



# **Progress of bias correction for satellite data at ECMWF**

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Tony McNally

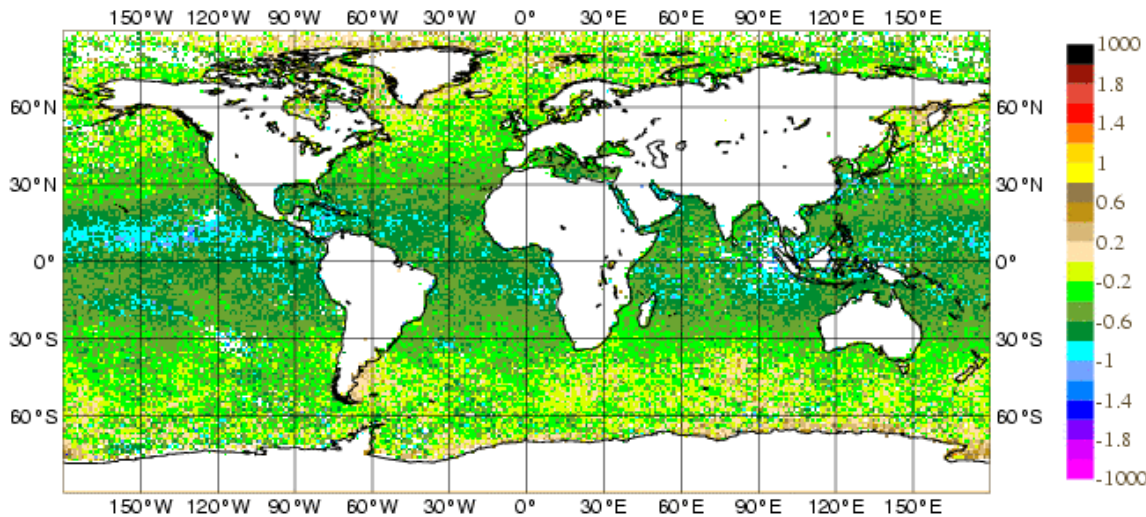
Acknowledgments to Phil Watts, Dick Dee



# Introduction

ECMWF 4DVar assimilation system requires that model and observations are unbiased with normally distributed errors.

But first-guess departures (*i.e.* observation minus equivalent from the model guess) show systematic errors.



- Bias model
- Adaptive bias correction
- Variational bias correction

Average departures over 2 weeks for NOAA17/HIRS14



# Bias model: correction strategy

➤ Scan correction

➤ Air-mass regression (Harris & Kelly)

Linear regression with a limited set of predictors  $P_i$  derived from the

NWP model:  $\text{Bias} = \sum \beta_i \cdot P_i(x)$

➤  $[\gamma, \delta]$  model: Radiative Transfert Model correction  
(for errors in absorbing gas density, SRF, absorption coefficient).

For each channel, definition of

$\delta$ : global constant

$\gamma$ : fractional error in layer absorption coefficient

Transmittance from level  $p$  to space:  $\Gamma(p) \rightarrow \Gamma(p)^\gamma$

Physically based scheme, discriminating observation bias from model error.



