

EWGLAM, October 2011



Met Office

Met Office data assimilation activities

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Contents

This presentation covers the following areas

- flow dependent covariances
- humidity control variable
- Doppler radial winds
- road sensor data
- symmetric cloud observation errors
- high resolution assimilation impact
- observation sensitivity study
- plans



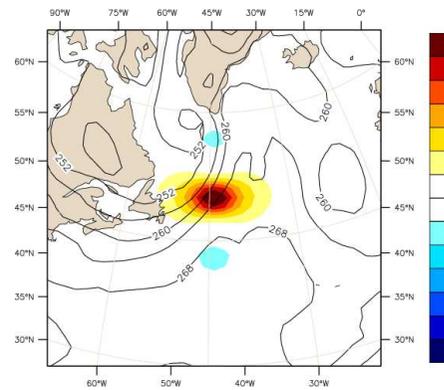
July 2011 Global Data Assimilation Upgrade Package (PS27)

- Assimilation method
 - Hybrid 4D-VAR algorithm.
 - Moisture control variable: Replacing RH with scaled humidity variable
- Observation changes
 - Introduce METARS
 - GOES/Msat-7 clear-sky radiances, extra IASI (land)
 - Revisions to MSG clear-sky processing and GPSRO
 - Reduced spatial thinning (ATOVS/SSMIS/IASI/AIRS/aircraft)

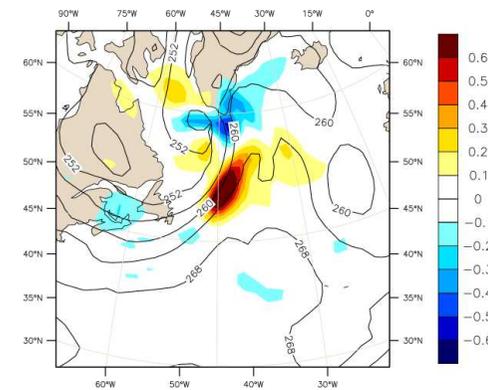
Hybrid data assimilation

- **Basic idea:** Use data from MOGREPS-G to improve the representation of background error covariances in global 4D-VAR:

u response to single u observation:

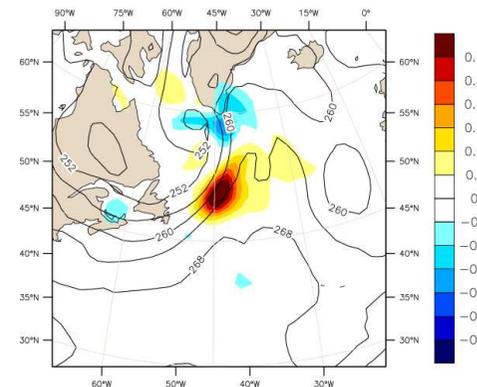


Climatological COV



MOGREPS COV

- MOGREPS is sensitive to the position of the front, and gives covariances that stretch the increment along the temperature contours.
- Ensemble currently too small to provide the full covariance, so we blend the MOGREPS covariances with the current climatological covariances; i.e., we use a **hybrid** system:



Hybrid COV

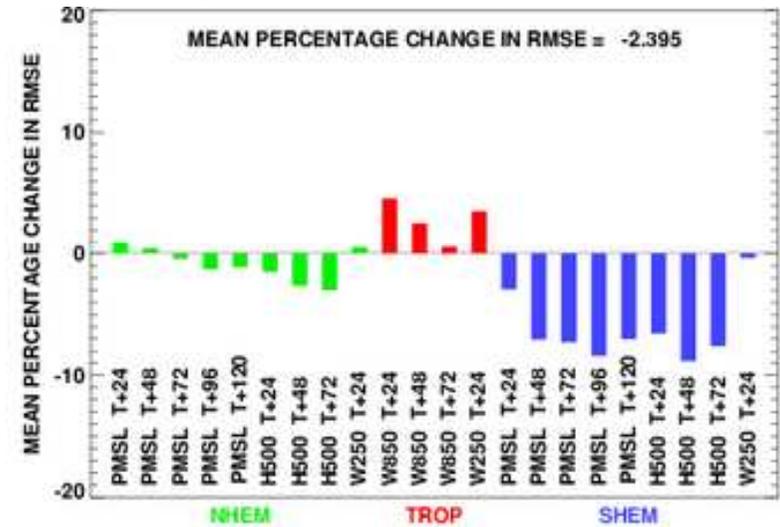
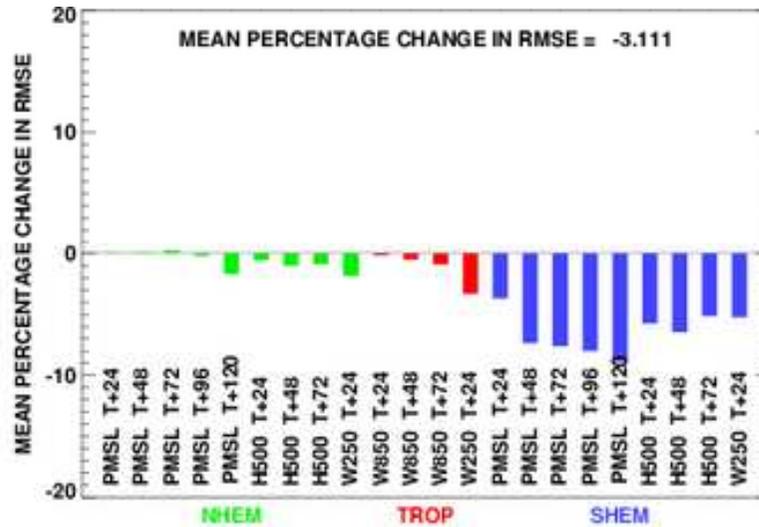


PS27 Global NWP Index Impact

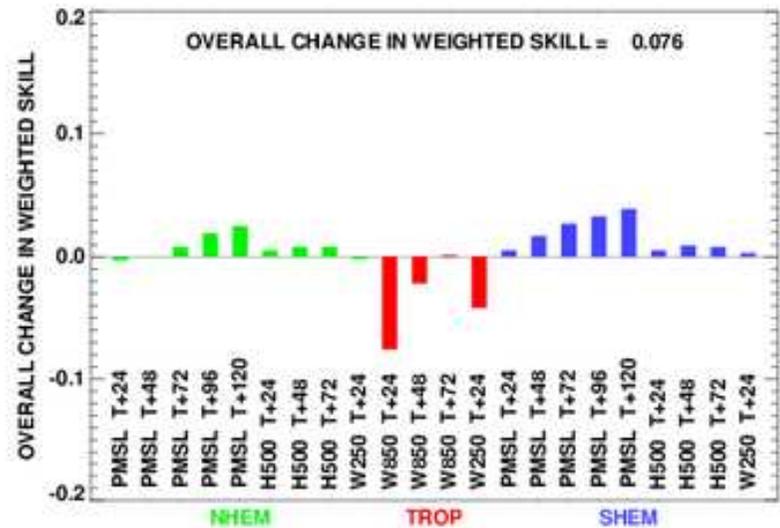
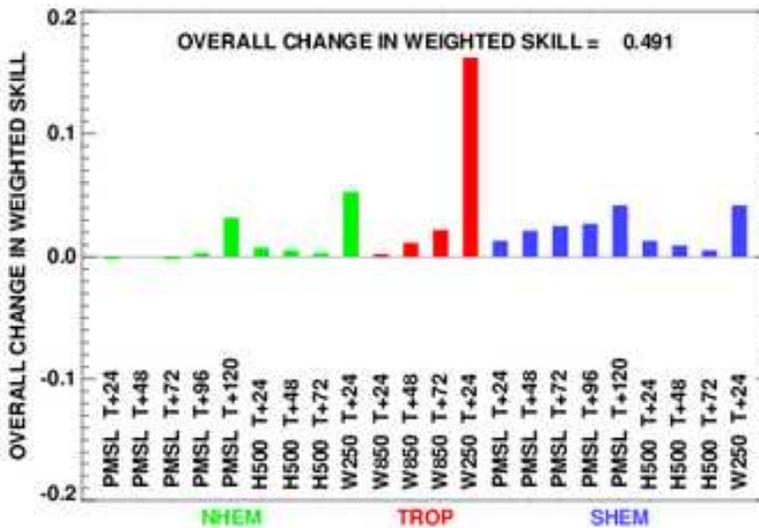
rms

