



Application of PCA to IASI: An NWP Perspective

Andrew Collard, ECMWF

Acknowledgements to Tony McNally, Jean-Noel Thépaut



Overview

- **Introduction**
 - **Why use PCA/RR?**
- **Expected performance of PCA/RR for IASI**
- **Assimilation of RR: AIRS Experience**
 - **Assimilation of “Normal” Radiances**
 - **Assimilation of Reconstructed Radiances**
- **Conclusions**



Why Use PCA/RR?



Why is data compression important?

- Very large data volumes need to be communicated in near-real time (e.g., EUMETSAT to NWP centres)
- Simulation of spectra (needed for assimilation) is costly
- Data storage



Expected Performance of PCA/RR for IASI?



Spectral data compression with PCA*

The complete AIRS spectrum can be compressed using a truncated principal component analysis (e.g. 200PCAs v 2300 rads)

**Leading eigenvectors (200, say)
of covariance of spectra from
(large) training set**

$$\mathbf{p} = \mathbf{V}^T (\mathbf{y} - \bar{\mathbf{y}})$$

Coefficients (blue arrow pointing to \mathbf{p})

Mean spectrum (orange arrow pointing to $\bar{\mathbf{y}}$)

Original Spectrum (green arrow pointing to \mathbf{y})

- To use PCs in assimilation requires an efficient RT model to calculate PCs directly
- PCs are more difficult to interpret physically than radiances

N.B. This is usually performed in noise-normalised radiance space

This allows data to be transported efficiently



Spectral data compression and de-noising

The complete AIRS spectrum can be compressed using a truncated principal component analysis (e.g. 200PCAs v 2300 rads)

**Leading eigenvectors (200, say)
of covariance of spectra from
(large) training set**

**Reconstructed
spectrum**

$$\mathbf{p} = \mathbf{V}^T (\mathbf{y} - \bar{\mathbf{y}})$$

Coefficients (blue arrow pointing to \mathbf{p})

Mean spectrum (orange arrow pointing to $\bar{\mathbf{y}}$)

Original Spectrum (green arrow pointing to \mathbf{y})

$$\mathbf{y}_R = \bar{\mathbf{y}} + \mathbf{V}\mathbf{p}$$

N.B. This is usually performed in noise-normalised radiance space

Each reconstructed channel is a linear combination of all the original channels and the data is significantly de-noised.

If N PCs are used all the information is contained in N reconstructed channels (theoretically)



Transformation of Instrument Noise

Error in observation, $\boldsymbol{\varepsilon}_y$, is transformed by:

$$\boldsymbol{\varepsilon}_{y_R} = \mathbf{V}\mathbf{V}^T \boldsymbol{\varepsilon}_y + \boldsymbol{\varepsilon}_R$$

Where $\boldsymbol{\varepsilon}_R$ is the reconstruction error.

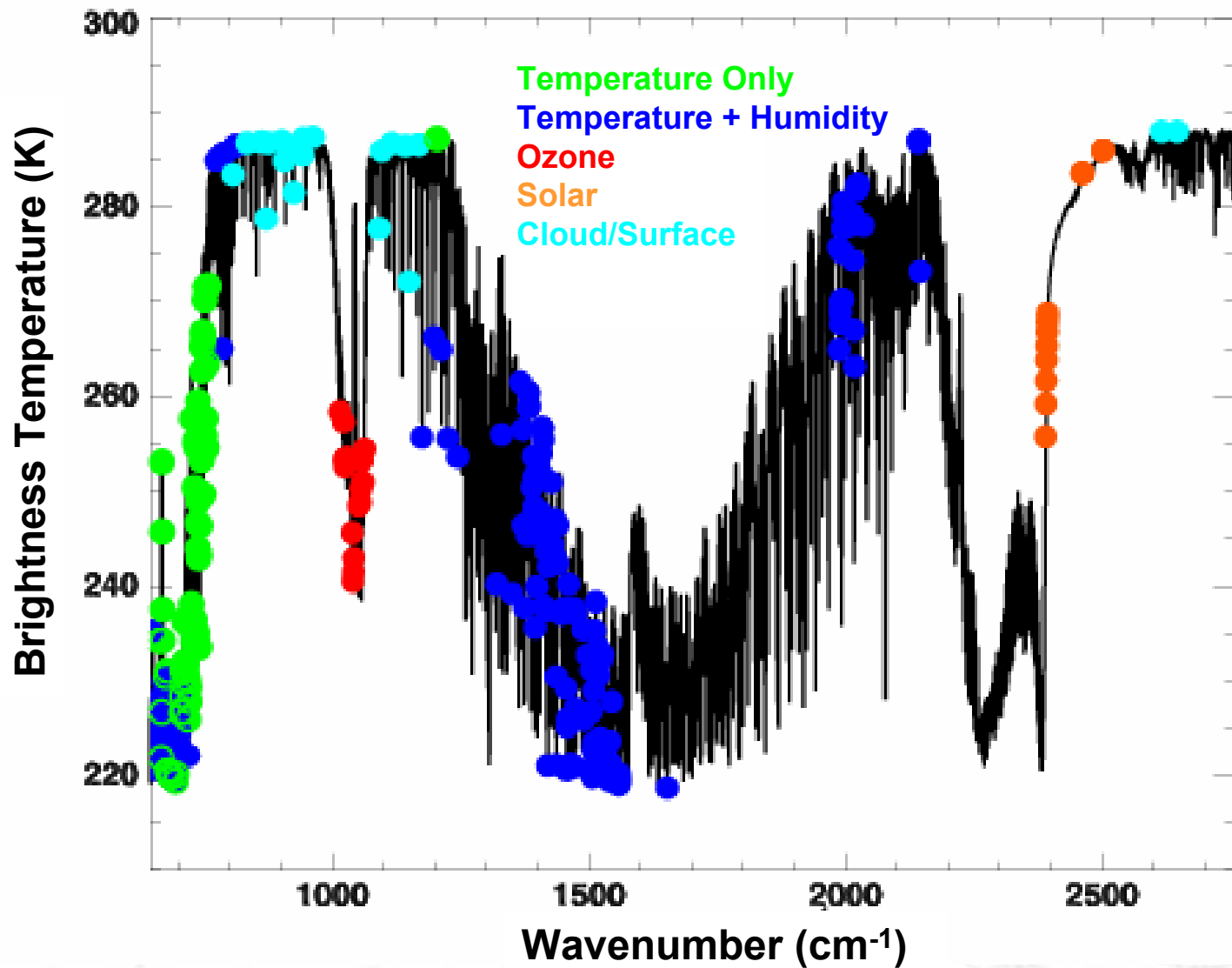
\therefore Observation error covariance, \mathbf{O}_R , of reconstructed radiances:

$$\mathbf{O}_R = \mathbf{V}\mathbf{V}^T \mathbf{O}\mathbf{V}^T \mathbf{V} + \mathbf{F}_R$$

In addition, the *forward model term* is not modified by reconstruction.