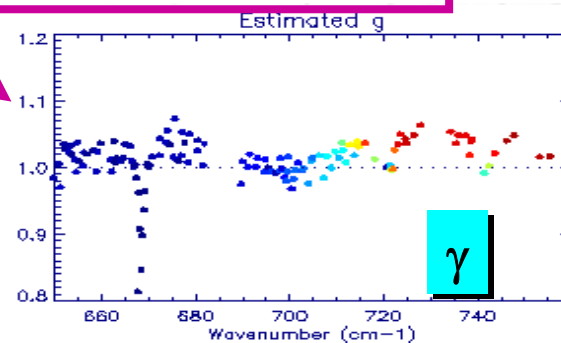
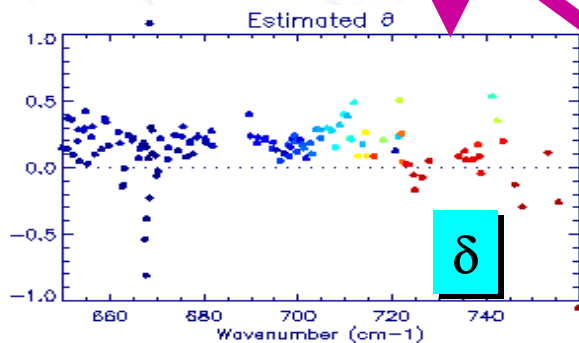
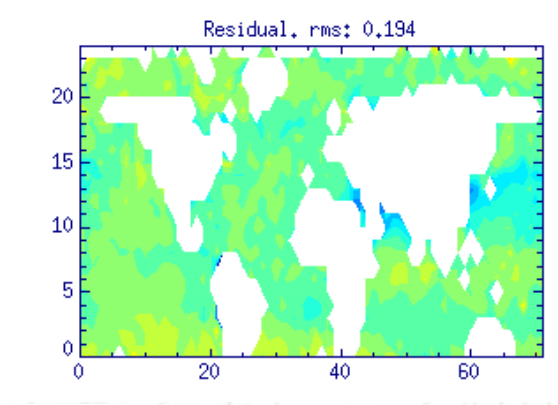
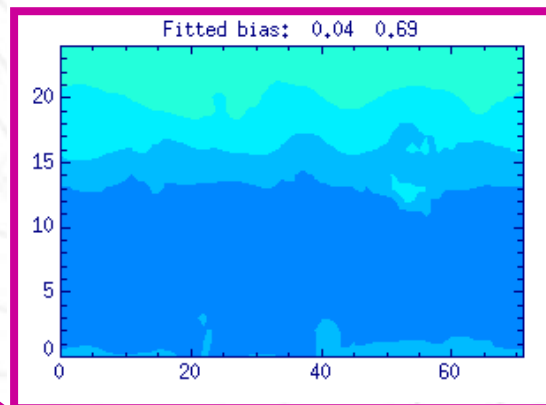
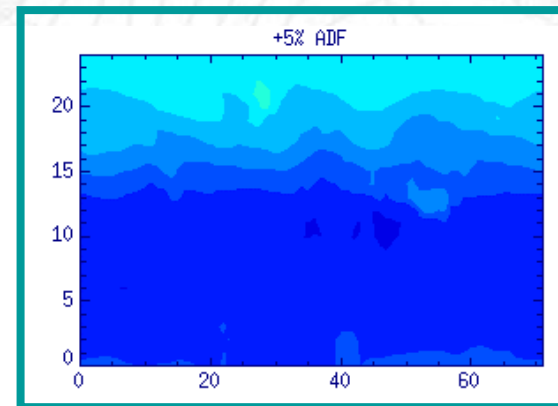
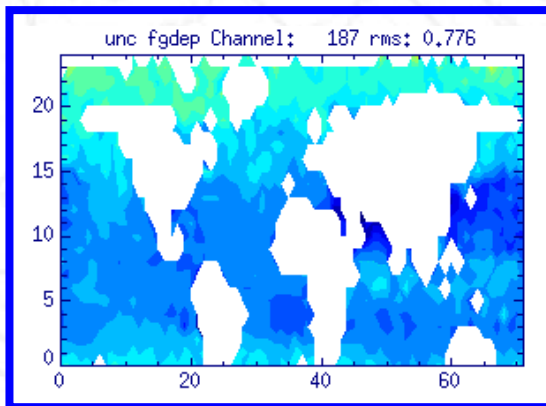
# Bias model: $[\gamma, \delta]$

Simulate  $\gamma = +5\%$   
transmission error – air-mass  
dependent bias:  $\mathbf{A}$