Impact Vs. Observations=+2.11

Impact Vs. Analyses=+0.67



skill



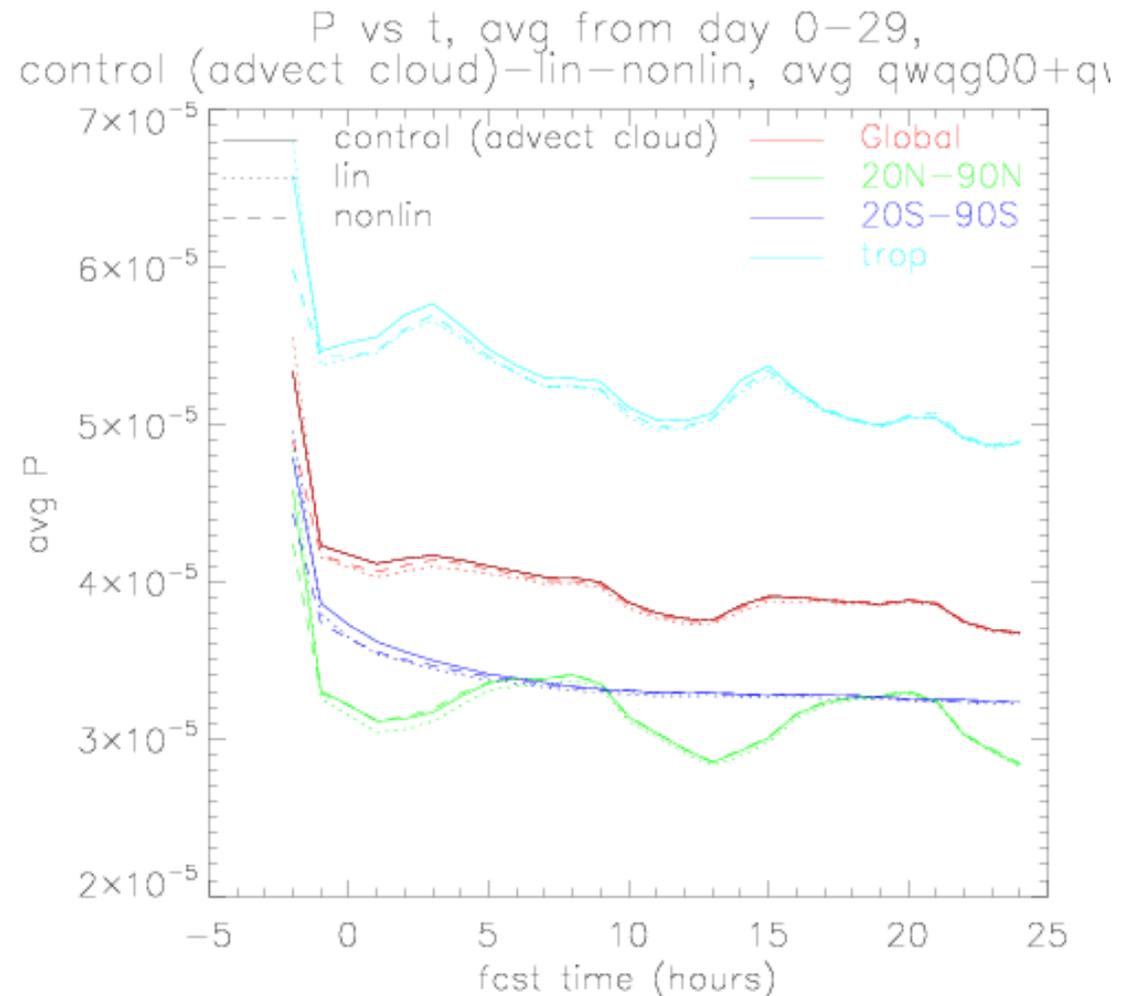
- Biggest single reduction in overall global forecast error for many years.
- First time in memory that all parameters have reduced error vs obs. (usually a mix).

New Humidity Transform

- $\mu = (q_T' - hcT') a / q_{\text{sat}}(\mathbf{b})$
- q_T' increment of total q – including cloud
- T' temperature increment
- $h = h(\text{RH}\mathbf{b})$ gives “balanced” q_T increment from T'
- $a = a(\text{RH}\mathbf{a}, \text{RH}\mathbf{b})$ is normalising factor so that $\sigma(\mu) \approx 1$ this reduces under/overshoots
 - h and a are derived from training data
- If $a = a(\text{RH}\mathbf{b})$ then we have linear transform

Precipitation spin-down

- Excessive ppn over first hour (esp. over oceans) then slower decline
- Nonlinear trial (dashed) reduces jump by ~40%

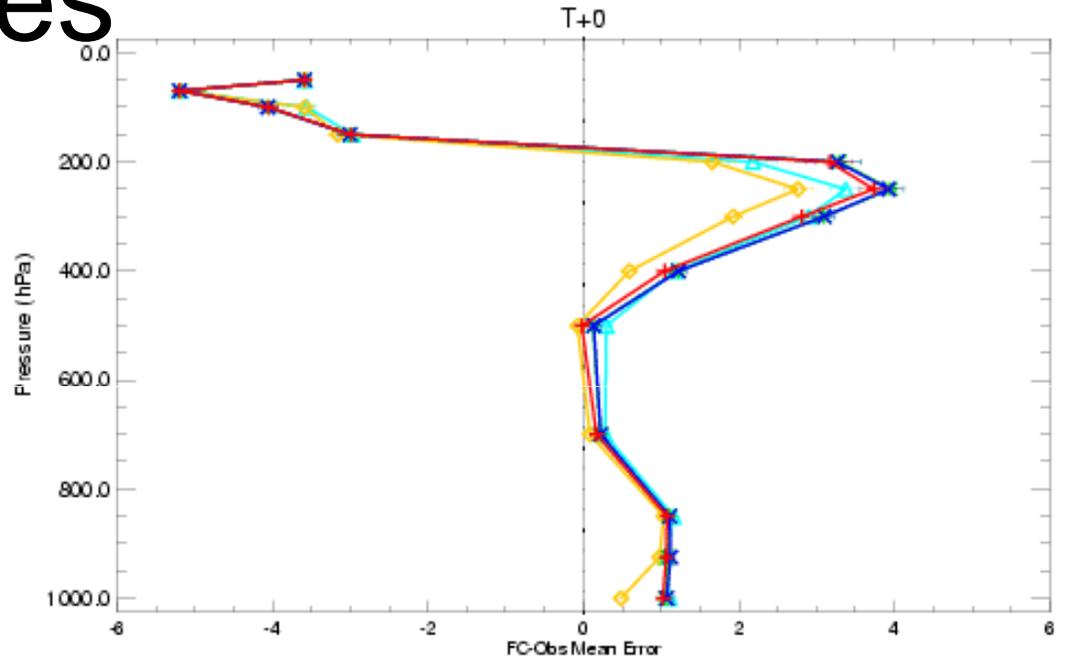




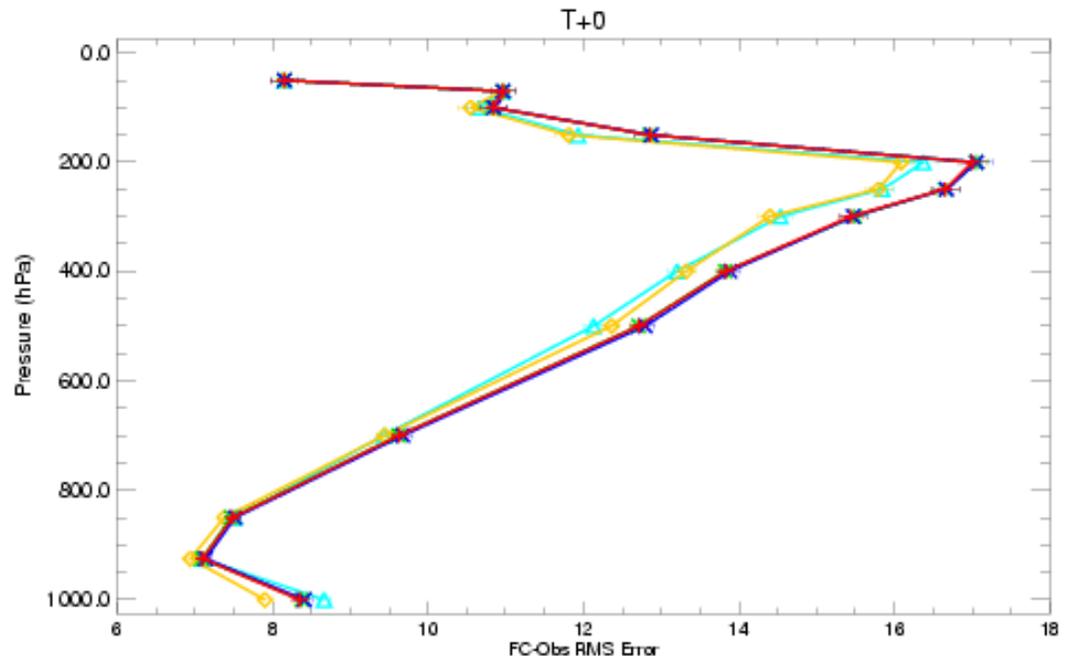
RH v Sondes 20-90°N

Cases: + PS25 BI control x Uncap RH and extraRH for SURF * qcl and qcf in LS
◇ linear moisture control variable △ nonlinear moisture control variable

- T+0 mean fit:



- T+0 rms fit:

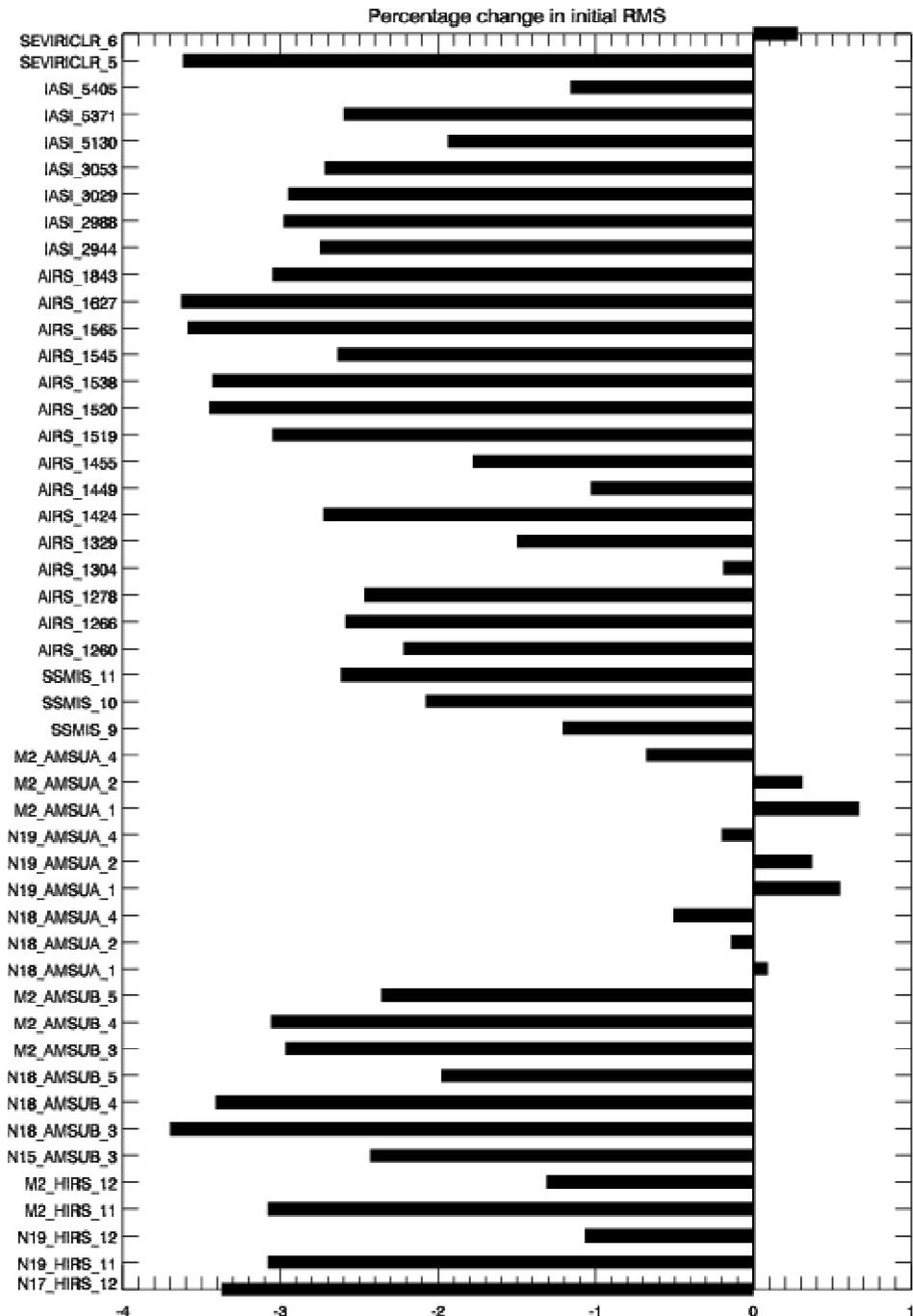


- largest diffs in upper-trop
– better fit for trials there



% change in fit to humidity channels

- Fit to upper tropospheric humidity-sensitive channels improved by ~ 3% (large)
- Less improvement for channels that “see” the surface.





Adaptive Mesh Transform

(Piccolo & Cullen, 2011: Q. J. R. Met. Soc., 137, 631-640)

- ❖ aims to change the vertical background-error correlations by moving the vertical levels to concentrate mesh points around temperature inversions.
- ❖ movement of the levels is guided by a scalar ***monitor function***, chosen to be a function of the *static stability* which strongly controls vertical mixing in the atmosphere and thus probably the vertical correlation structure of model variables.
- ❖ introduce the adaptive method within the Met Office VAR system as an extra transformation in the sequence of variable transformations used to simplify the background term of the cost function:

$$\delta\mathbf{x} = \mathbf{U}\boldsymbol{\chi} = \mathbf{U}_p \mathbf{U}_a \mathbf{U}_v \mathbf{U}_h \boldsymbol{\chi} \quad \text{and} \quad \mathbf{B} = \mathbf{U}\mathbf{U}^T$$

where \mathbf{U}_a is the "adaptive mesh transform" which is placed between the parameter transform \mathbf{U}_p and the vertical transform \mathbf{U}_v

Adaptive Grid Formulation

The first step of the \mathbf{U}_a transform is to calculate a *monitor function* $M (>0)$ in physical space $z \in [0,1]$:

$$\int_0^1 M(z') dz' = 1$$

The second step is to generate the adaptive mesh in physical space by defining a computational coordinate $\zeta \in [0,1]$:

$$\zeta(z) = \int_0^z M(z') dz'$$

The map from computational domain to physical domain is thus defined by the a unique one-dimensional map which connects intervals of a prescribed length.

Finally, the control variables χ which will be generated at points ζ by the vertical transform are then interpolated to the true levels z .



Choice of the Monitor Function

$$M = \sqrt{1 + c^2 (\partial \theta / \partial z)^2}$$

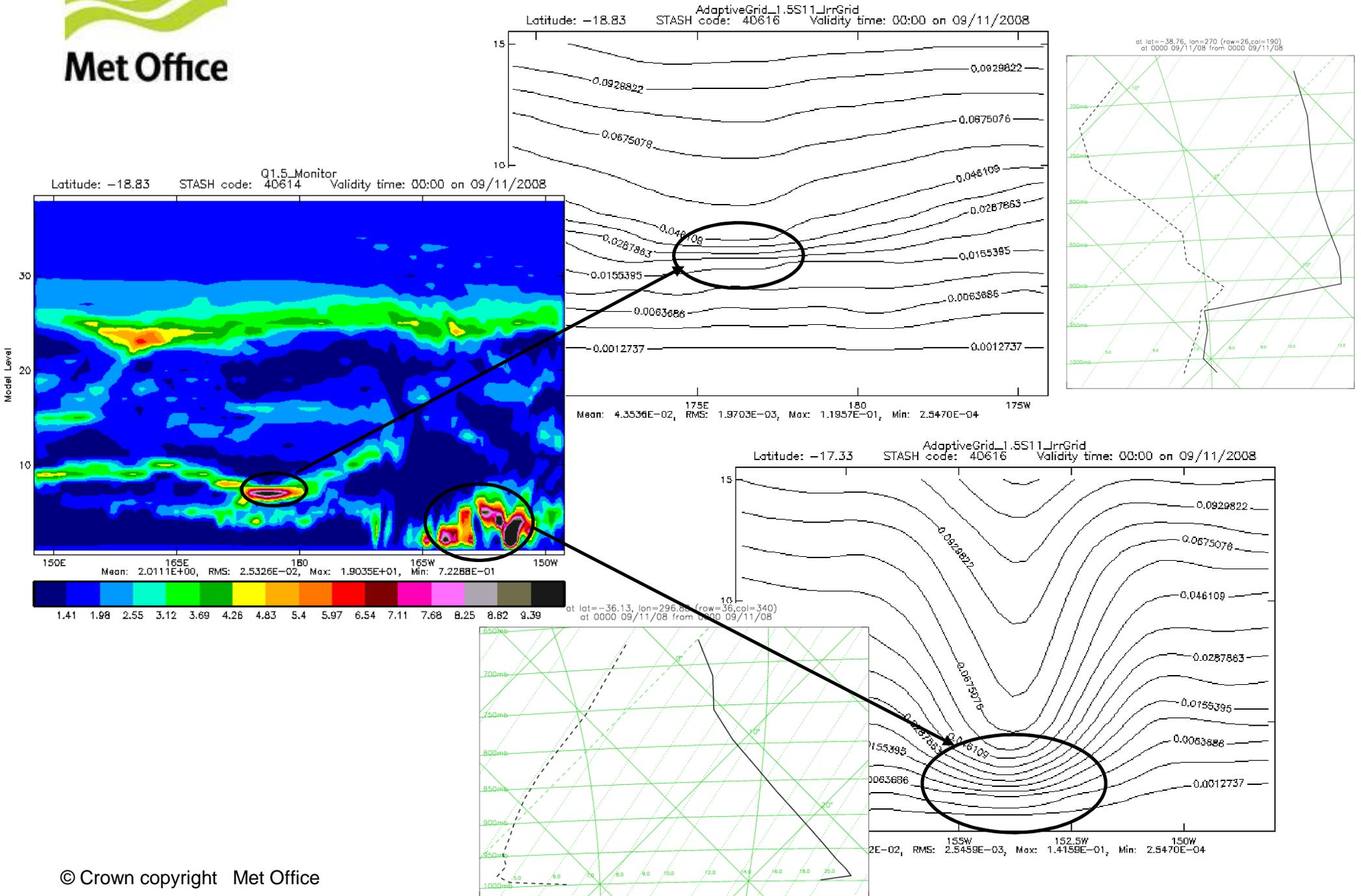
$M > 0$ and can be modulated by a scaling factor c .

If $c = 0$, the computational grid and the physical grid are the same.

Since mesh points will be clustered where the monitor function is large, this choice of M will cluster mesh points in regions of large static stability.