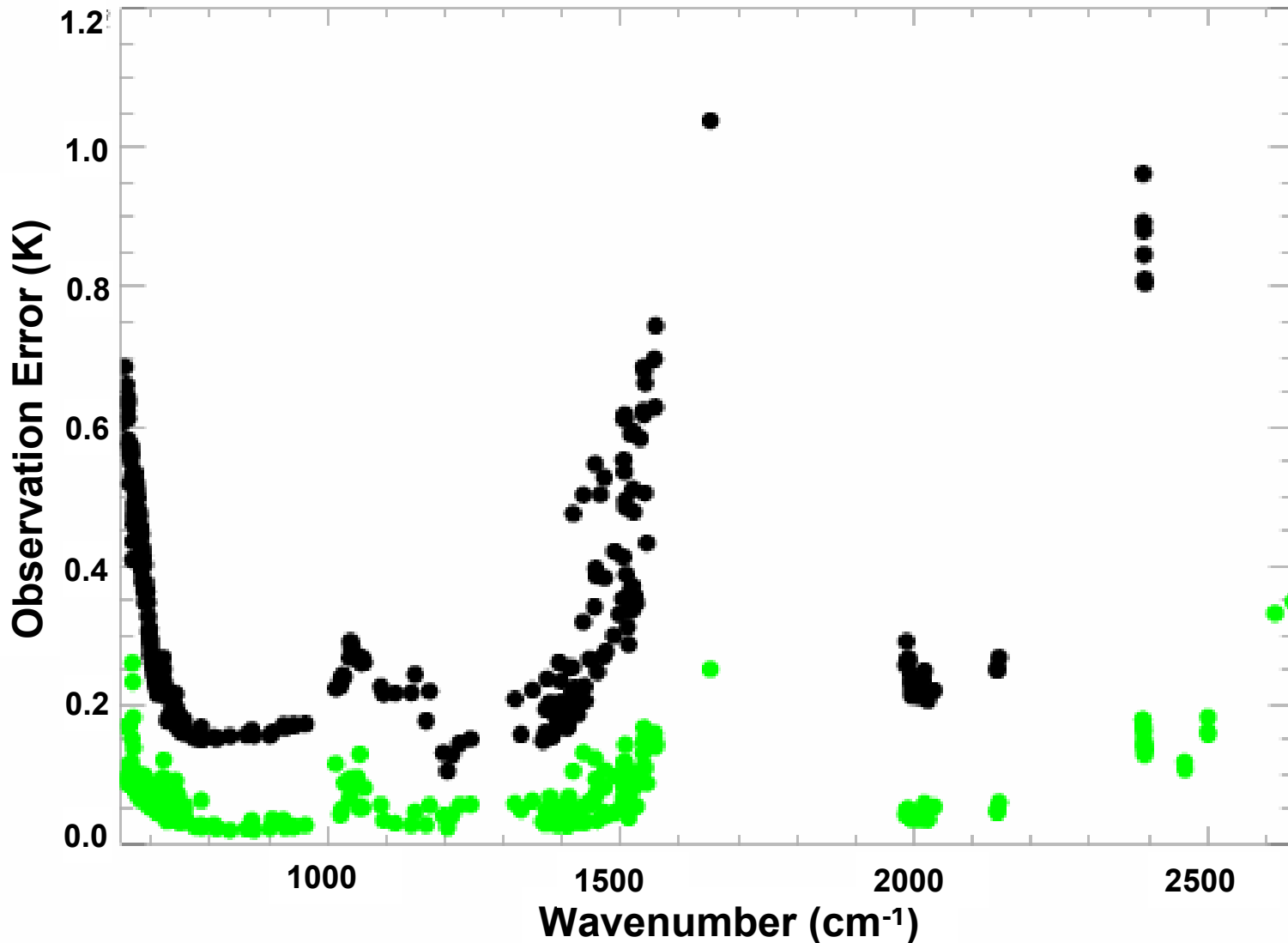


300 Channels for IASI



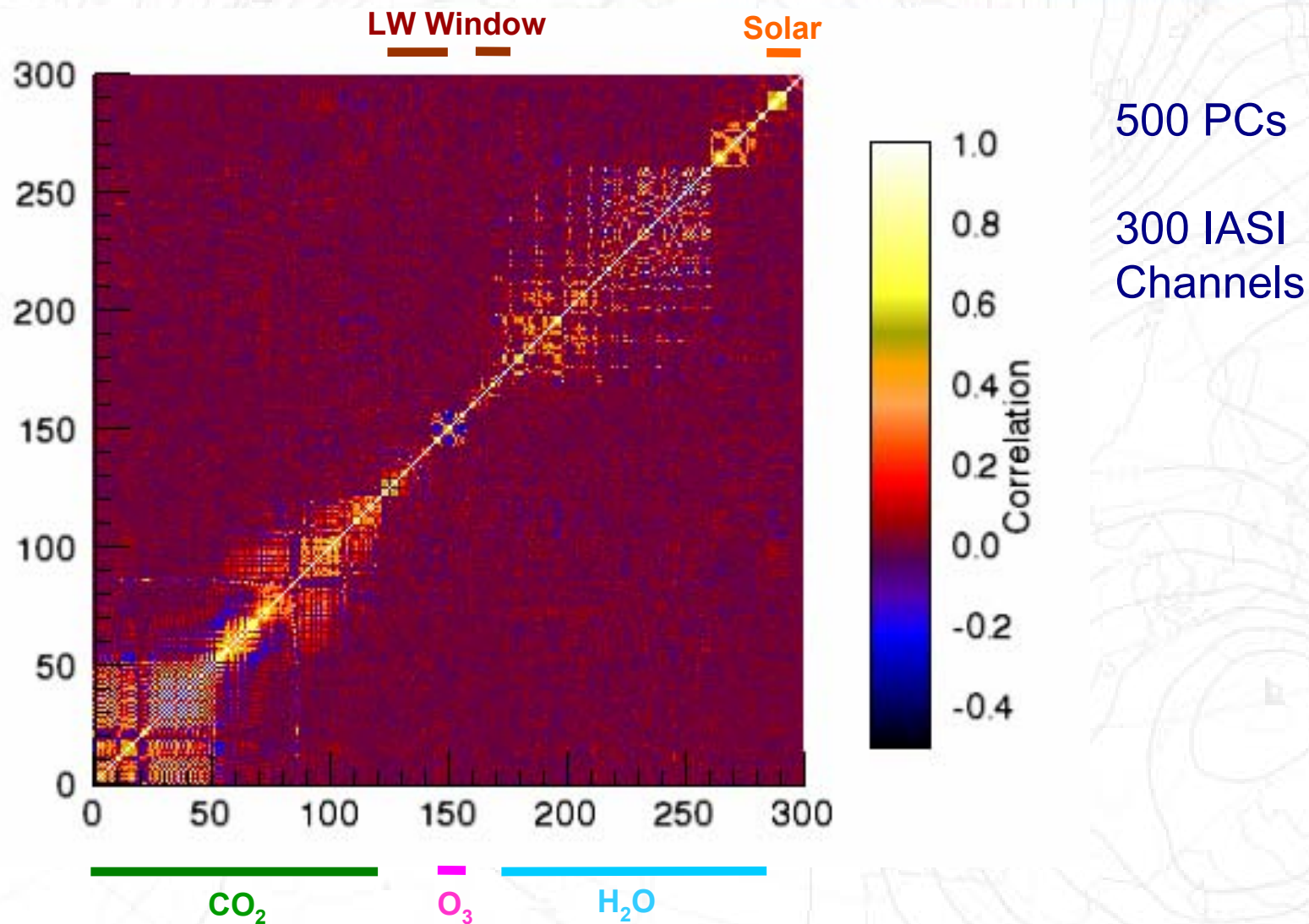


Noise Reduction – 300 Channels



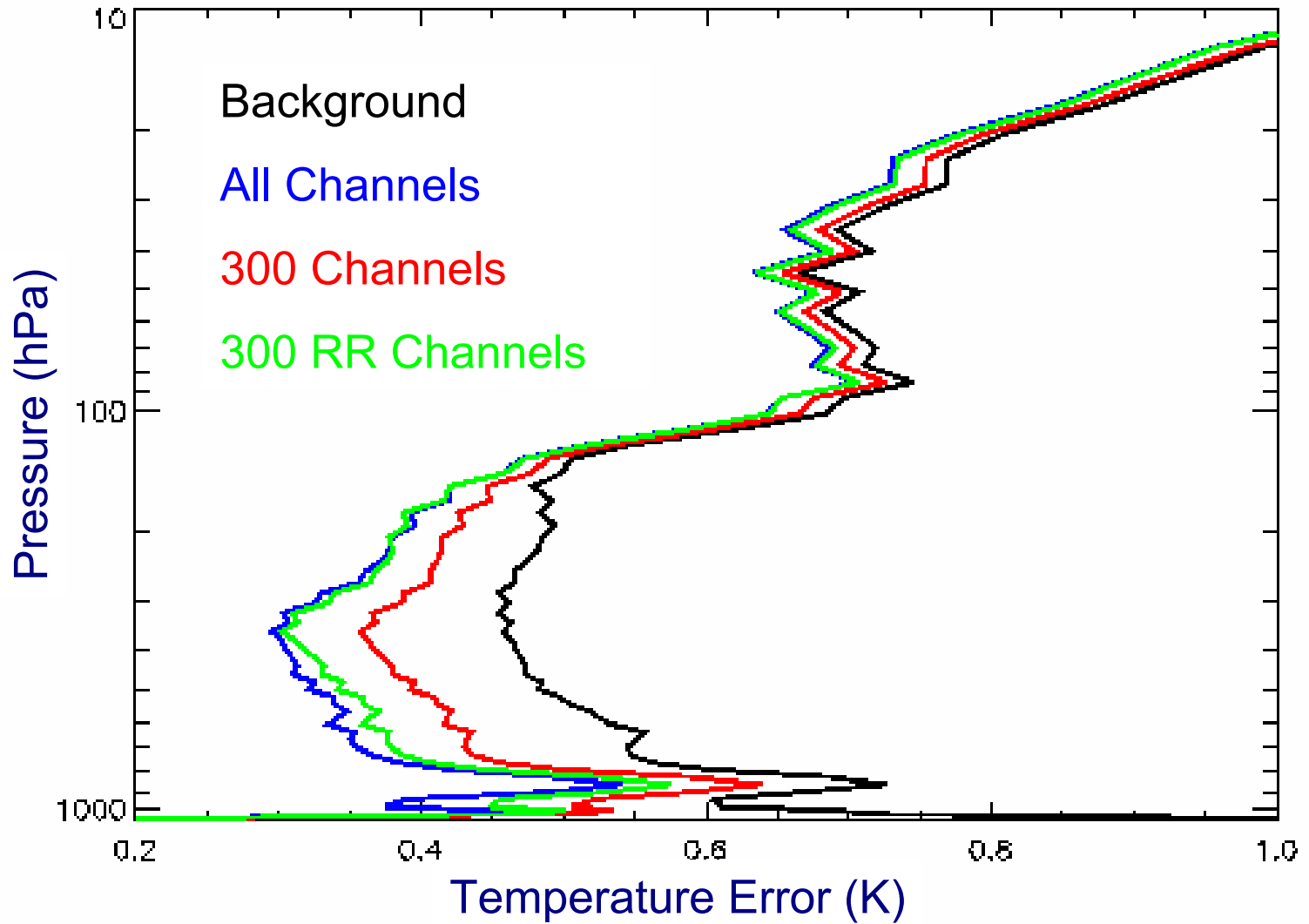


IASI RR Error Correlations





Retrieval Performance





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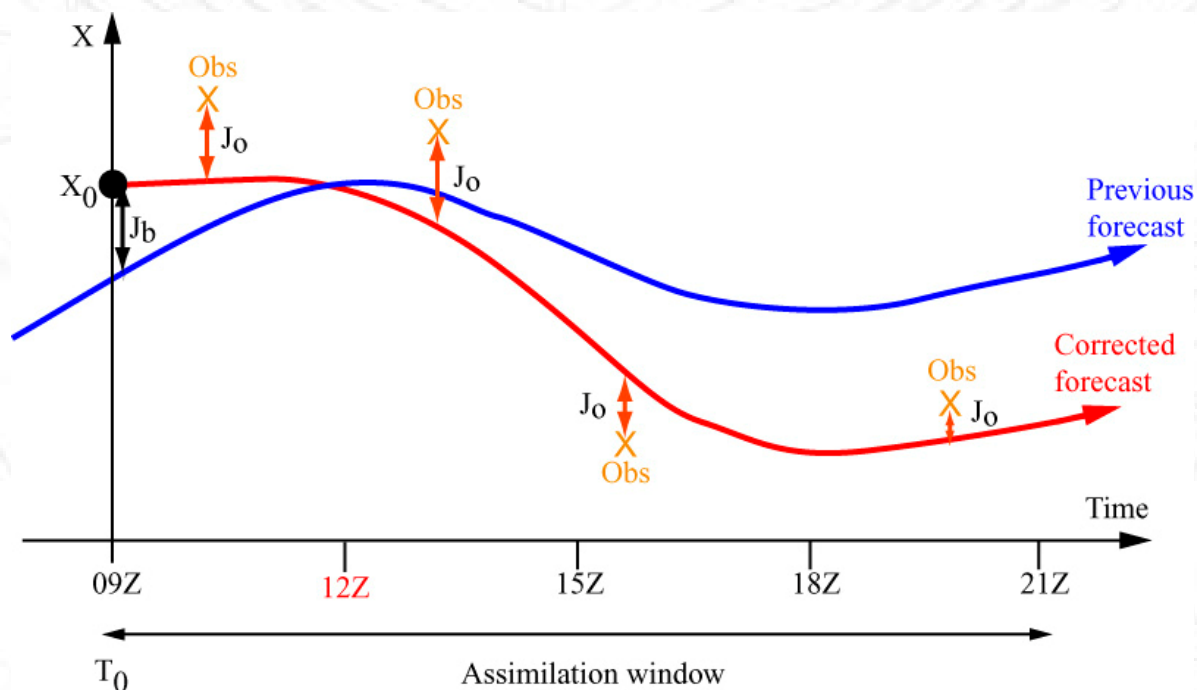
Experience with AIRS: Current Operational System



4D-Var Data Assimilation

4-dimensional variational data assimilation is in principle a least-squares fit in 4 dimensions between the predicted state of the atmosphere and the observations.

The adjustment to the predicted state is made at time T_0 , which ensures that the analysis state (4-dimensional) is a model trajectory.





