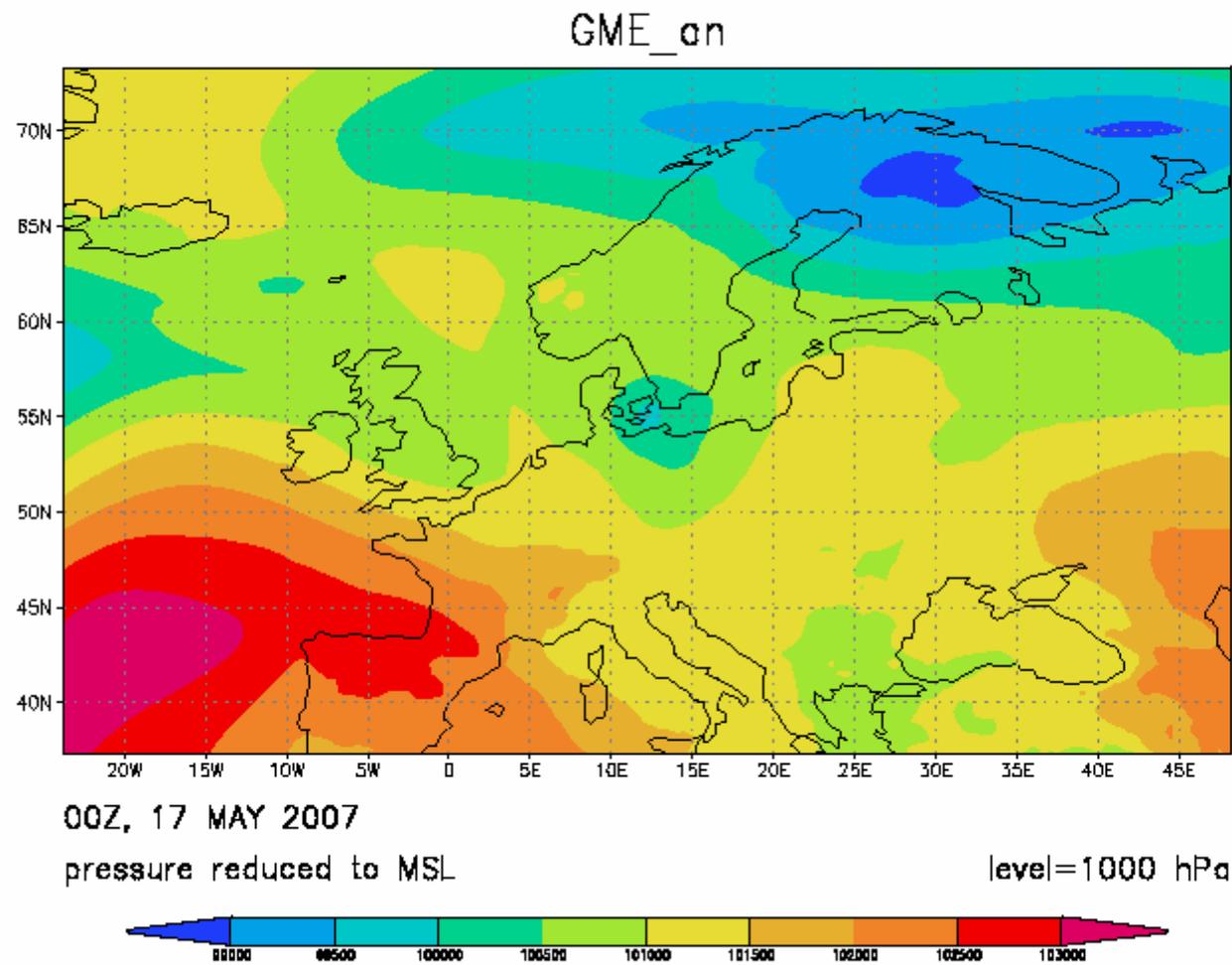
+ 48 h, REF (no 1DVAR)