Monitor Function and Adaptive Grid

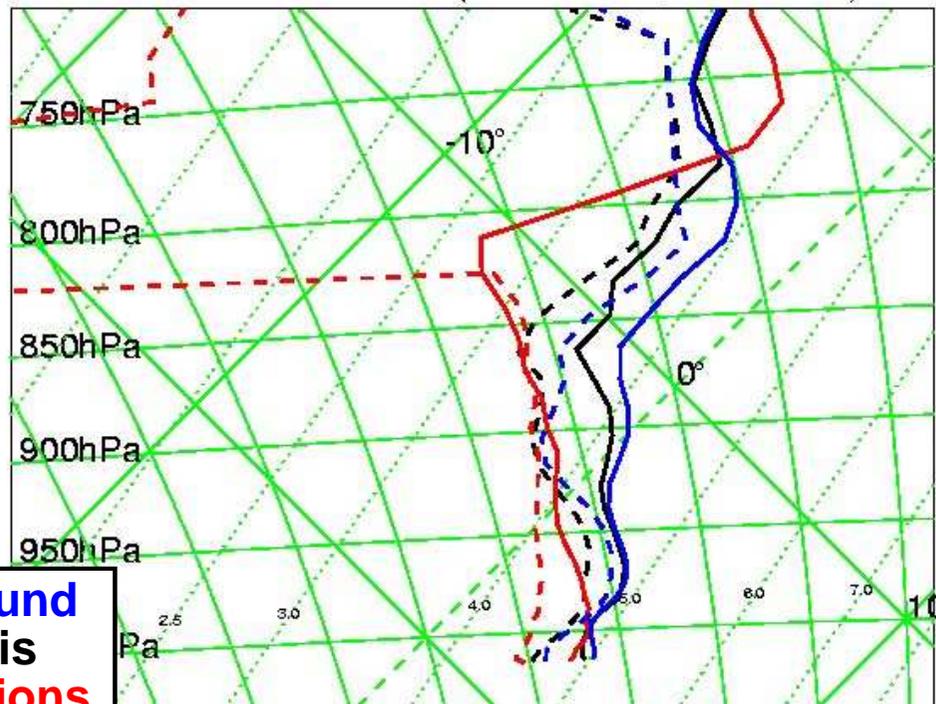




3 Jan 2011 00z: Camborne

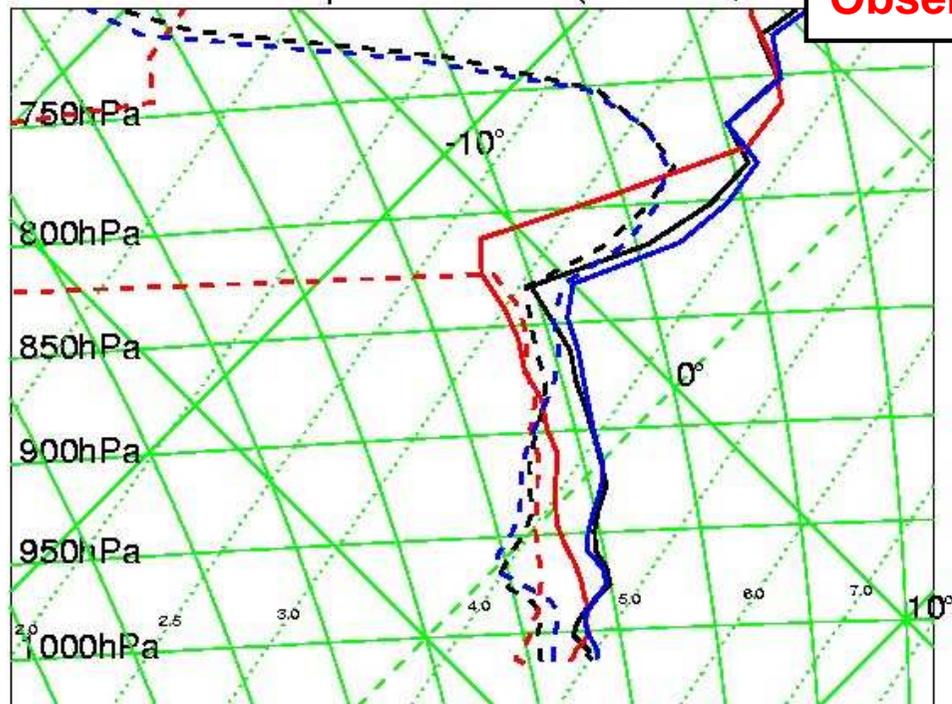
The monitor function is based on the background state: if the inversion is not present, the vertical grid does not change.

Control Camborne (lat=50.22,lon=-5.33)



Background
Analysis
Observations

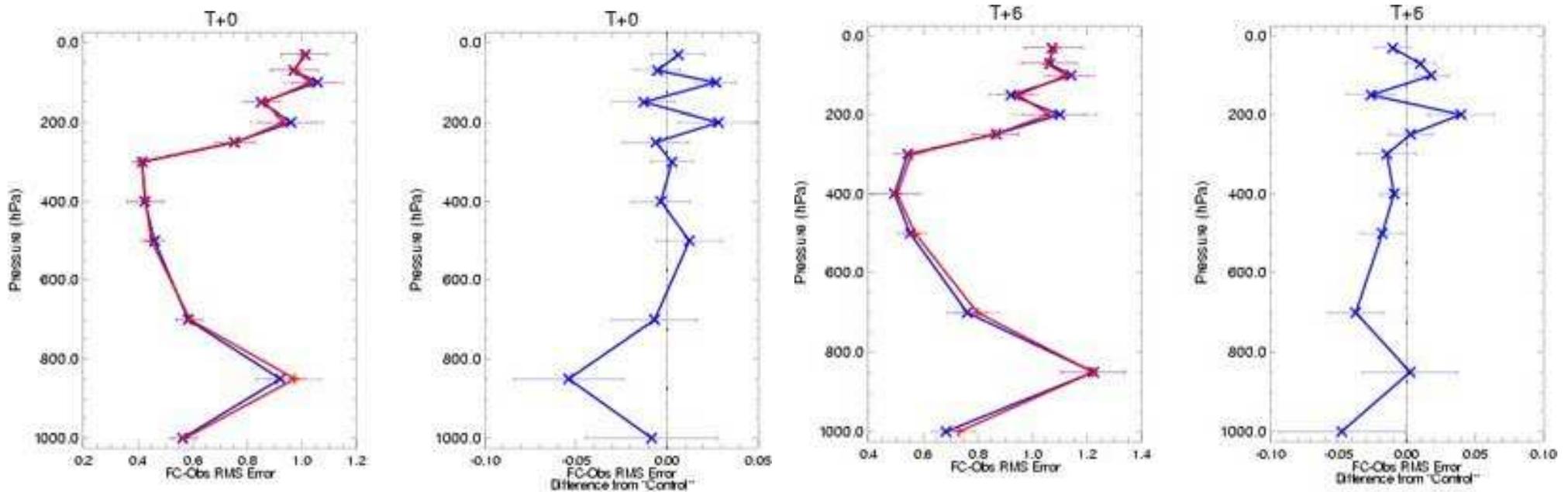
AG Theta 0.75 LSplus Camborne (lat=50.22,lon=-5.33)



When the monitor function is based on an updated background-state using the observation's information in the minimisation process, the analysis has a clearer inversion.

Analysis vs Sonde Observations

Temperature RMS error over the winter period



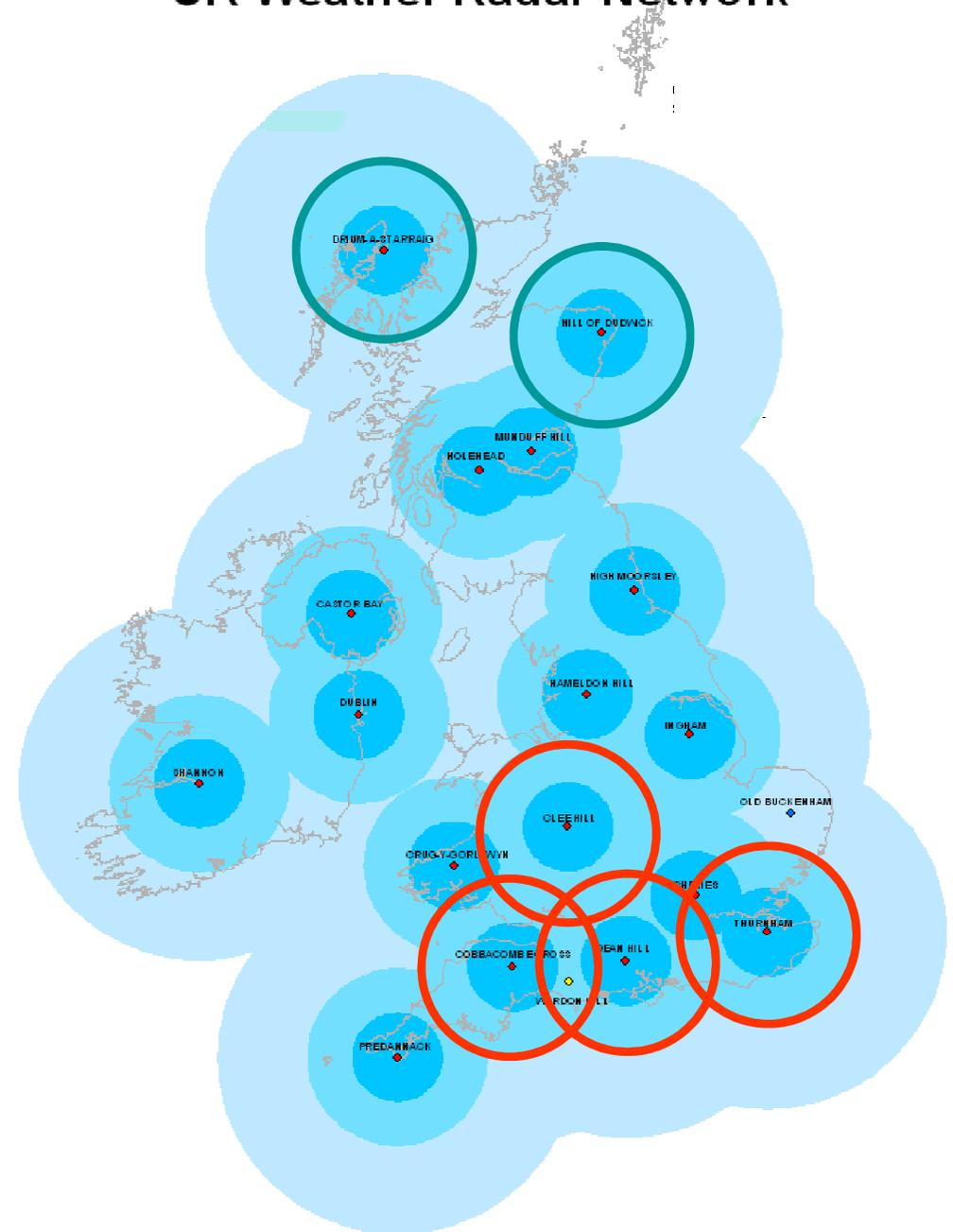
Results from the full coupled analysis/forecast system:

- small improvement of temperature RMS error versus sonde profiles in the lower atmosphere for both winter and summer cases up to T+ 6h
- (also slight improvement for cloud base height and T2m)



Doppler radial winds

UK Weather Radar Network



6 radars currently providing radial winds

(plans to upgrade whole network by 2013)

4 assimilated operationally so far

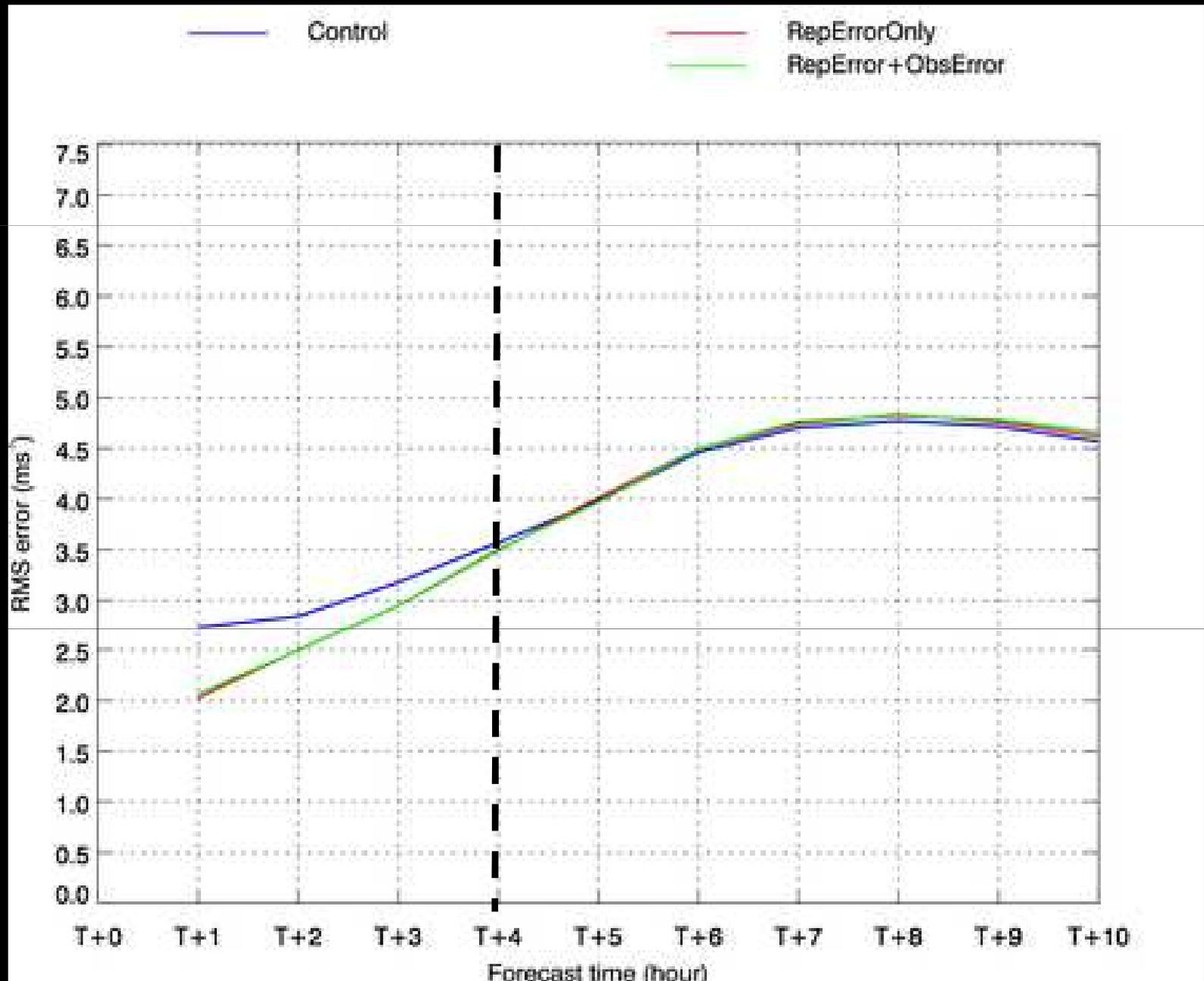
100 km radius

elevations between 1° and 9°

1° azimuthal 600 m radial

every 5 minutes

RMSE against Doppler Wind



David
Simonin



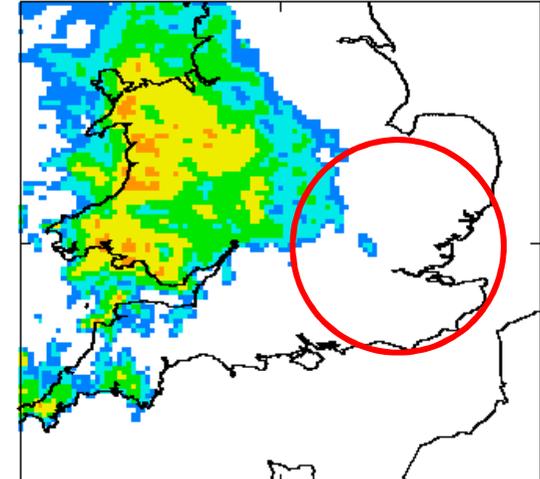
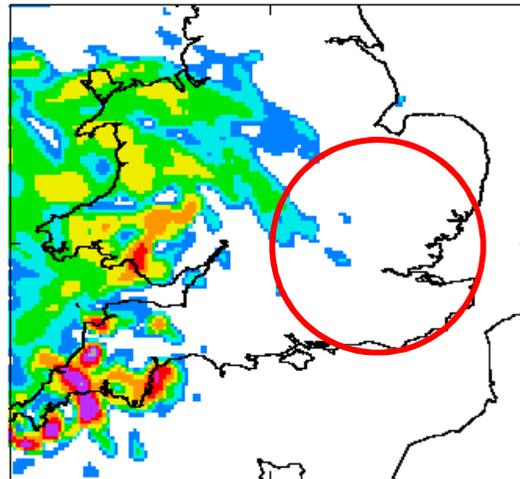
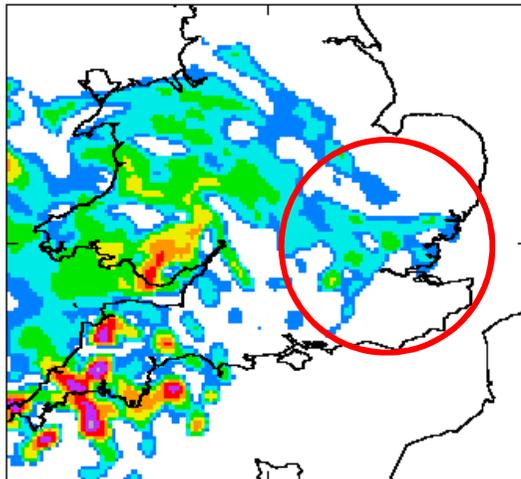
Individual case where rainfall location is seen to be improved

T+4

Control

Trial

Radar



0.2 0.5 1.0 2.0 4.0 8.0 16.0 32.0
mm/hr

Helen Buttery



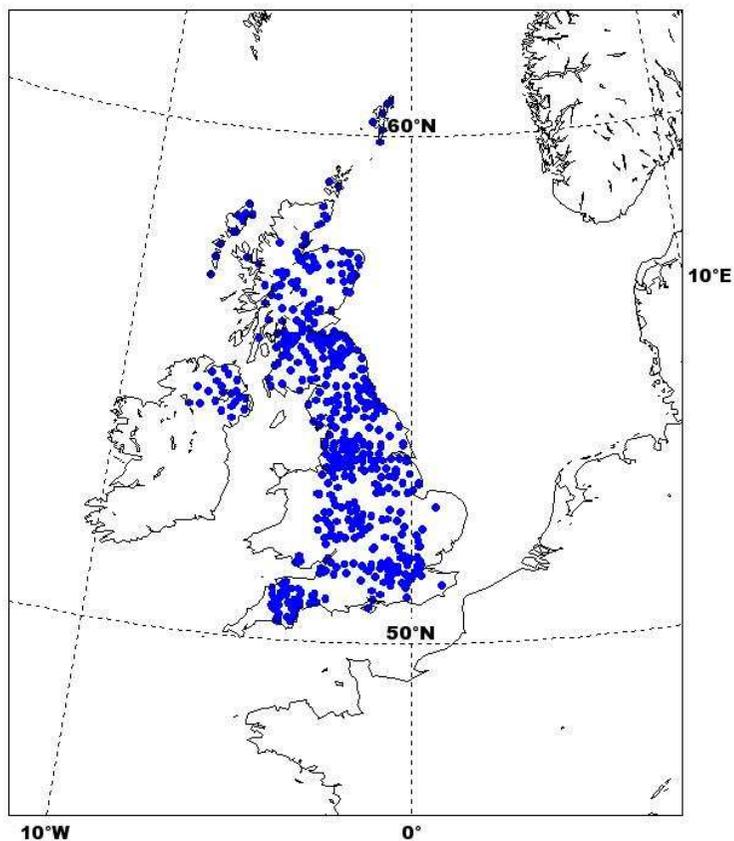
Met Office

Roadside sensor network

OpenRoad

Data Coverage: Surface (20/2/2010, 6 UTC)
Total number of observations assimilated: 1507