Satellite data assimilated operationally at ECMWF

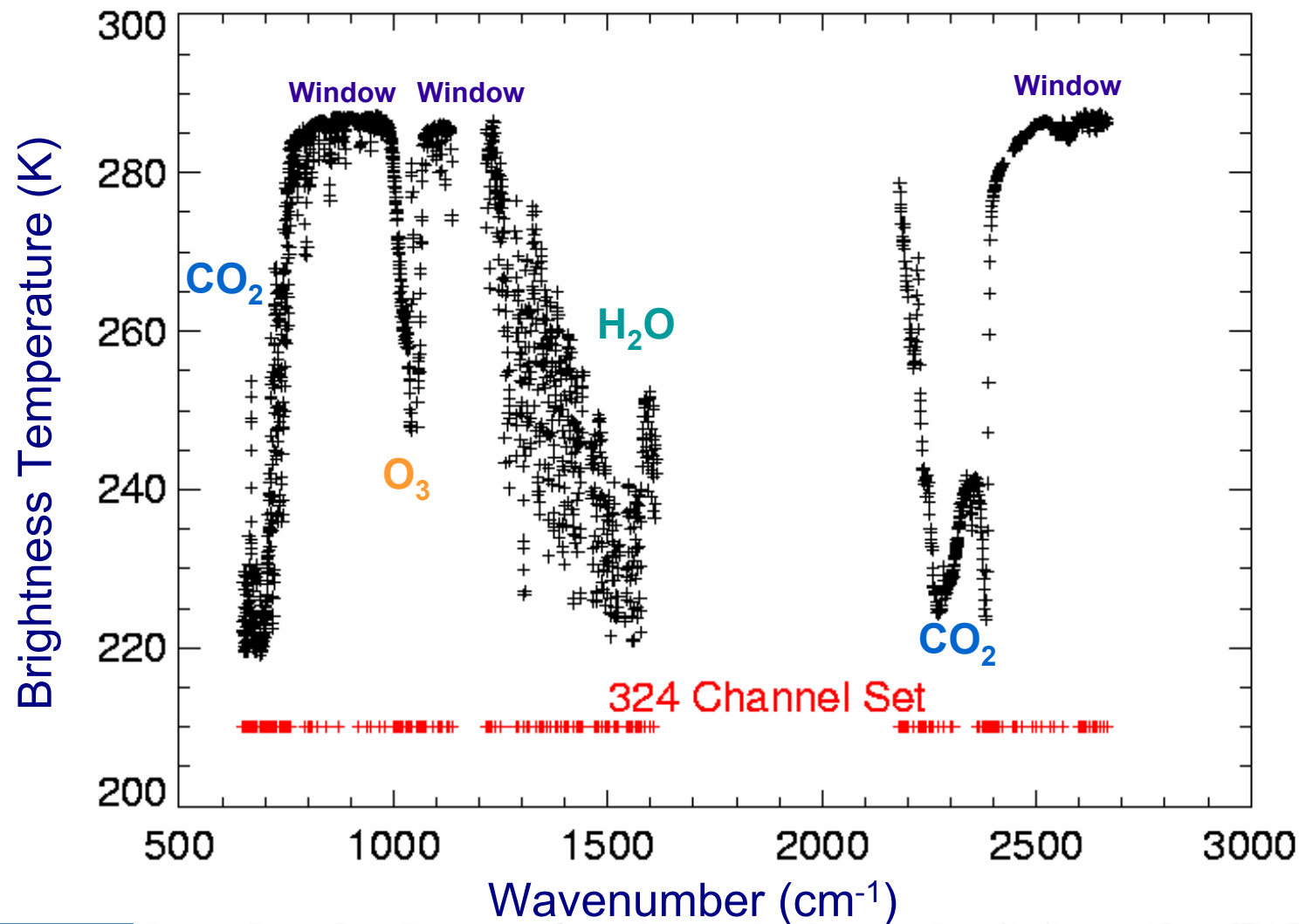
- 4xAMSU-A (NOAA-15/16/18 + AQUA)
- 3xAMSU-B (NOAA-16/17/18)
- 3 SSMI (F-13/14/15) in clear and rainy conditions
- 1xHIRS (NOAA-17)
- AIRS (AQUA)
- Radiances from 4 GEOS (Met-5, Met-8, GOES-10/12)
- Winds from 4 GEOS (Met-5/8 GOES-10/12) and MODIS/TERRA+AQUA
- Scat winds from QuikSCAT and ERS-2 (Atlantic)
- Wave height from ENVISAT RA2 and ASAR, JASON
- Ozone from SBUV (NOAA 16) and SCIAMACHY (ENVISAT)

29 different satellite sources

Coming soon: SSMIS, radio occultation (GPS),...and IASI!