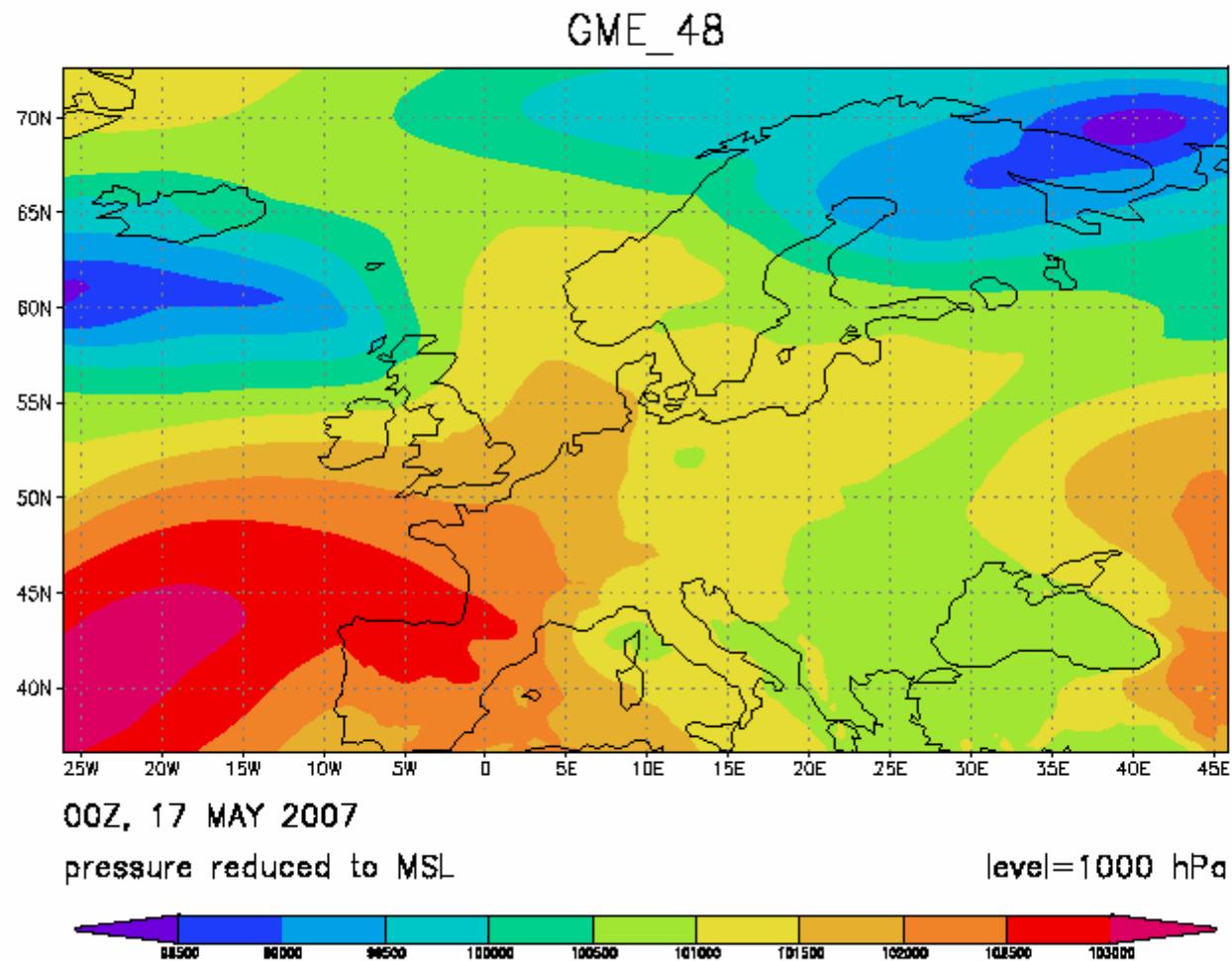
+ 60 h, 1DVAR-THIN2



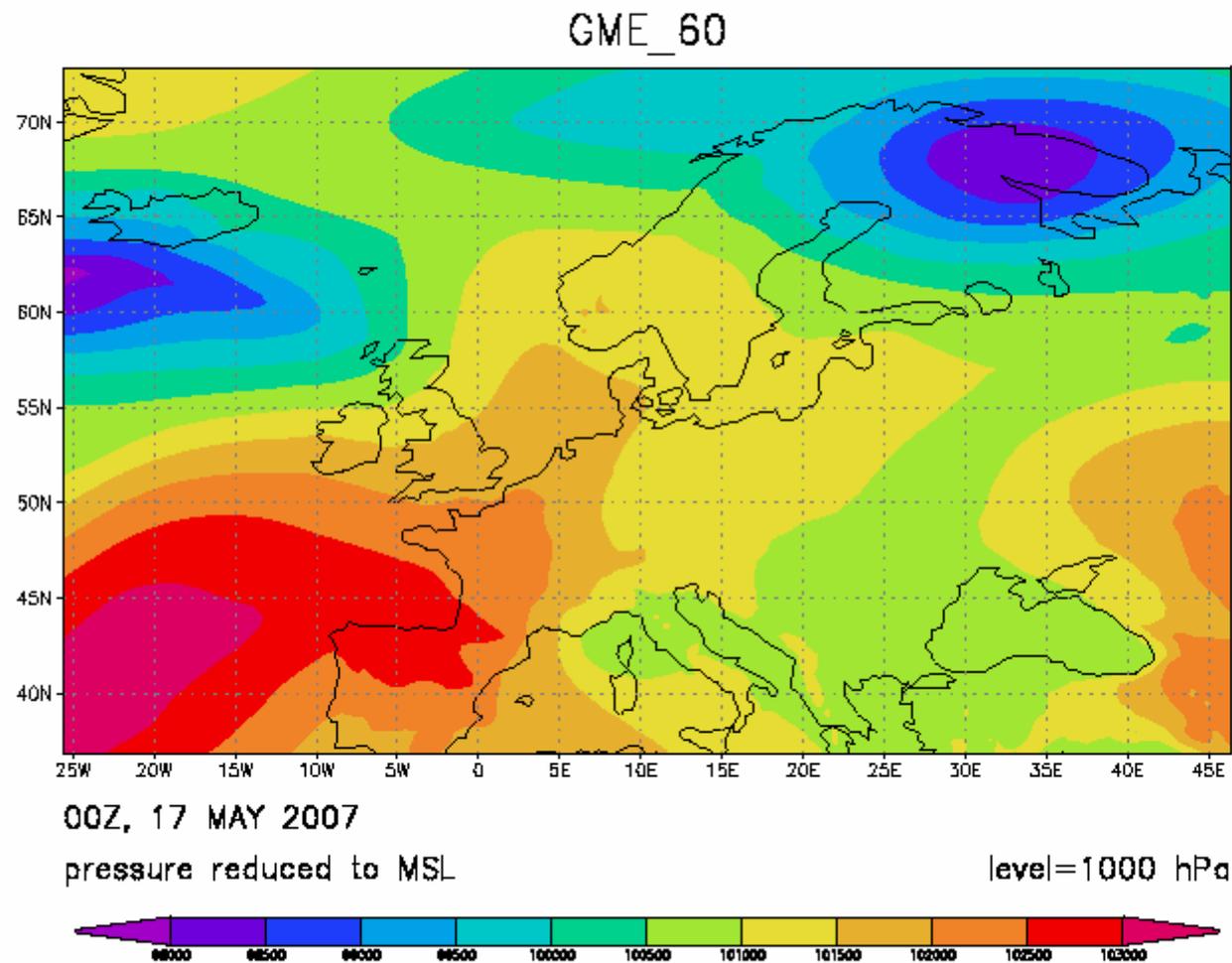
## GME analysis



## GME forecast for 48 hours

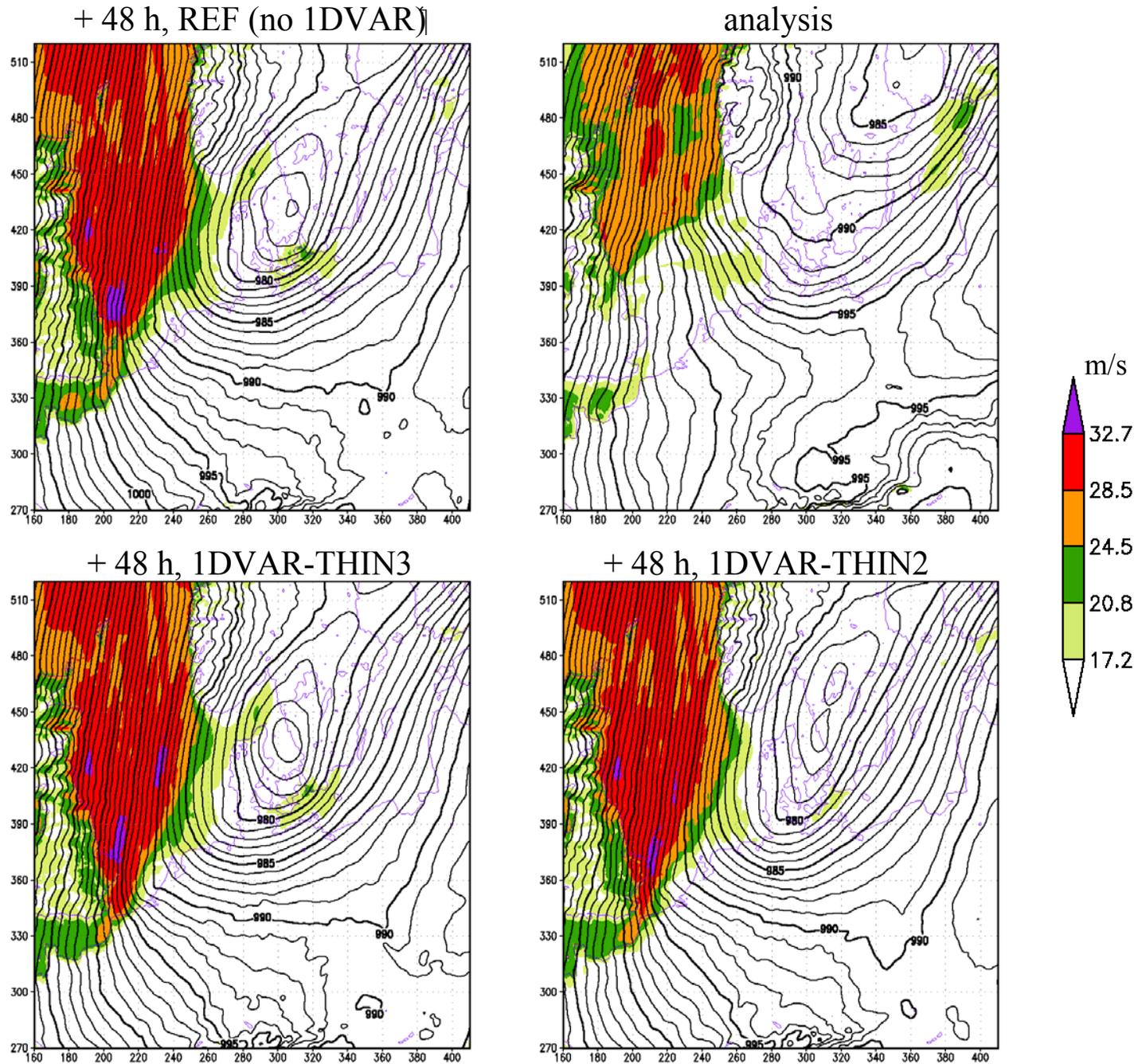


## GME forecast for 60 hours



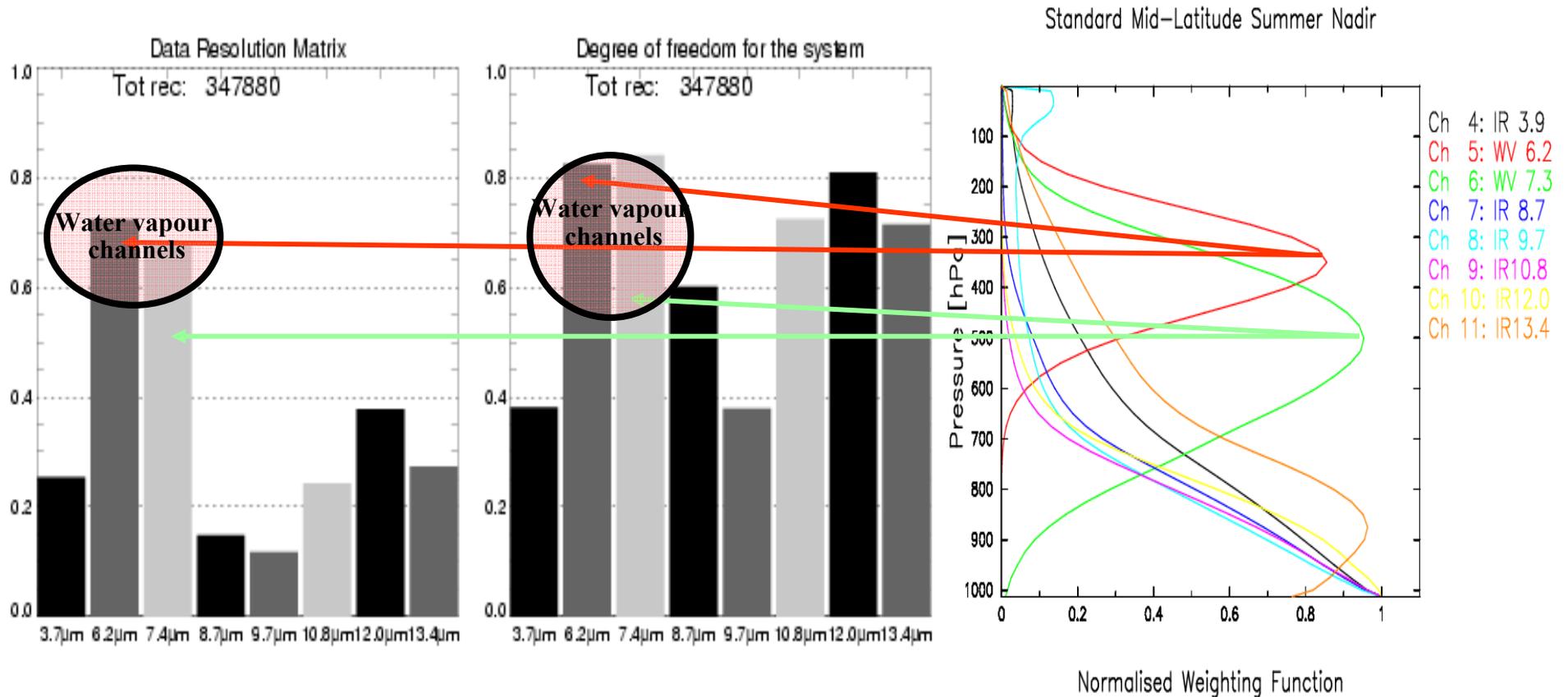
mean sea level pressure & max. 10-m wind gusts