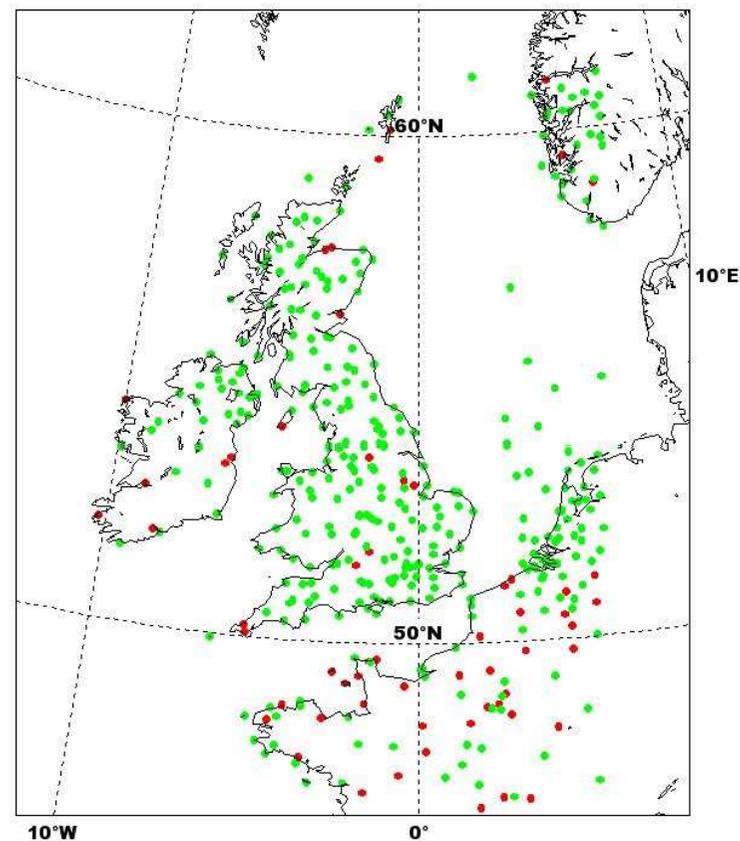
OPENROAD (1507)



SYNOP

Data Coverage: Surface (20/2/2010, 6 UTC)
Total number of observations assimilated: 1150

SYNOP (201) SYNOP AUTO (949) SYNOP MIXED (0)





Roadside sensor network impact

Mean T2m error

— control

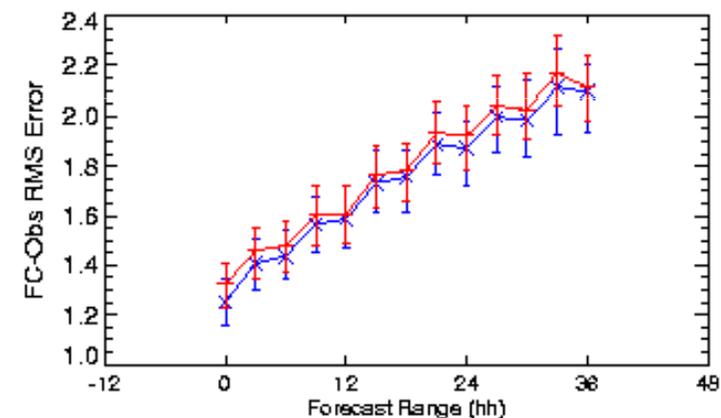
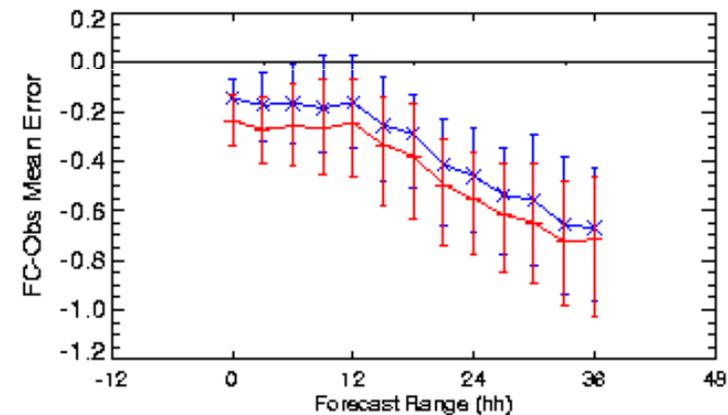
— test

Rms T2m error

(2nd half of Dec 2010)

Temperature (Kelvin) at Station Height
Reduced UK
Equalized and Meaned from 1.

Cases: —+— UK4 PS25 Control —x— UK4 PS25 with All OpenF





Symmetric Cloud Observation Errors

- $\varepsilon_{ob} = \text{fn} \{ (CF_{ob} + CF_b)/2 \}$
- $\varepsilon_{ob} = 0.55$ when $CF_{ob} = CF_b = 0$
- $\varepsilon_{ob} = 0.25$ when $CF_{ob} = CF_b = 1$.
- Previously ε_{ob} was $\text{fn}(CF_{ob})$ only, which tended to result in too much cloud



Symmetric Cloud Errors

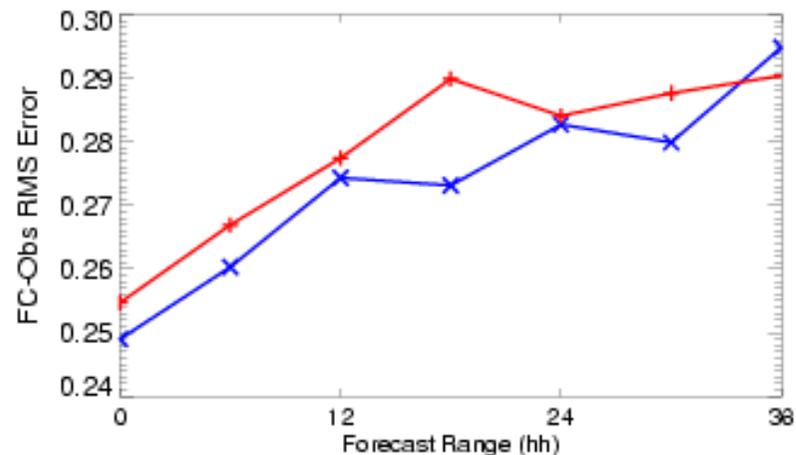
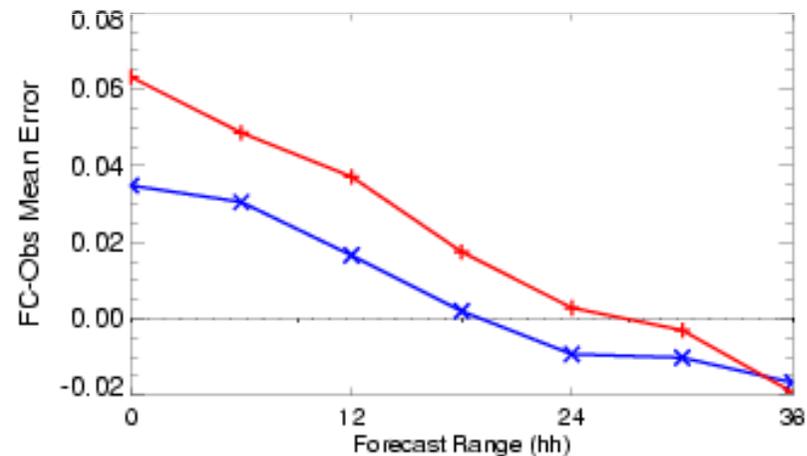
Sc Period – October 2010