AIRS Spectrum – 324 Channel Subset

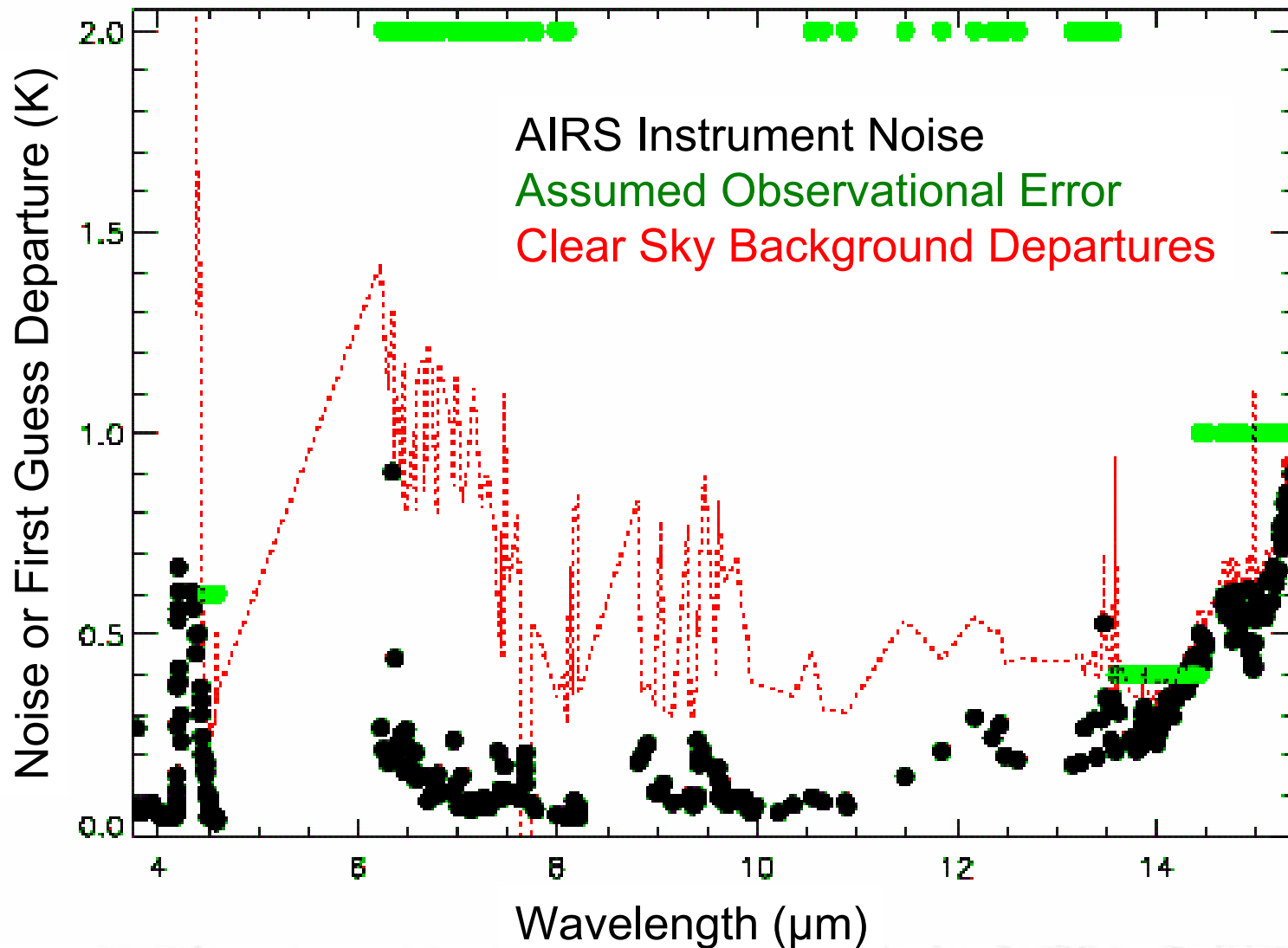


- 324 Channels (BUFR dataset)
- One AIRS FOV for every AMSU FOV (there are 9 AIRS FOVS per AMSU FOV)
- Up to 150 channels are assimilated (depending on cloud top height)

N.B. The same channels are supplied in NRT as reconstructed radiances



Assumed Noise for AIRS Assimilation





Optimal vs Non-Optimal Assimilation

- Optimal Assimilation

$$\mathbf{A} = \mathbf{B} - \mathbf{W}\mathbf{H}\mathbf{B}$$

- Non-Optimal Assimilation (Watts and McNally, 1988)

$$\mathbf{A}' = (\mathbf{I} - \mathbf{W}\mathbf{H})\mathbf{B}'(\mathbf{I} - \mathbf{W}\mathbf{H})^T + \mathbf{W}\mathbf{O}'\mathbf{W}^T$$

\mathbf{A}' , \mathbf{B}' and \mathbf{O}' are true values for \mathbf{A} , \mathbf{B} and \mathbf{O}

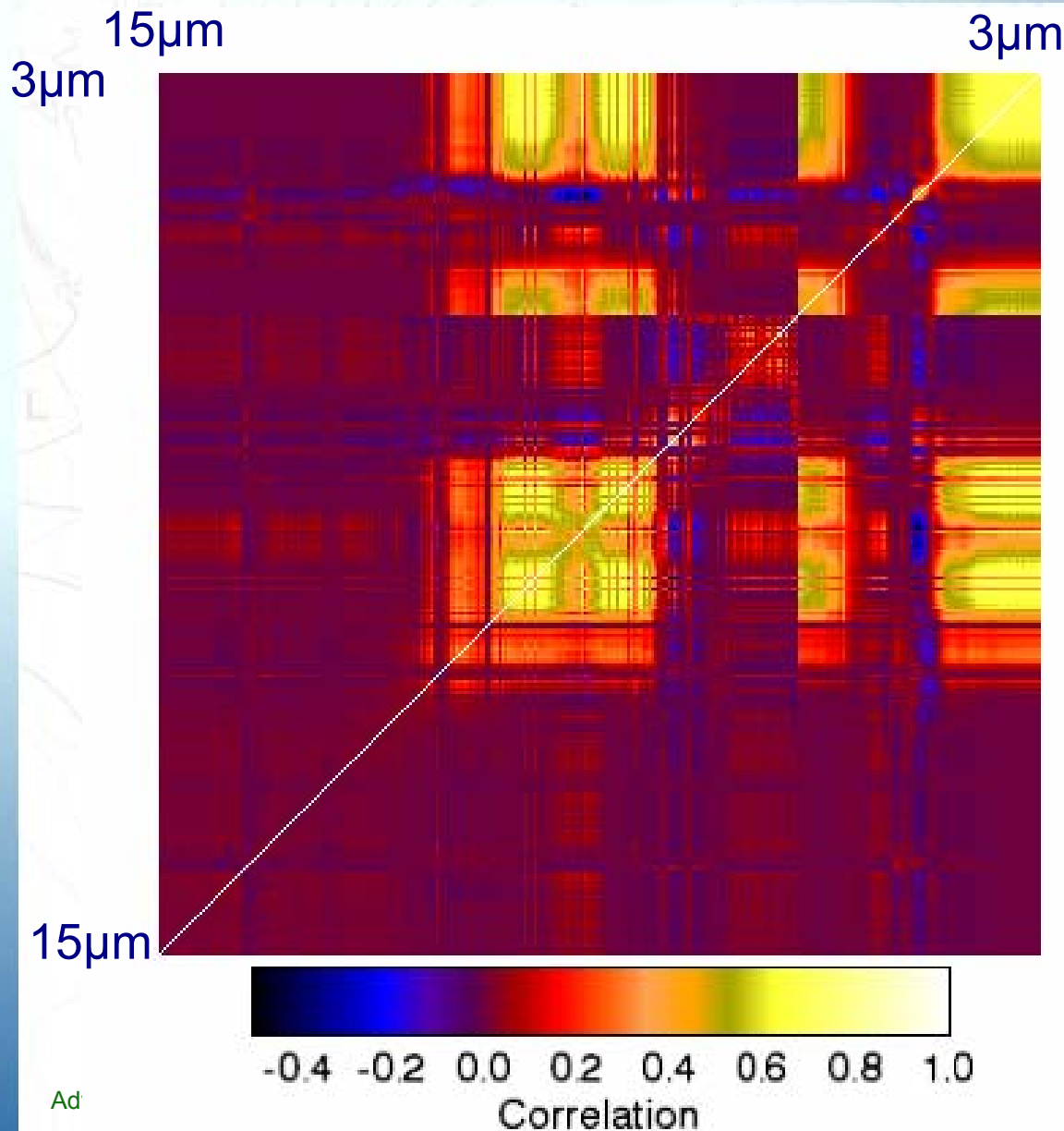
Where

$$\mathbf{W} = \mathbf{B}\mathbf{H}^T (\mathbf{H}\mathbf{B}\mathbf{H}^T + \mathbf{O})^{-1}$$