valid for 20 March 2007, 0 UTC



# Assimilation of SEVIRI/MSG

DRM e DFS: period 1st -21st of September 2006

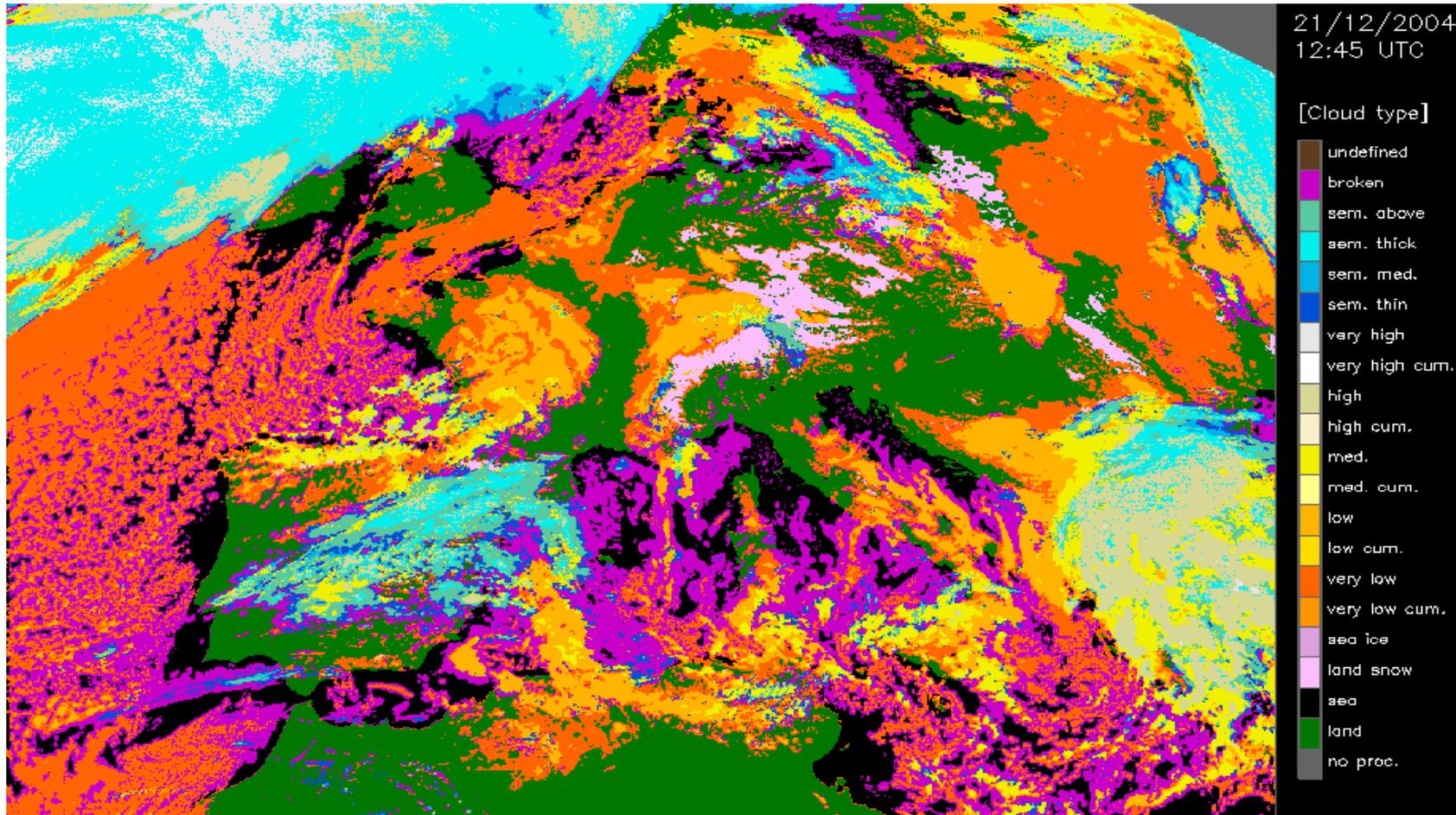


*DRM tells the most useful channels in between the ones used*

*DFS instead is a measure of how much a channel in isolation is able to reduce the background error*