Cloud Verification

mean cloud
fraction error

rms cloud
fraction error

Cases: + Control 0z x Symmetrical Clouds





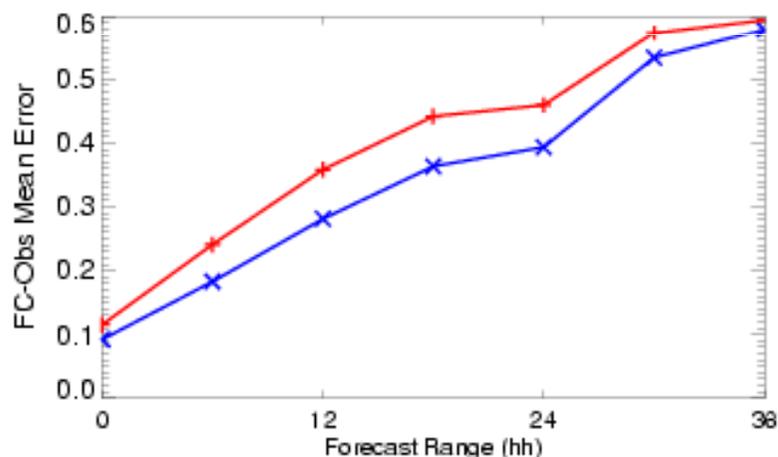
Symmetric Cloud Errors

Sc Period – October 2010

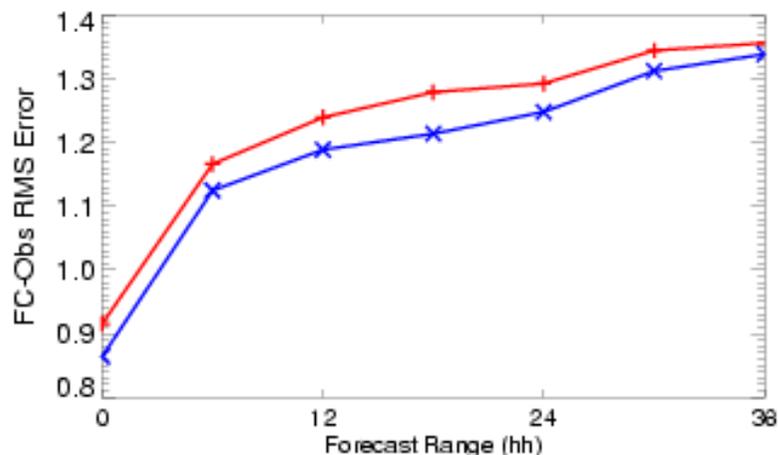
T2m Verification

mean T2m
error

Cases: + Control 0z x Symmetrical Clouds



rms T2m
error

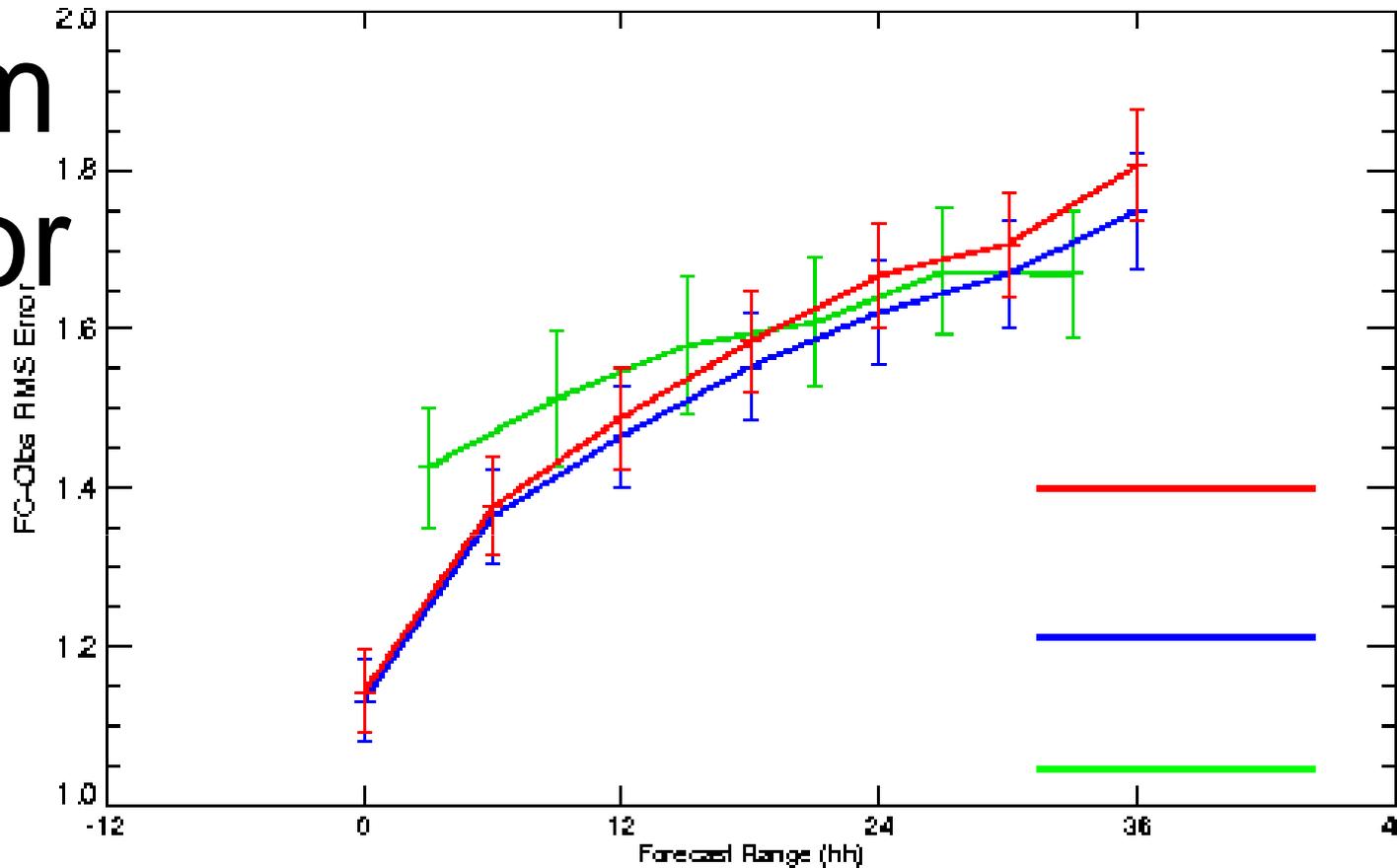




Impact of high resolution assimilation in UK4 model

Rms

T2m
error



CTRL

GL lbc

**GL lbc +
GL t-3
analysis**

Mark Weeks

fc time (hr)

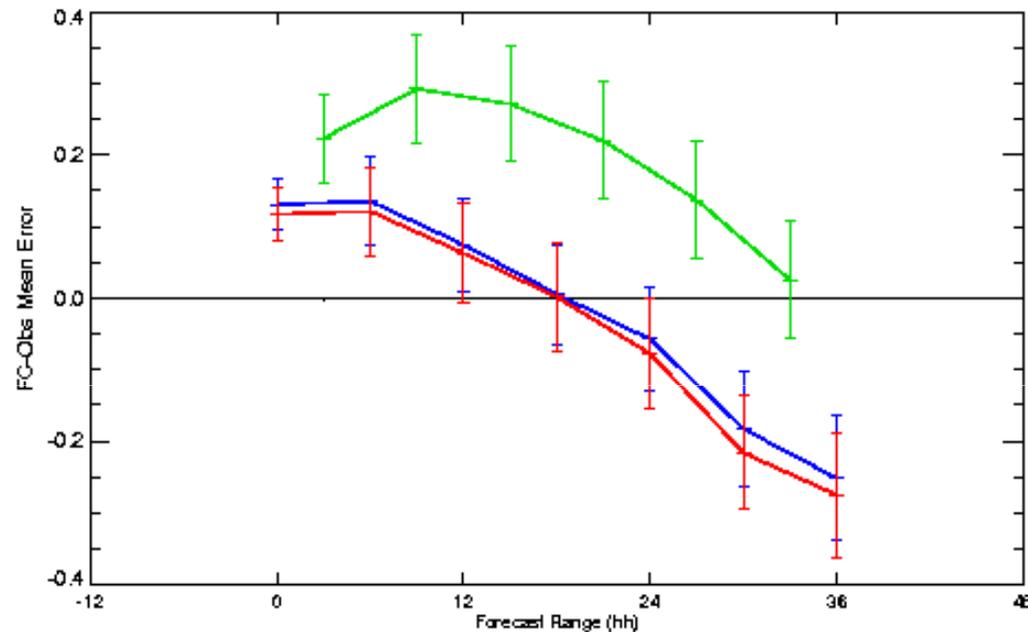


Impact of high resolution assimilation

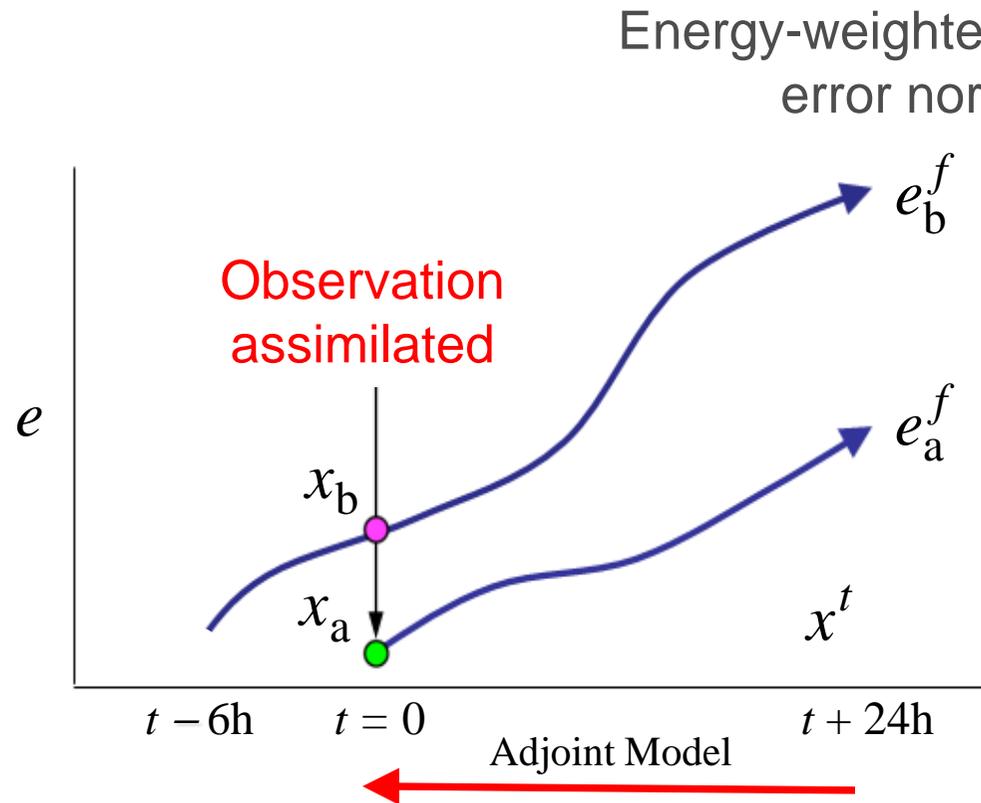
mean
T2m
error

Cases: — UK4 — UK4 Global LBCs — UKX

Areas: +—+ WMO Block 03 station list



Forecast Sensitivity to Observations (FSO)



Observation impact is quantified as the difference in forecast error norm

$$\delta e = e_a^f - e_b^f$$

Adjoint of NWP model/DA system used to derive analysis/observation sensitivity.