A=Analysis Error Covariance
B=Background Error Covariance
O=Observation Error Covariance
H=Jacobian



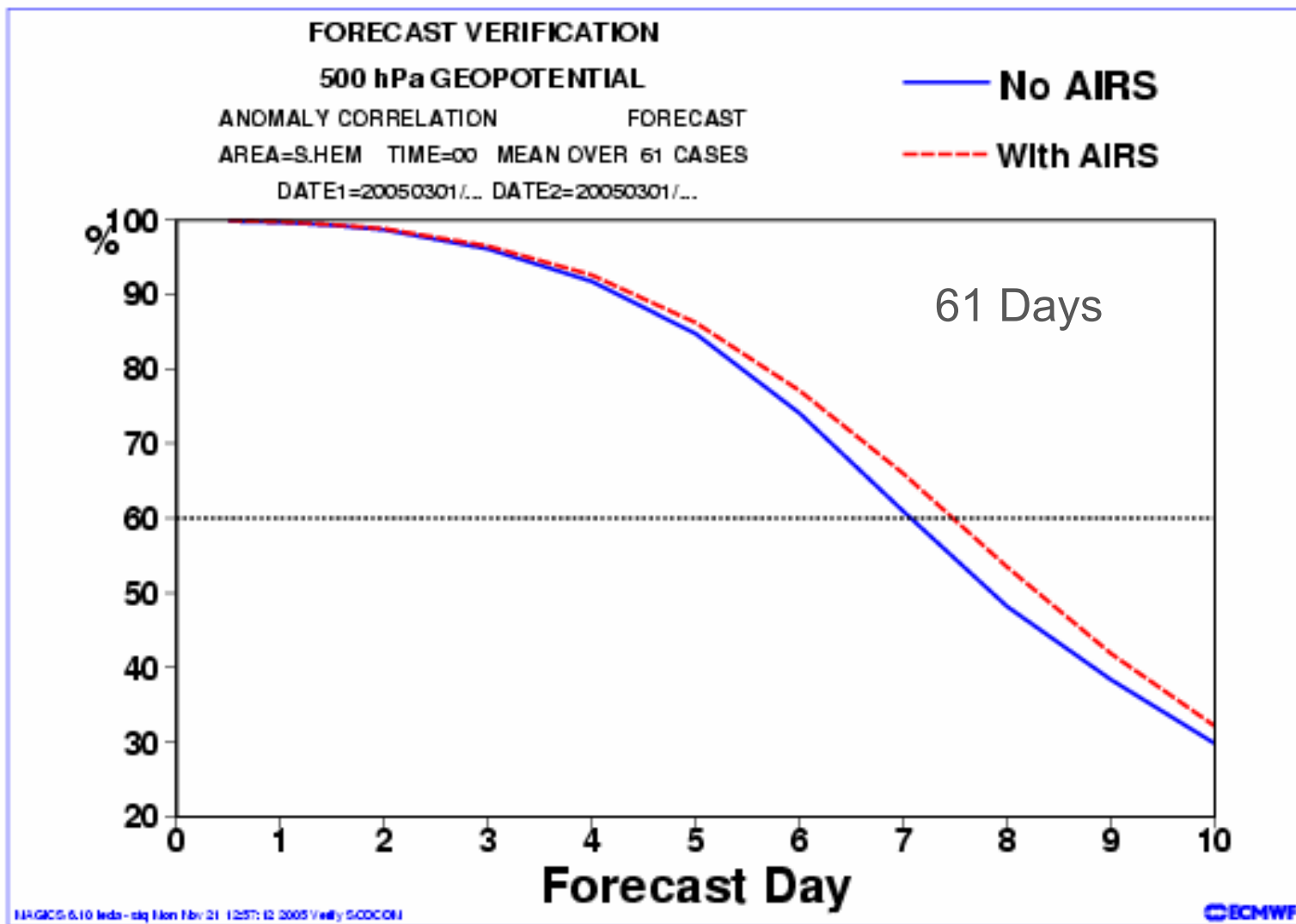
“Observation”* Errors are Correlated



*Instrument plus forward model noise



AIRS Impact up to 10 hours at 7 days

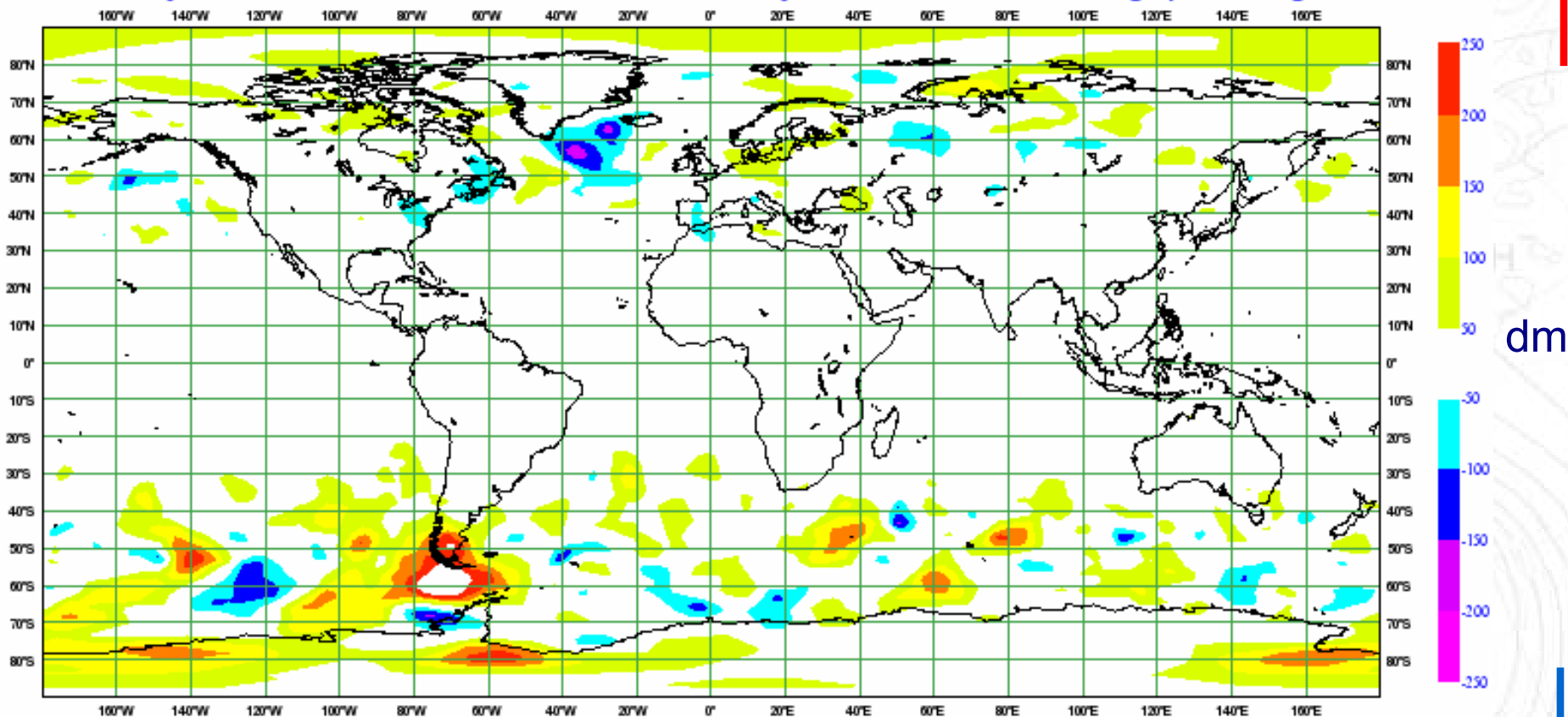




5 Day Forecast Improvements on Adding AIRS

AIRS Improves FC ↑

Tuesday 1 March 2005 00UTC ECMWF Forecast +120 VT: Sunday 6 March 2005 00UTC 500hPa **geopotential height