# Cloud Clearing for MSG

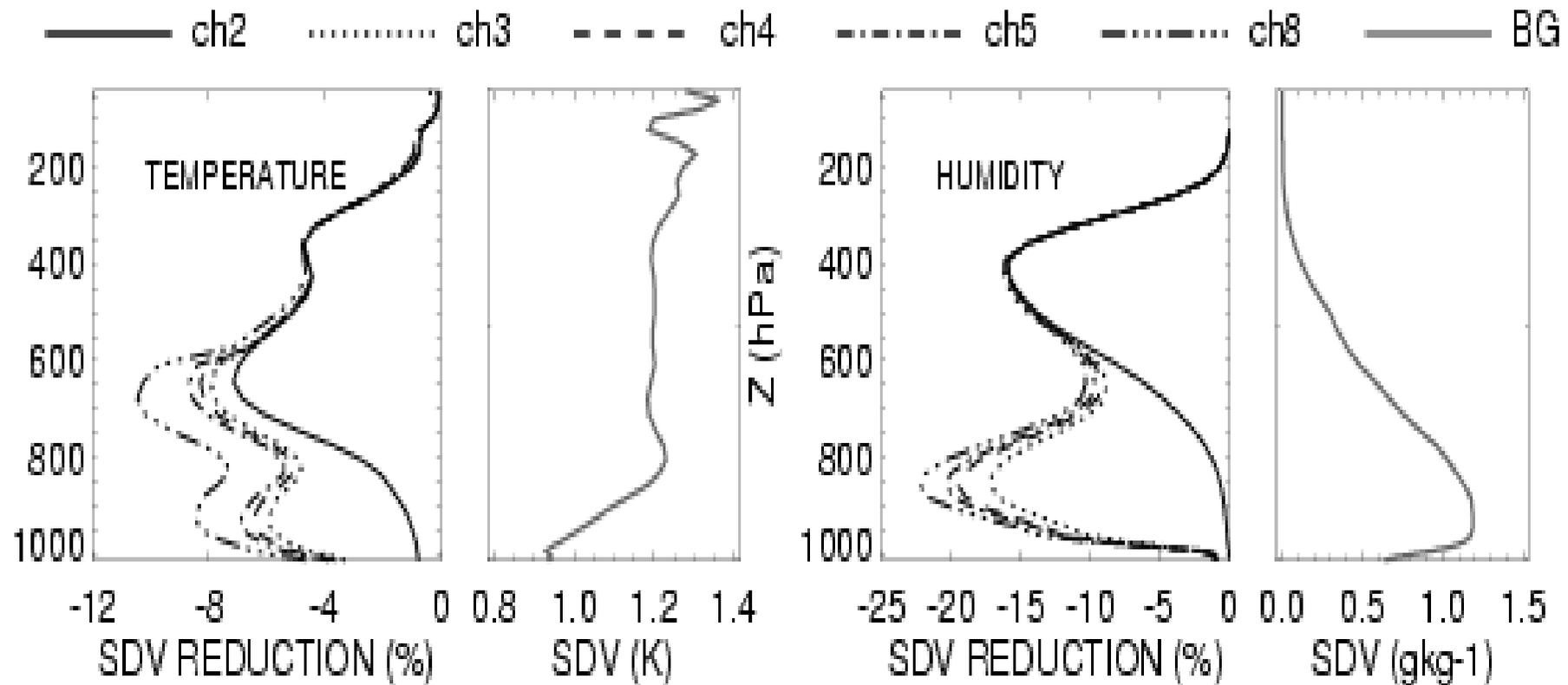
**MSG:** Cloud clearing using SAF-NWC software for MSG1 and MSG2



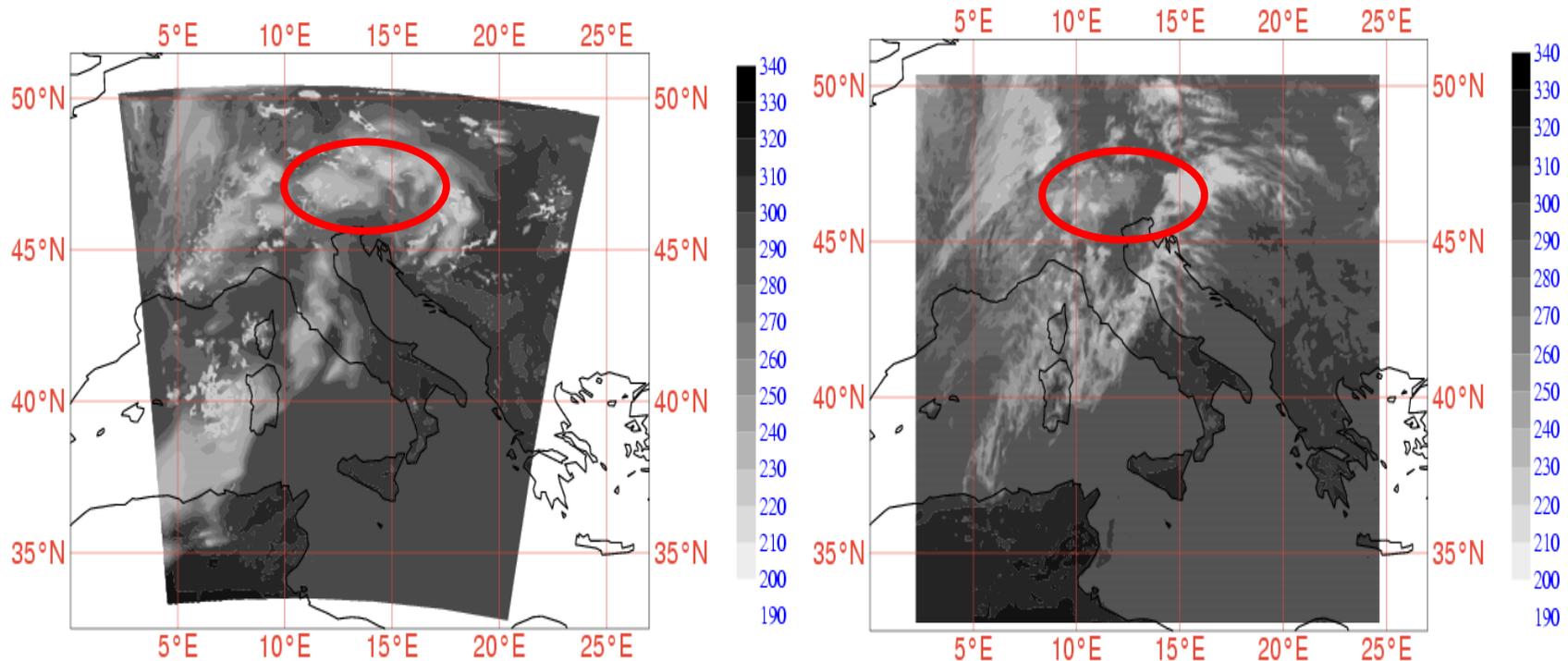
*Cloud Type from SAF-NWC-PGE02*

# Analysis error reduction (Rogers) period 1st -21st of September 2006

Set 2		6.2	7.3					
Set 3		6.2	7.3			10.8		
Set 4		6.2	7.3			10.8	12.0	
Set 5		6.2	7.3	8.7		10.8	12.0	
Set 8		6.2	7.3	8.7	9.7	10.8	12.0	13.4