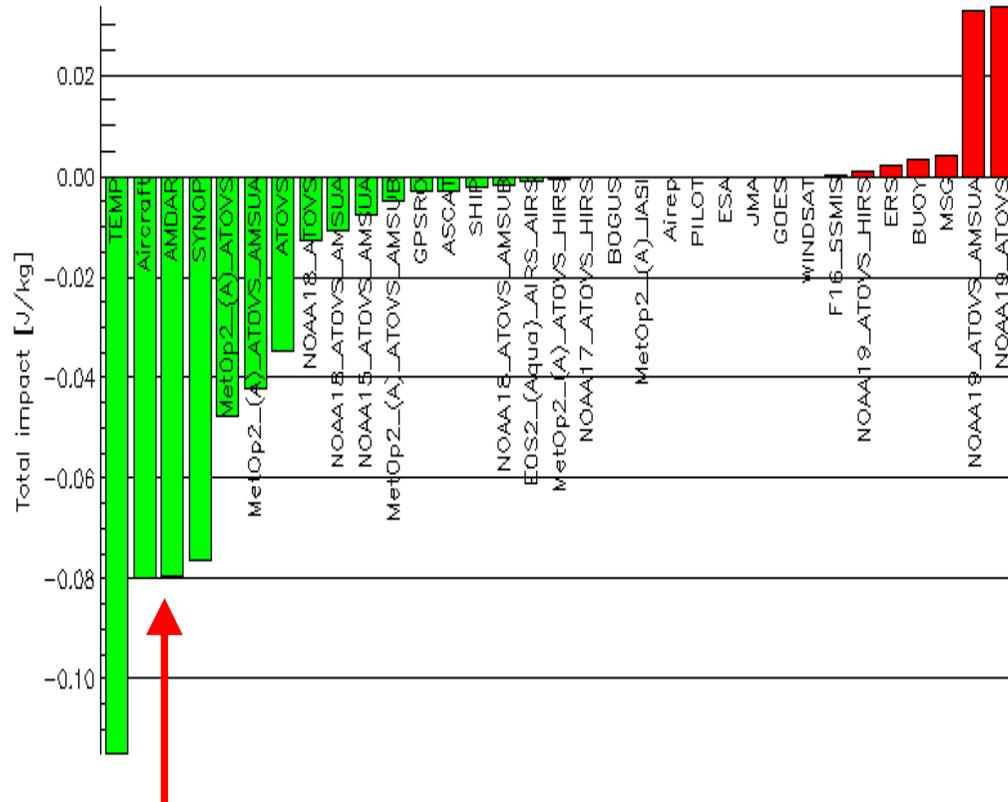
- FSO technique can estimate individual contribution of **every** observation.
- Reduces need for expensive data denial experiments.
- Assists optimal design of observation networks.



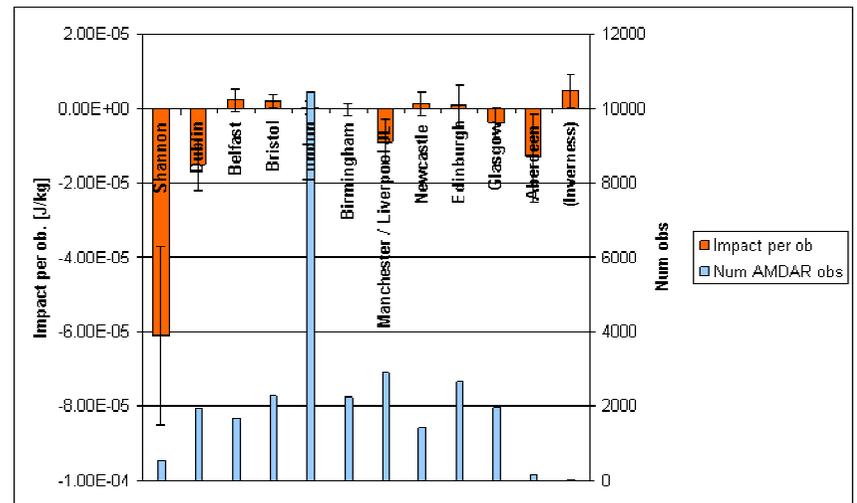
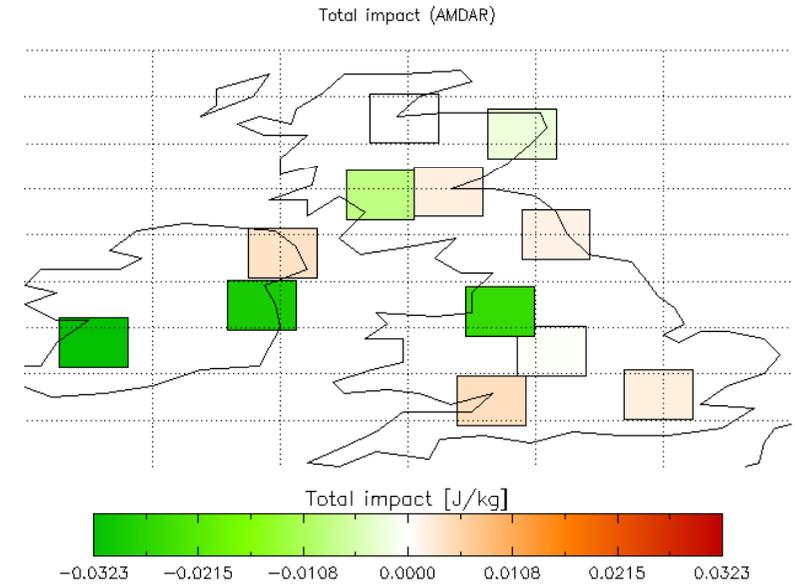
FSO Example Study: EasyJet AMDARs

Met Office

3rd November – 13th December 2010 Global Model T+24 Forecast Impact Study

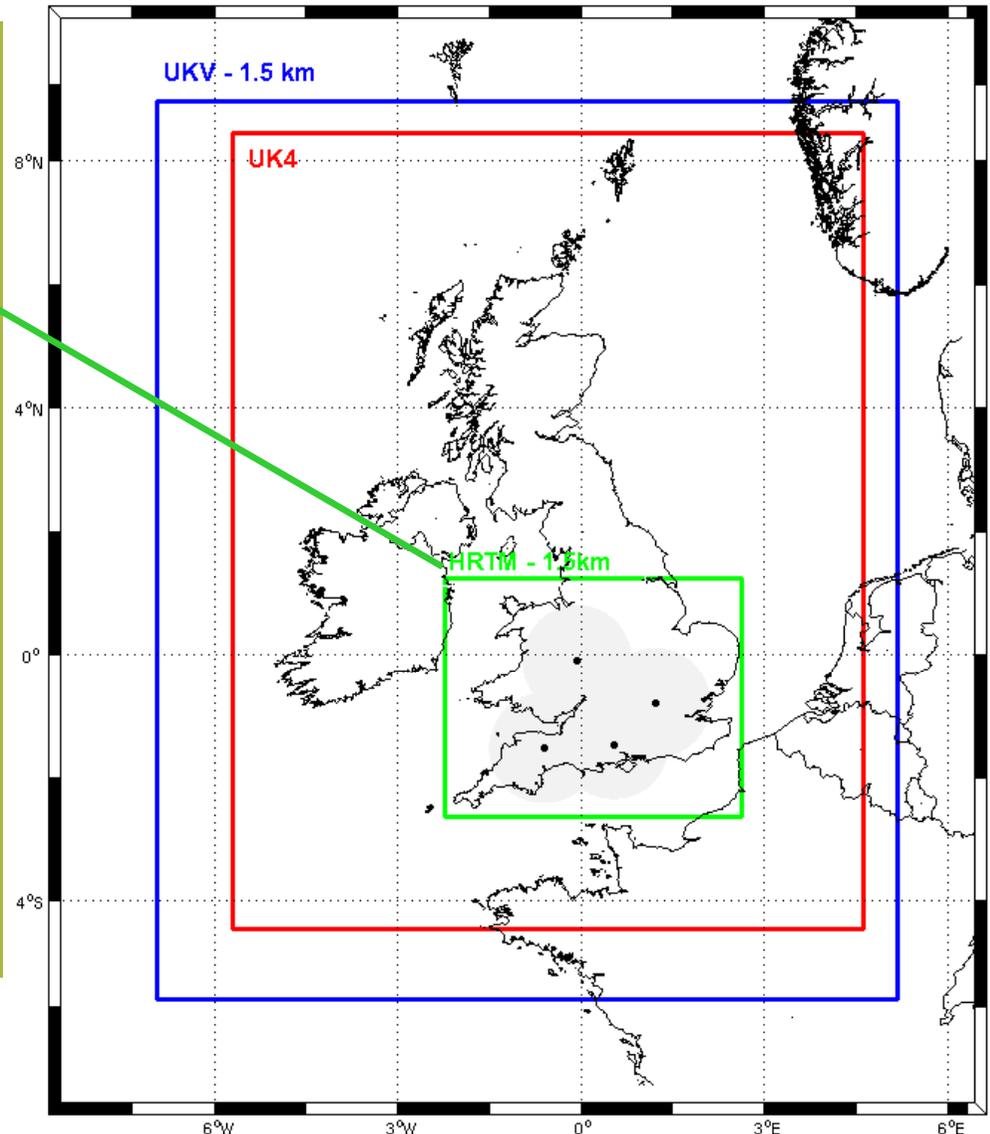


- AMDAR impact #2 after TEMP
- Significant regional variation in impact.



Nowcasting Demonstration Project

- 1.5 km NWP-based nowcasting system
- Southern UK only
- Hourly cycling, ~12 hr forecasts
- To be run in experimental mode during London's summer Olympics in 2012
- 3/4DVAR with Doppler winds/reflectivities.





Future plans for UK assimilation

- add large scale analysis increments from driving model whenever lateral boundary forcing is updated
- focus analysis increments on smaller horizontal scales
- replace MOPS cloud with separate assimilation of surface cloud obs and satellite cloud top
- expand network of Doppler radars
- assimilation of reflectivity (indirect / direct)
- SEVIRI radiances (clear and cloudy, sea and land)
- high resolution AMV's



Met Office



Questions?