AIRS Degrades FC ↓

1st March – 30th April 2005

Advanced Sounder Workshop, Madison, 27th April 2006



Experience with AIRS: Reconstructed Radiances

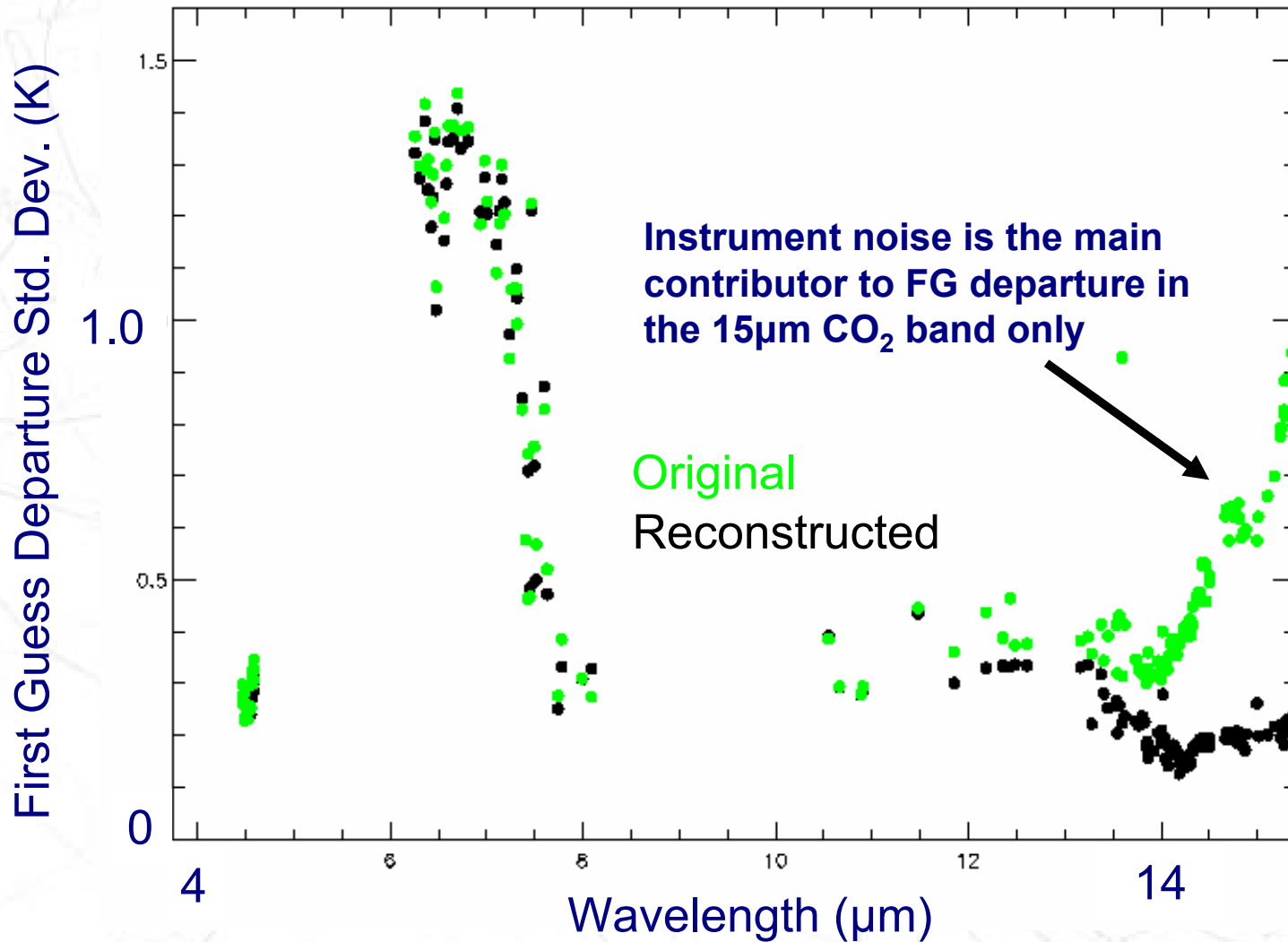


AIRS Reconstructed Radiances

- Data are supplied in near-real time by NOAA/NESDIS in the same format as the “real” radiances.
- The same channels are supplied, except some “popping” channels are missing
- Based on 200 PCs
- QC Flag supplied

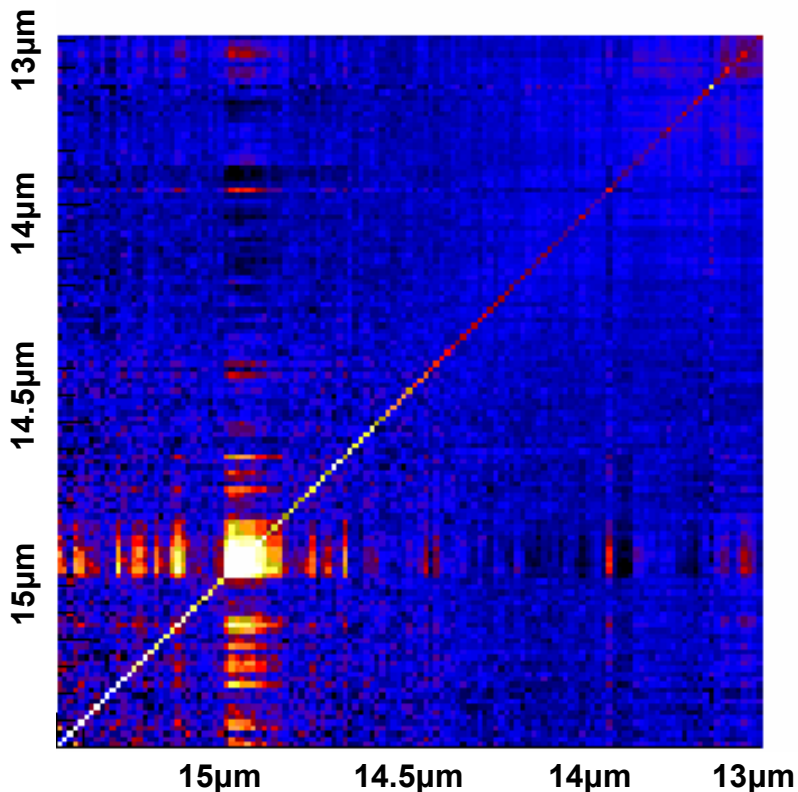


First Guess Departures for AIRS are Reduced





A look at Reconstructed Radiances' Errors

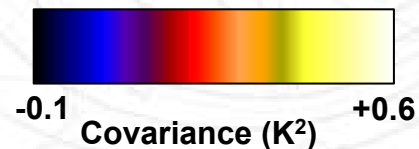
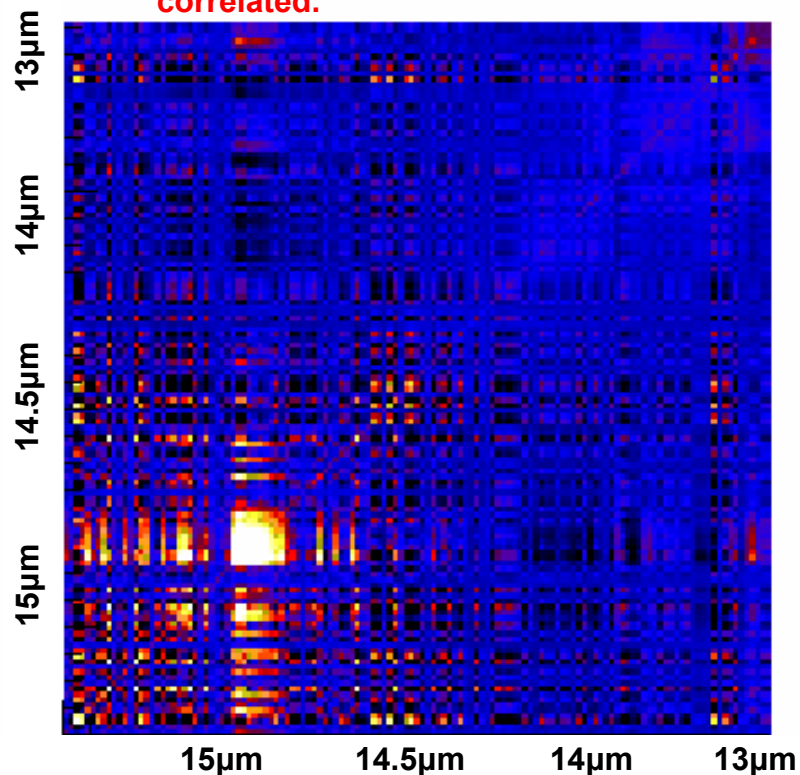


Original Radiances

Instrument noise is dominant and diagonal. Correlated noise is from background error

Reconstructed Radiances

Instrument noise is reduced (std. dev. is approximately halved) but has become correlated.