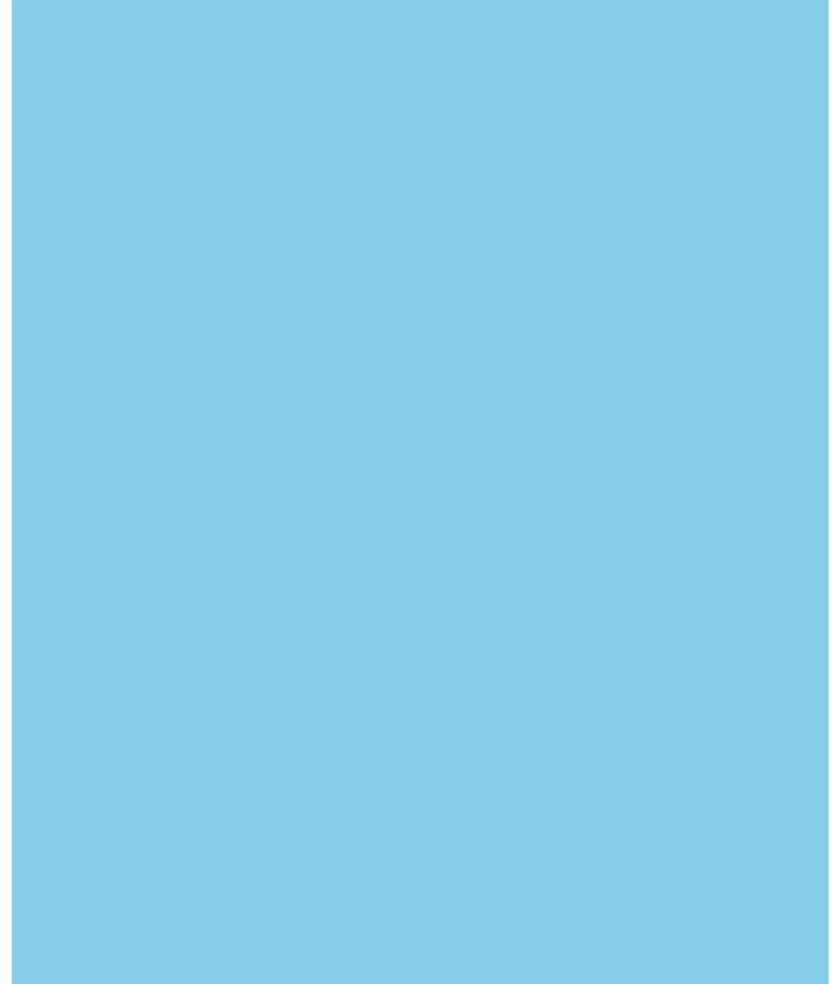
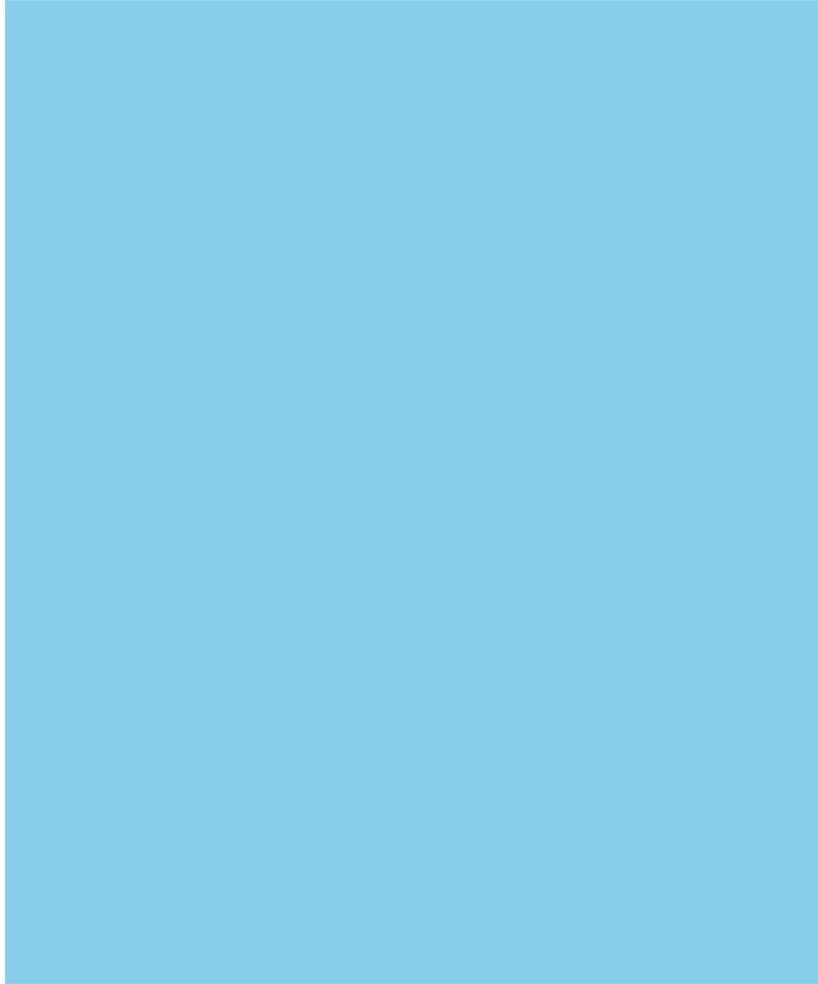


# FIRST CASE STUDY: 8 July 2004



The selected case study is a **false alarm** occurred in North North-Eastern Italy, Trentino Alto Adige and Friuli-Venezia-Giulia, on the 8<sup>th</sup> of July 2004. A risk scenario was diagnosed by LM outputs. In particular a large atmospheric instability and convection events were forecasted. In reality the event was of minor intensity and drier winds with associated scattered thunderstorms were recorded only on the early morning of the 9<sup>th</sup> July.

## CLOUD TABLE AND PROCESSING FLAG

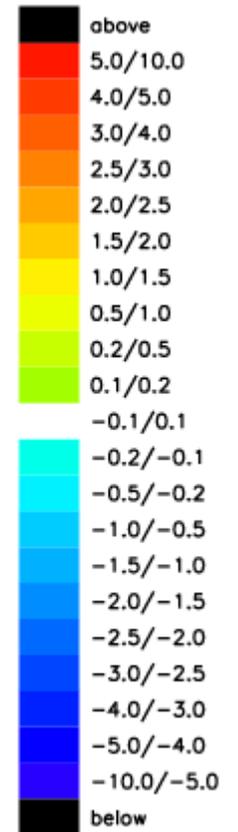


# INTEGRATED WATER VAPOUR AND INTEGRATED SATURATION WATER VAPOUR INCREMENTS

kg/m<sup>2</sup>

INT WV

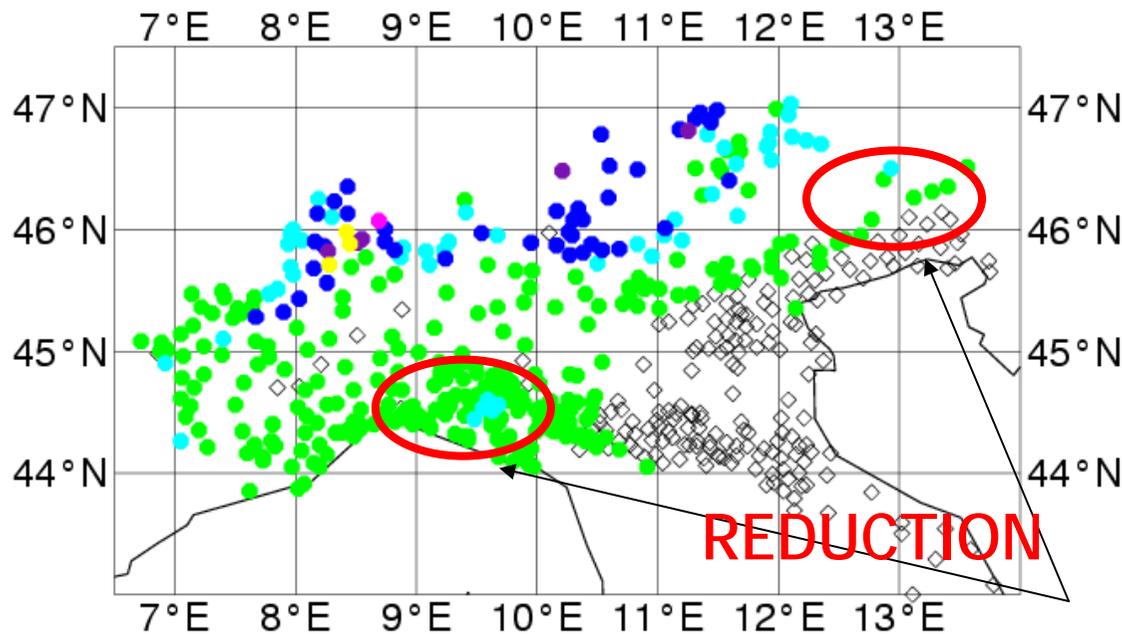
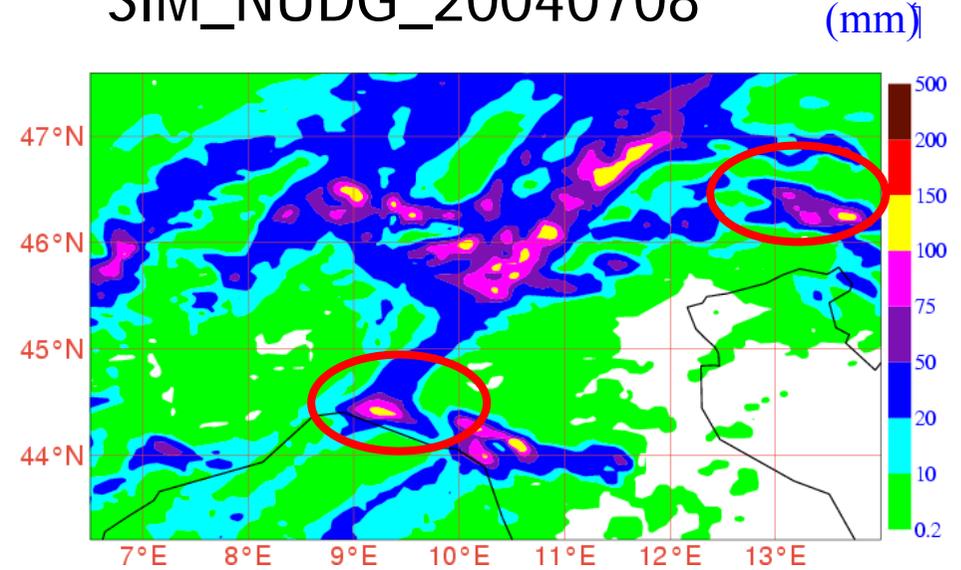
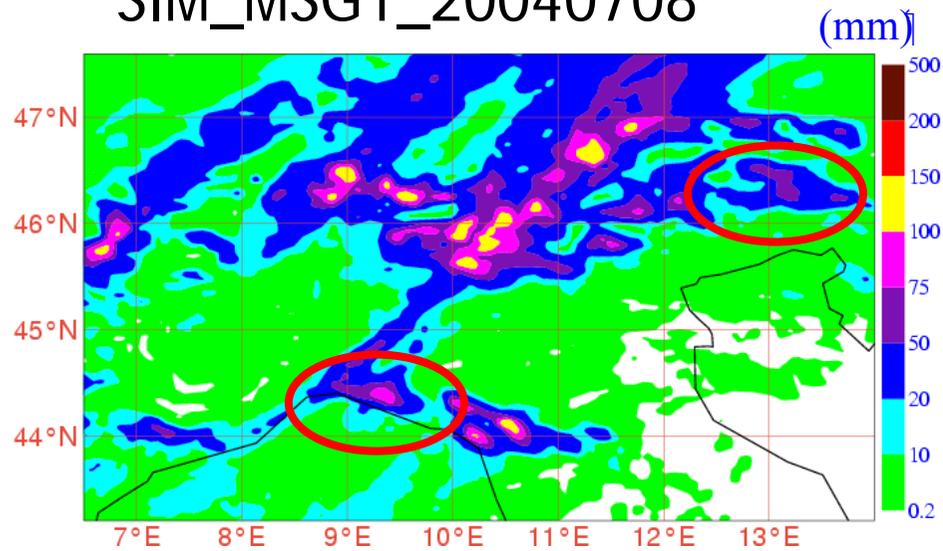
INT SAT WV (T)