Monitor biases in operational  
System:  $\mathbf{B}$

Assume bias model:  
 $\mathbf{B} = \delta + \gamma \cdot \mathbf{A}$

Get best estimates of  $\delta$  and  $\gamma$



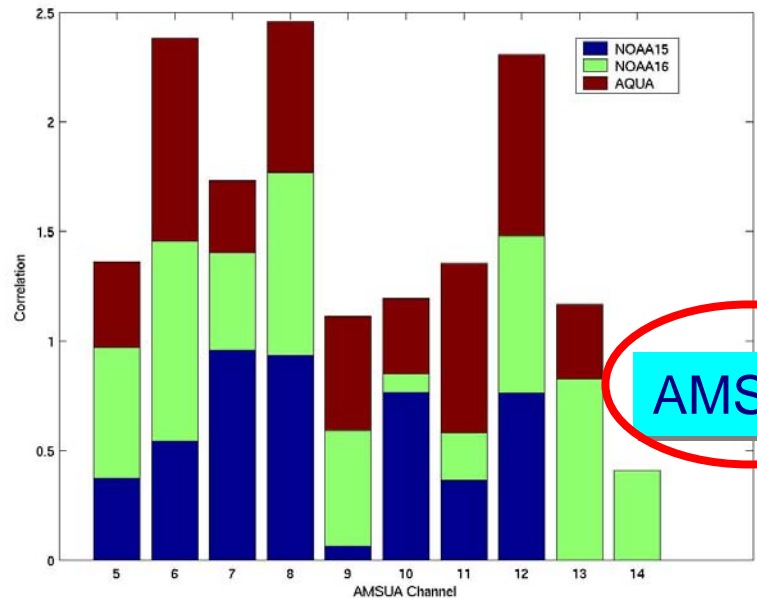
AIRS 15  $\mu\text{m}$  channels

Credits: P. Watts

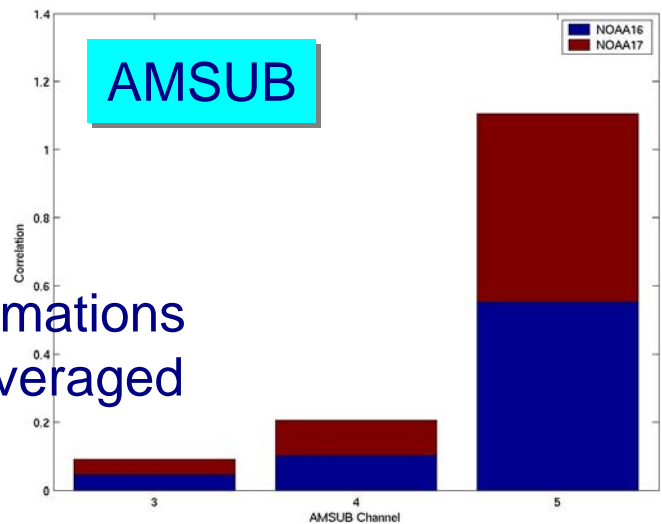


# Bias model: $[\gamma, \delta]$

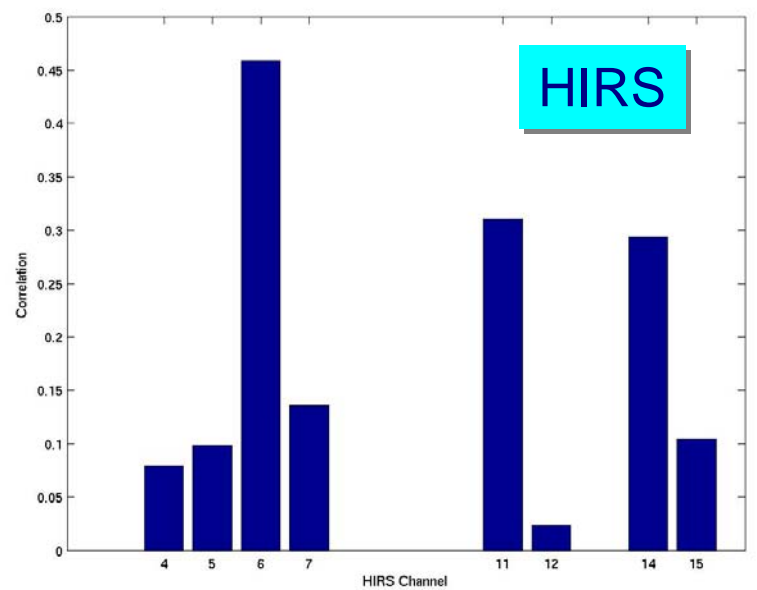
Correlations between bias estimations and departures averaged over one month