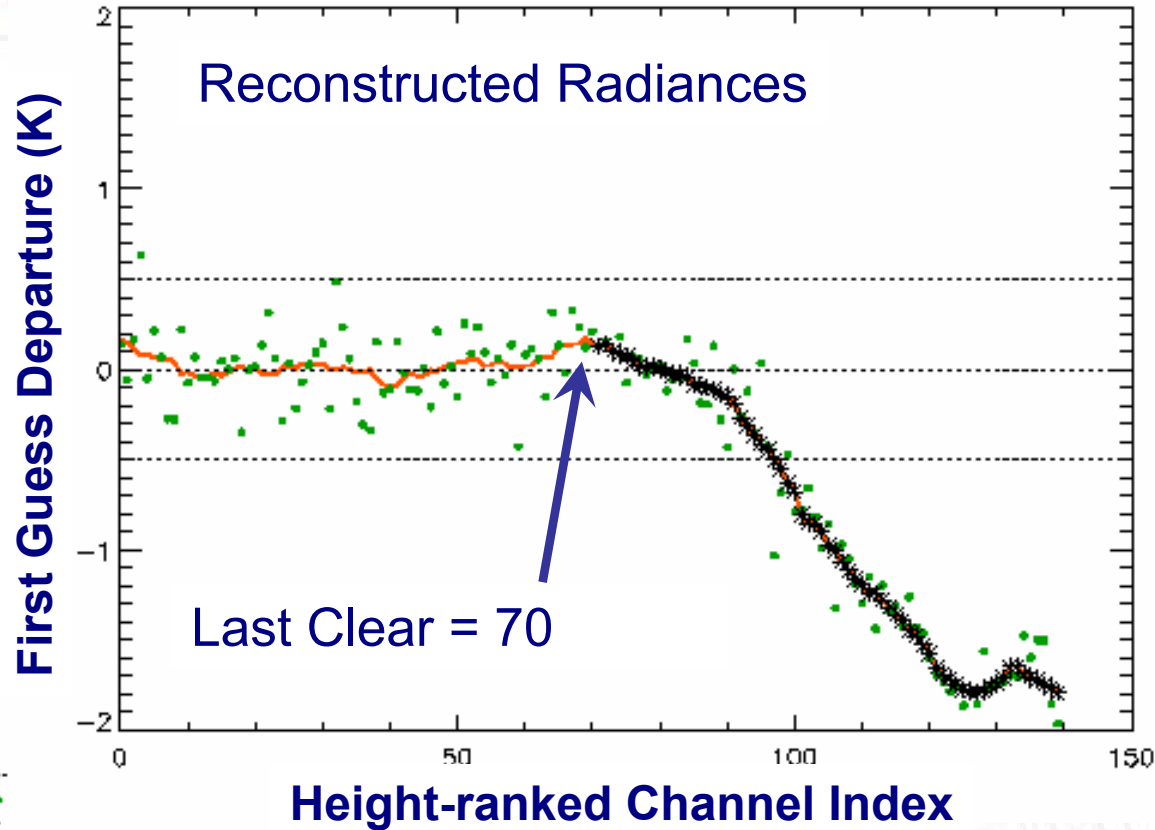
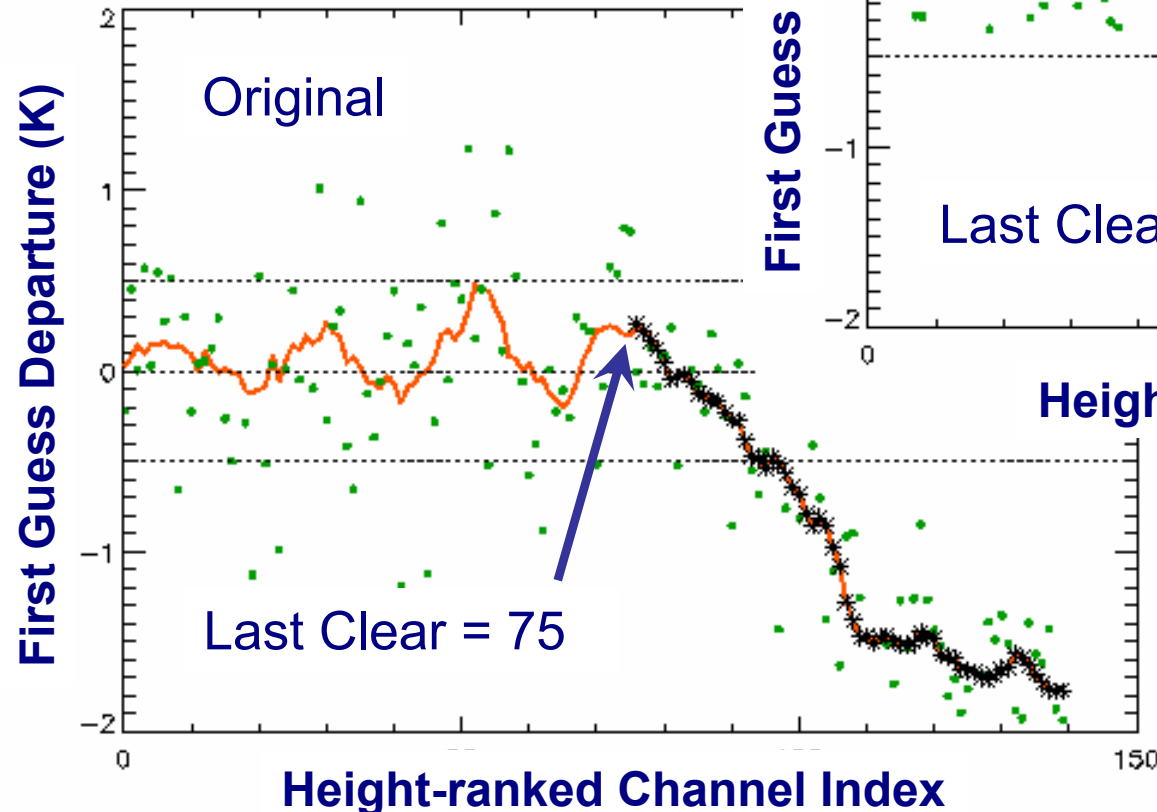
Covariances of background departures for clear observations in $15\mu m$ CO_2 band



Improvements in Cloud Detection

ECMWF Scheme:

- Ranks Channels Height
- Applies low-pass filter
- Tests for non-zero gradient



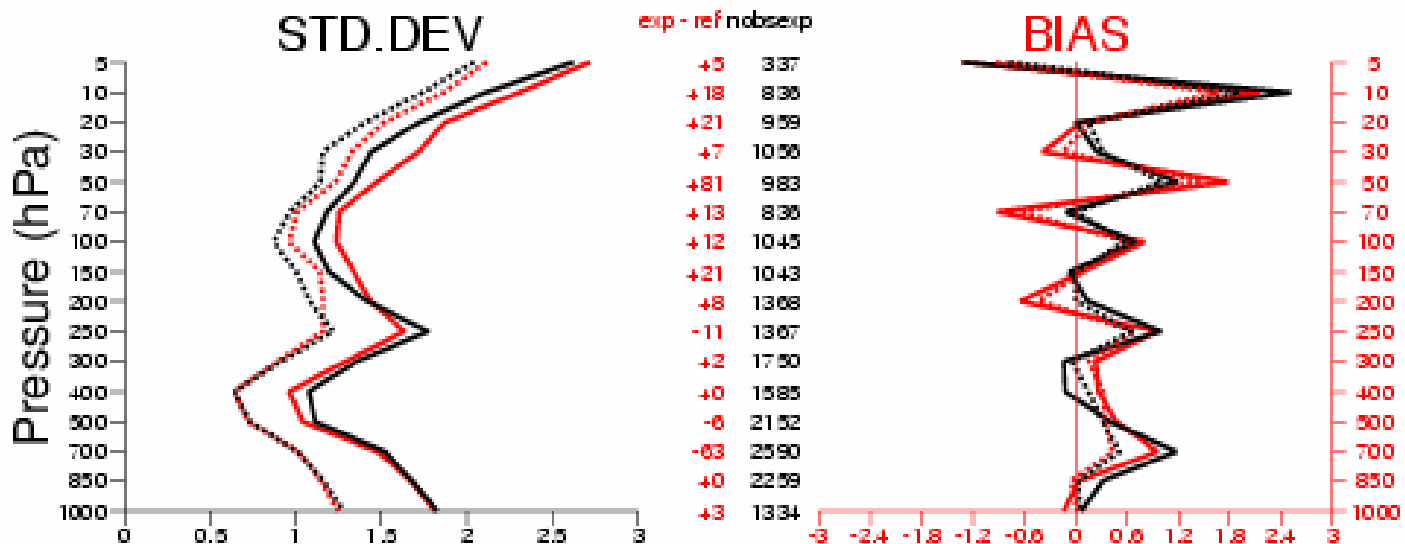
< 2% of Channels are flagged differently for RR vs Normal radiances



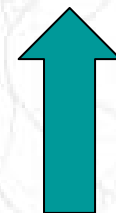
Improvements to Antarctic Stratosphere

exp:enx8 DA v en0z(ref) DA: 20050301 00-20050430 00(12)
TEMP-T S.PolarC
used T

— background departure o-b(ref)
— background departure o-b
- - - analysis departure o-a(ref)
..... analysis departure o-a