# Precipitation forecast: 24 hrs integrated precipitation

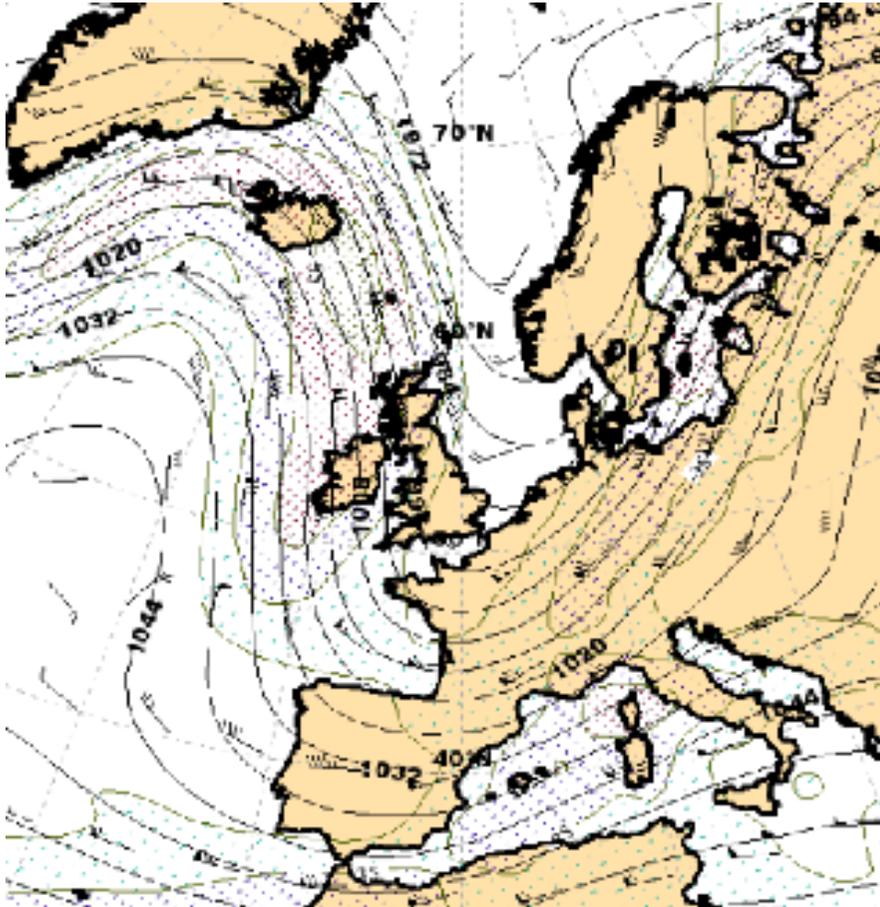
SIM\_MSG1\_20040708

SIM\_NUDG\_20040708

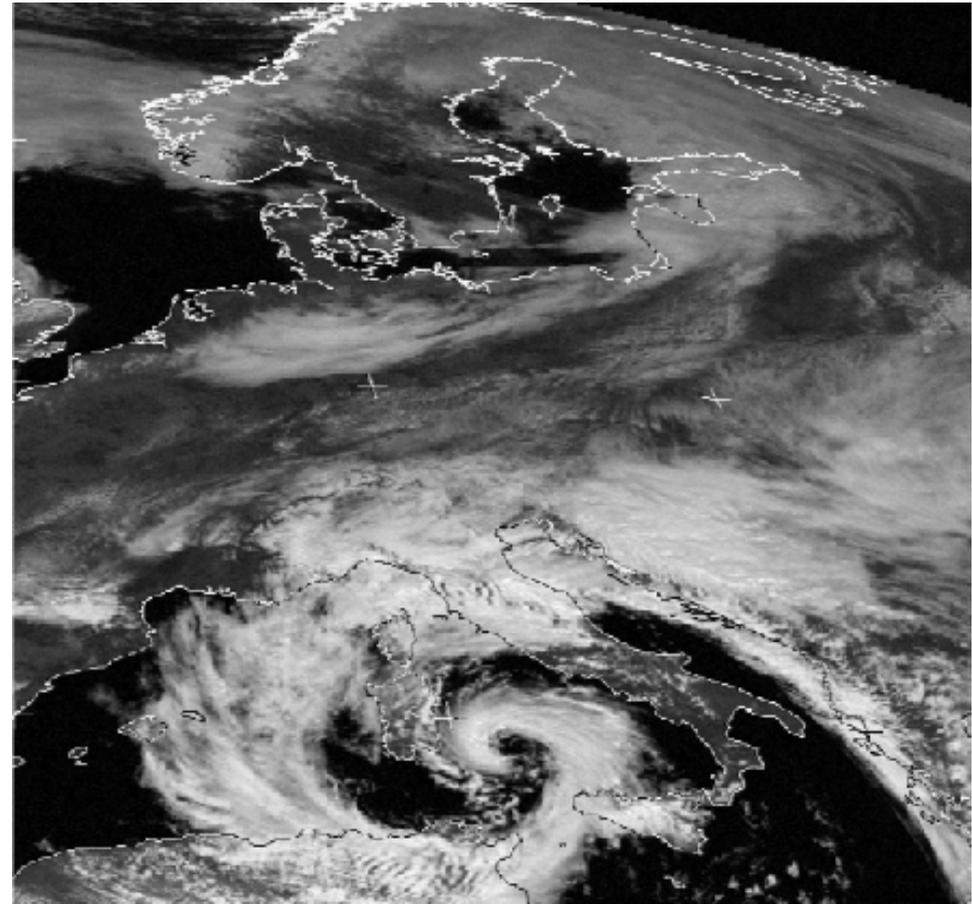


1. Both the forecasts do not correctly forecast the maximum of precipitation between 8 and 9 degrees of longitude and 46 degrees of latitude;
2. overestimation of the precipitation on Friuli-Venezia-Giulia region remains
3. the use of satellite data is able to reduce most of the overestimation of precipitation in the central Alpine area.

## 2nd case study : 9th of April 2005



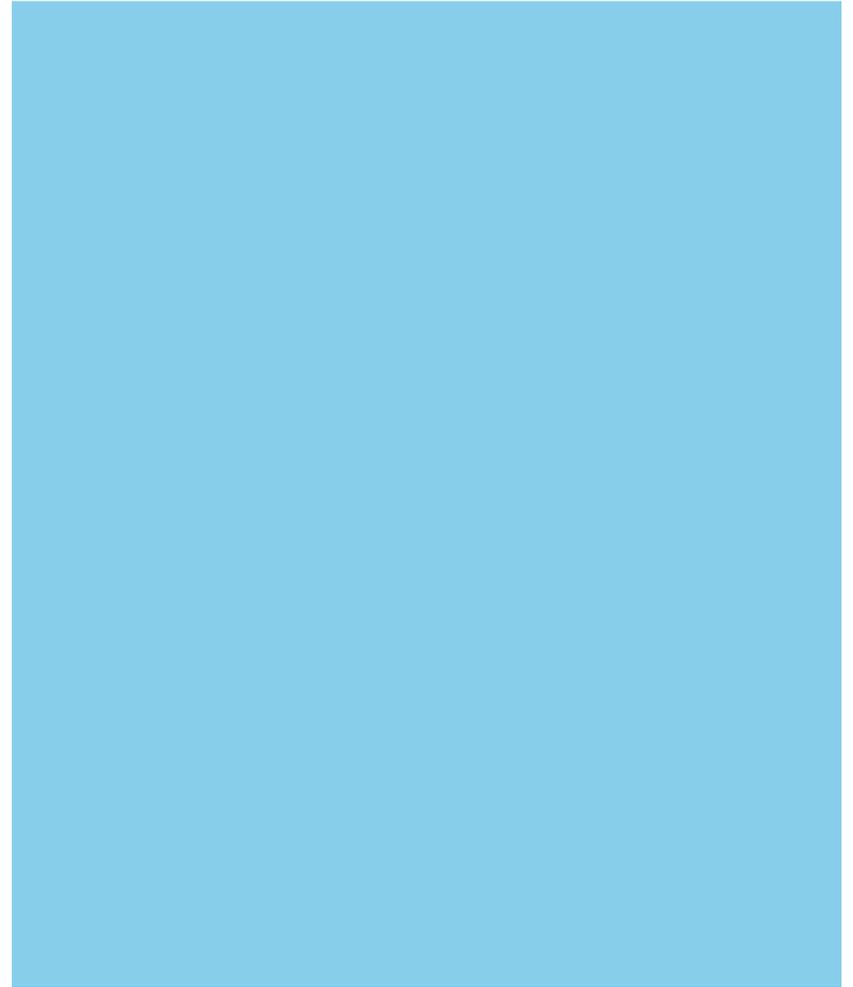
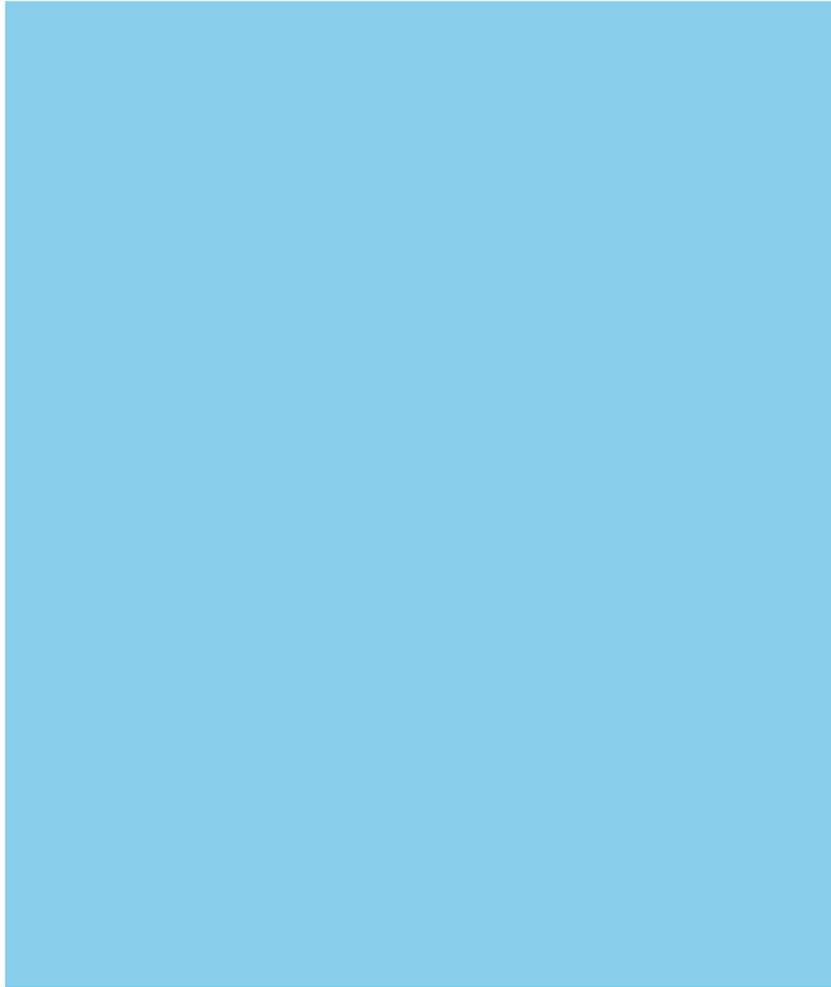
250 hPa, 12UTC 8<sup>th</sup>



METEOSAT 7, 12 UTC 11<sup>th</sup>

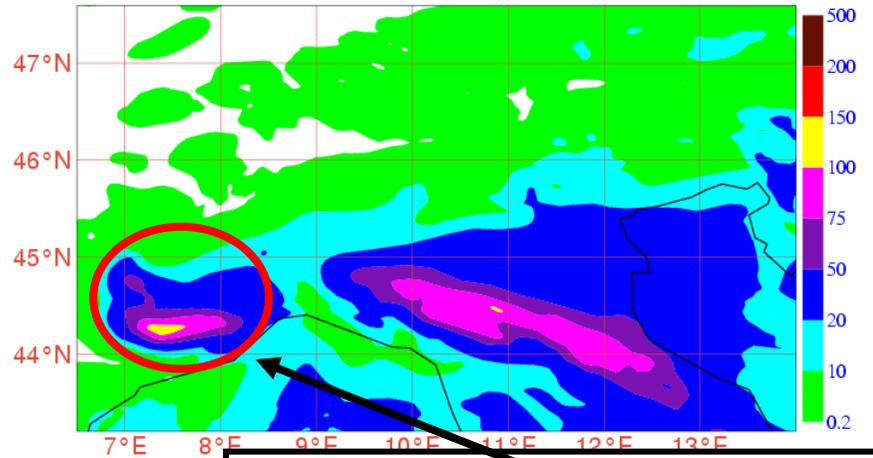
Forecast of heavy precipitation in the liguria region. Typically produced by southwesterly up-stream flow due to orographic forcing