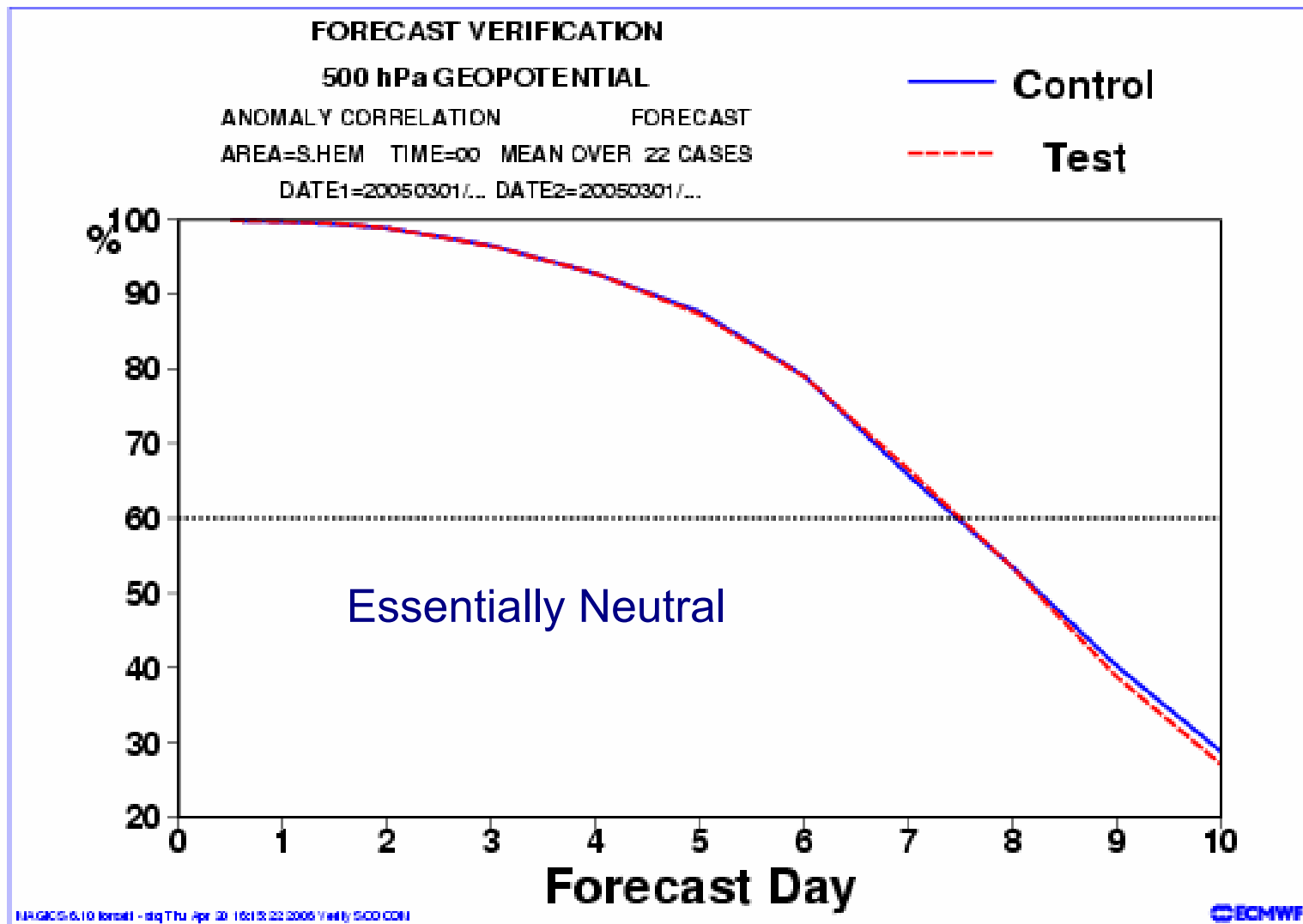


“Stratospheric Oscillation” in comparison to Antarctic radiosondes is greatly reduced on moving to reconstructed radiances



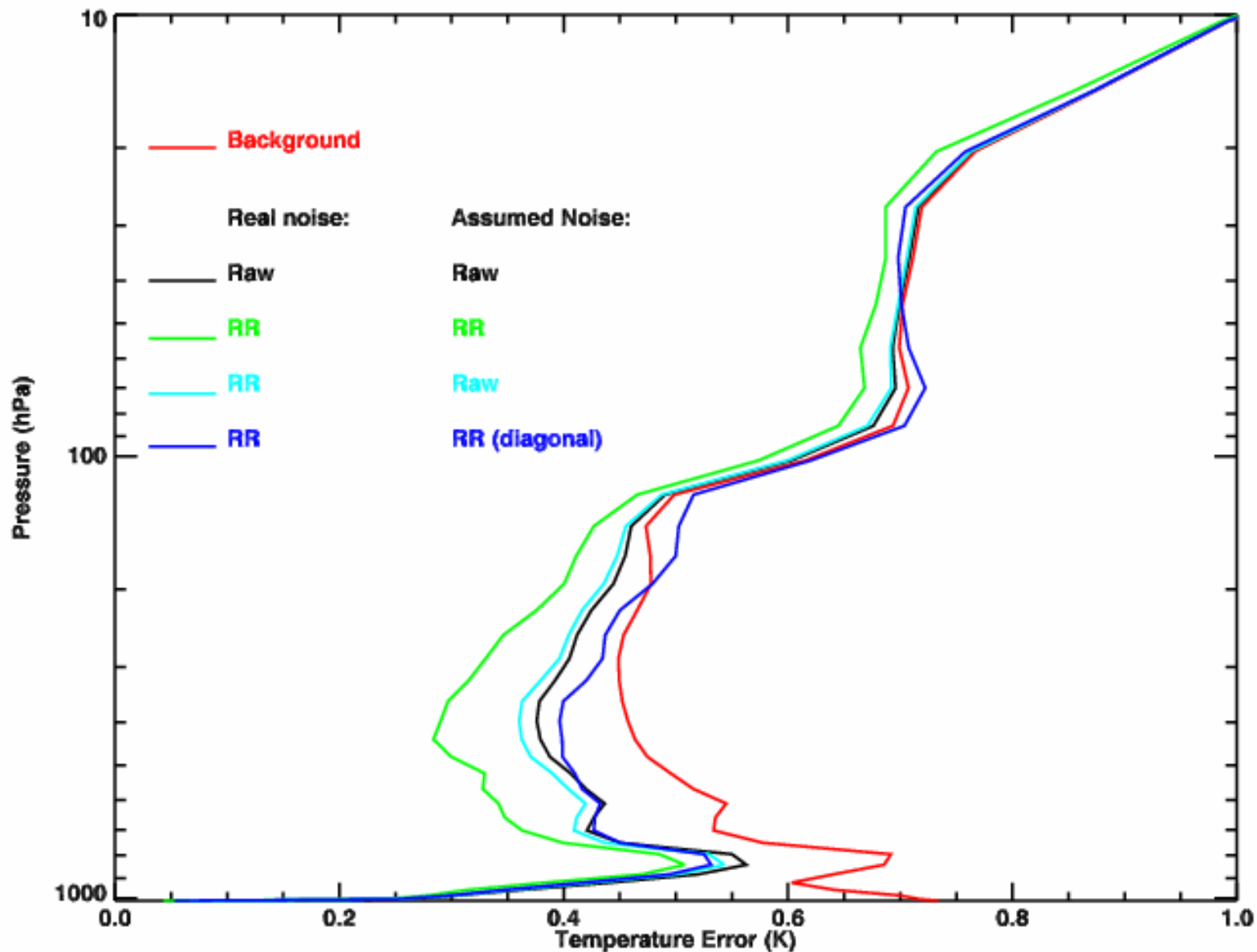


Forecast Impact of Reconstructed Radiances





Assimilating Reconstructed Radiances – Linear Theory





So why not use the RR data more aggressively?

- More aggressive errors have been tried:
 - Reduction of existing diagonal errors
 - “Bottom up” construction of full covariances
 - Hollingsworth-Lönnerberg approach
- Results have been neutral or negative with respect to:
 - Fit to other observations
 - Performance of short/medium range forecasts



Next steps

- Other error sources need to be investigated:
 - Bias
 - Spatially correlated error
 - Representivity error
 - ????



Conclusions

- The reconstructed radiances method has the potential to be particularly useful for the efficient representation of the information in IASI and AIRS.
- The noise smoothing properties of the reconstructed radiances method shows some positive benefits on assimilation into a high-resolution NWP system.
- However, the assimilation of reconstructed radiances has yet to yield positive impact on *forecast skill* (relative to “normal” radiances).
- Work on the construction of a suitable (non-optimal) error covariance continues.