## CLOUD TABLE AND PROCESSING FLAG

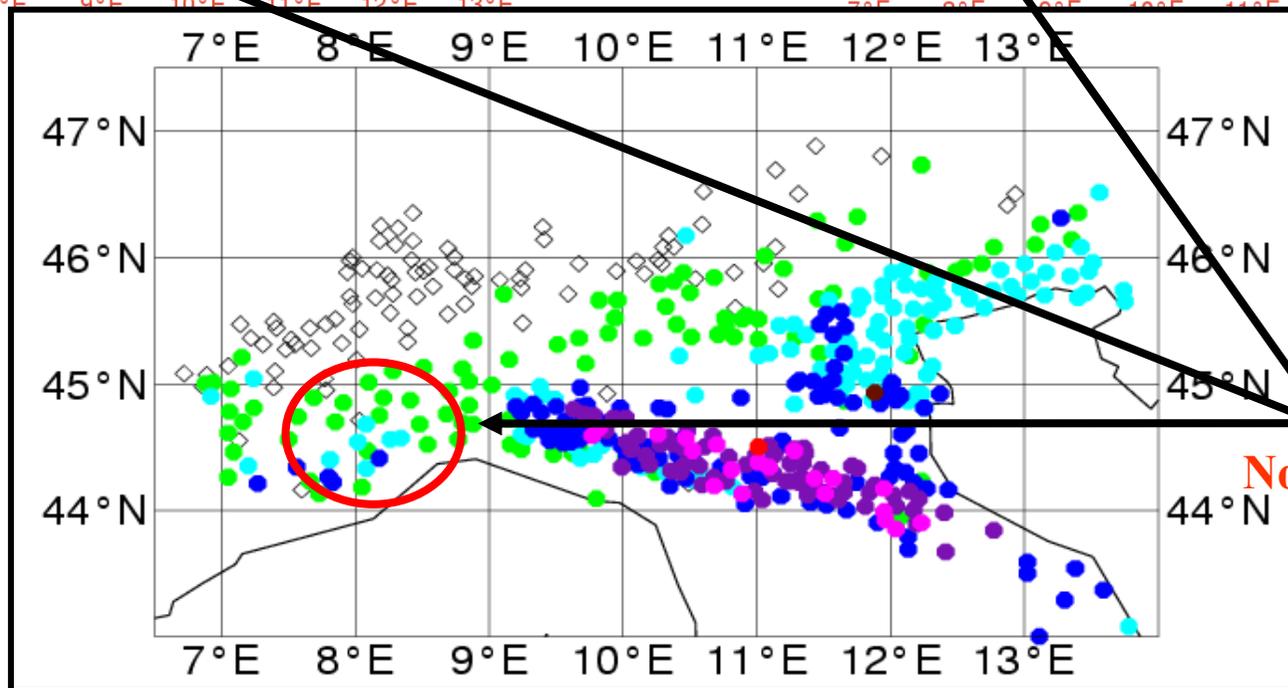
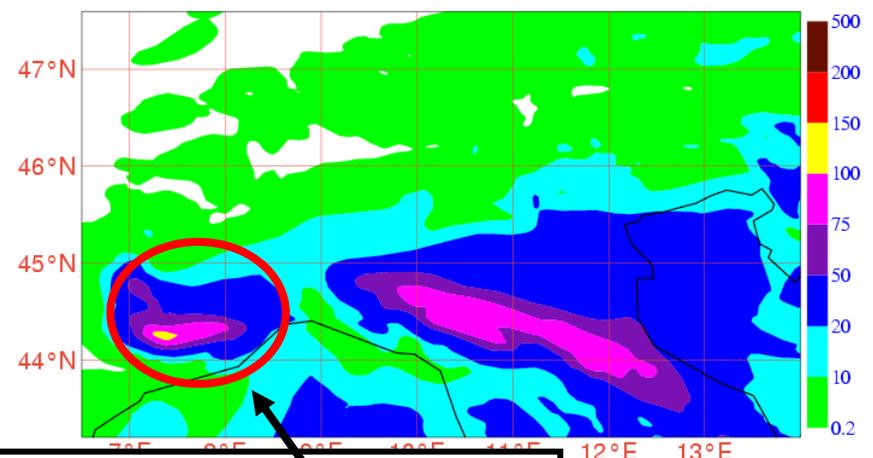


# Precipitation forecast

SIM\_MSG1\_20050409 (mm)



SIM\_NUDG\_20050409 (mm)



No improvement

## Conclusions for MSG:

- Wealthy information for temperature and humidity
- Use information of clouds and rain
  - to improve data coverage
  - to avoid spreading „dry“ (cloud free) increments into a „wet“ (cloudy) region

## Conclusions for AMSU:

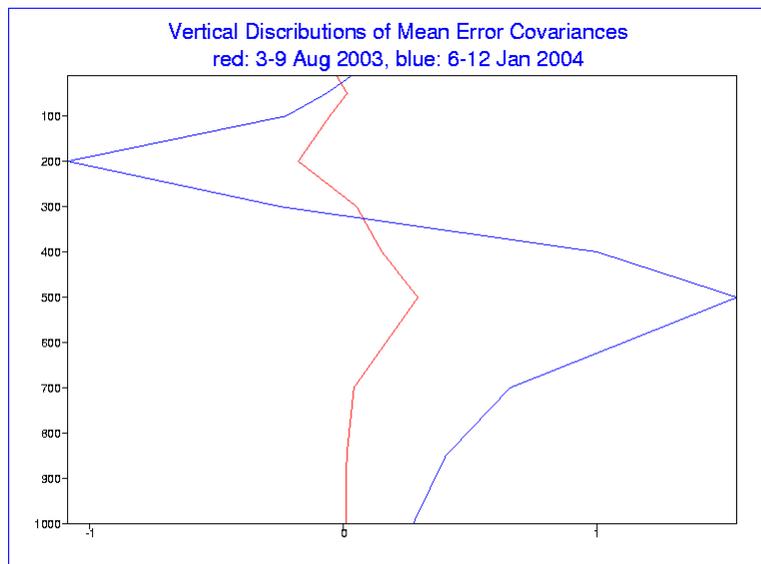
- similar approaches for limited area models, some complications: (stratosphere, bias correction, background errors)
- use information over land
- expect less impact on forecast quality than for global models

## Conclusions for the assimilation of satellite data in limited area models:

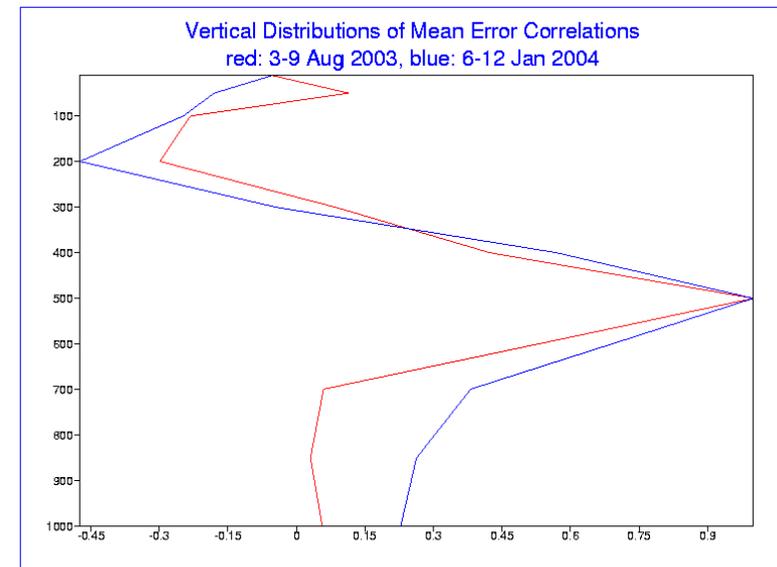
- with increasing resolution and more sophisticated physics  
more flexible background error structures are required  
(especially for sounders with high vertical resolution)
- variational methods: increasing interest for adaptive, situation or flow dependent  
background error covariances
- nudging: fit for high resolution observations  
tuning without mathematical framework required
- combine idea of nudging with variational developments  
(framework for nudging weights)
- interest for ensemble methods in order to prescribe adaptive background errors

## Background error covariance matrix $B$

vertical error structures derived from IFS  
blue: westerly winds, red: stable high pressure



covariances with 500hPa



correlations with 500hPa



Thank You for attending