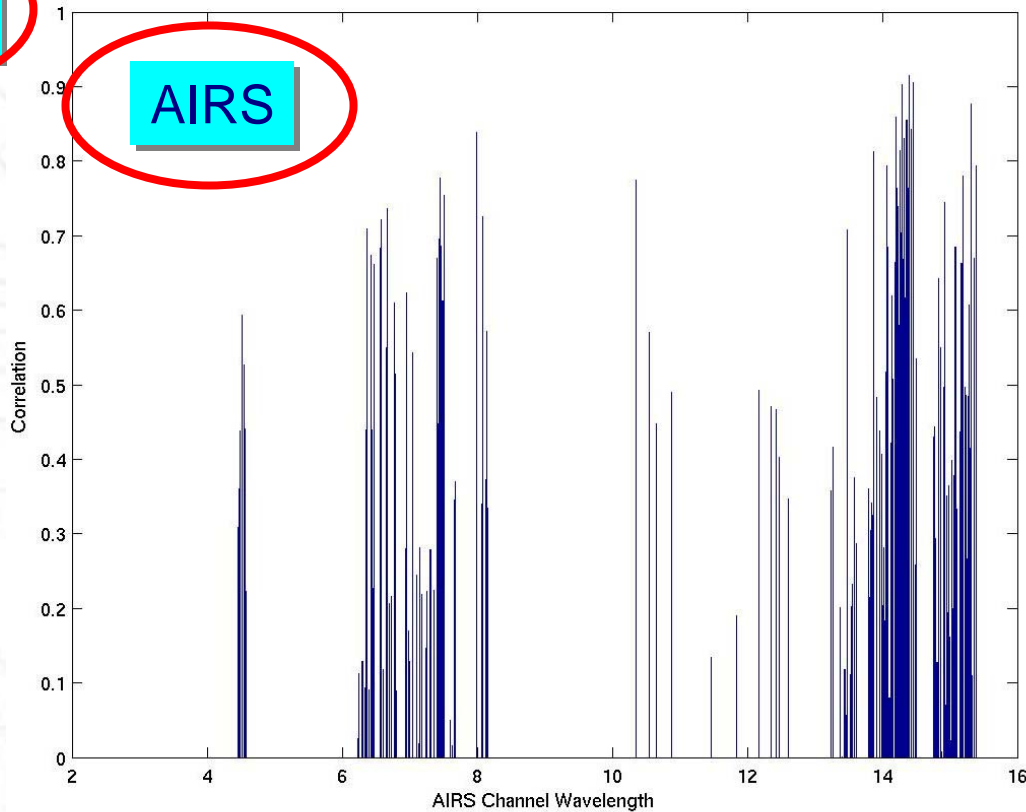
AMSUA



AMSUB



HIRS



AIRS





# Adaptive bias correction

A static bias correction cannot correct an instrument failure/drift.  
Problem of identifying manually a drift within hundreds of data types in real time.

Adaptive bias correction = bias estimate is updated for every cycle.

## ***Pros:***

Automatic, thus much easier to handle for new instruments or drifts.  
Continuity in time series (interesting for climate simulations).  
Based on the same bias model (Harris&Kelly).

## ***Cons:***

Prone to wrongly mapping systematic errors of the NWP model into radiance bias correction. Relies even more on the ability of the bias model to separate observation bias from model error.  
Need for a background term : reduces the reactivity of the system.

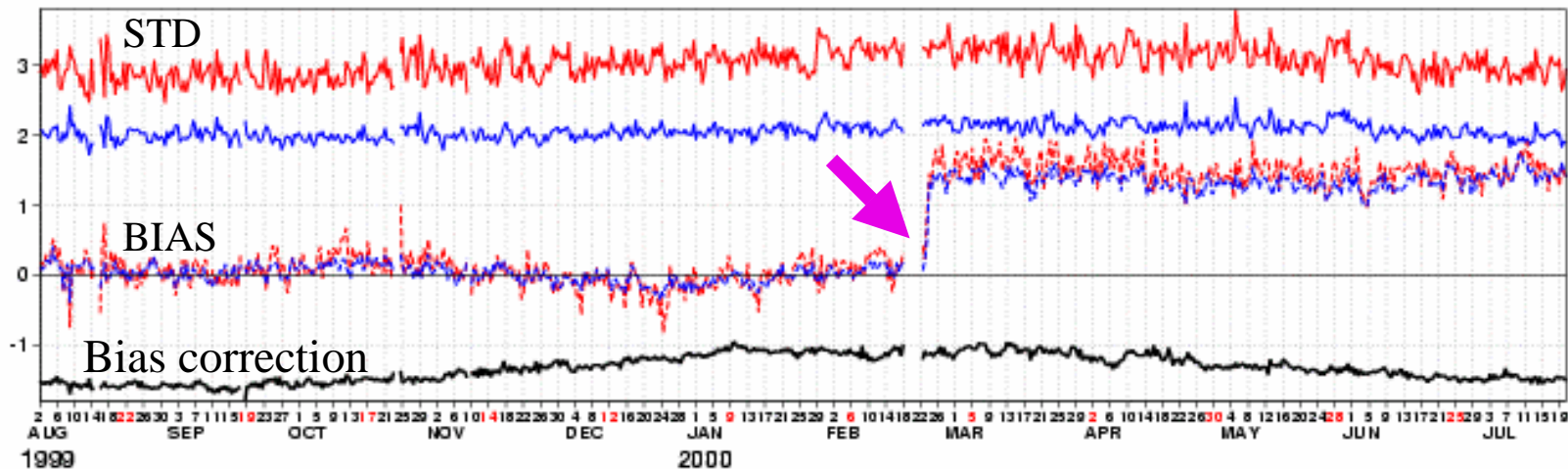


# Adaptive bias correction

Static

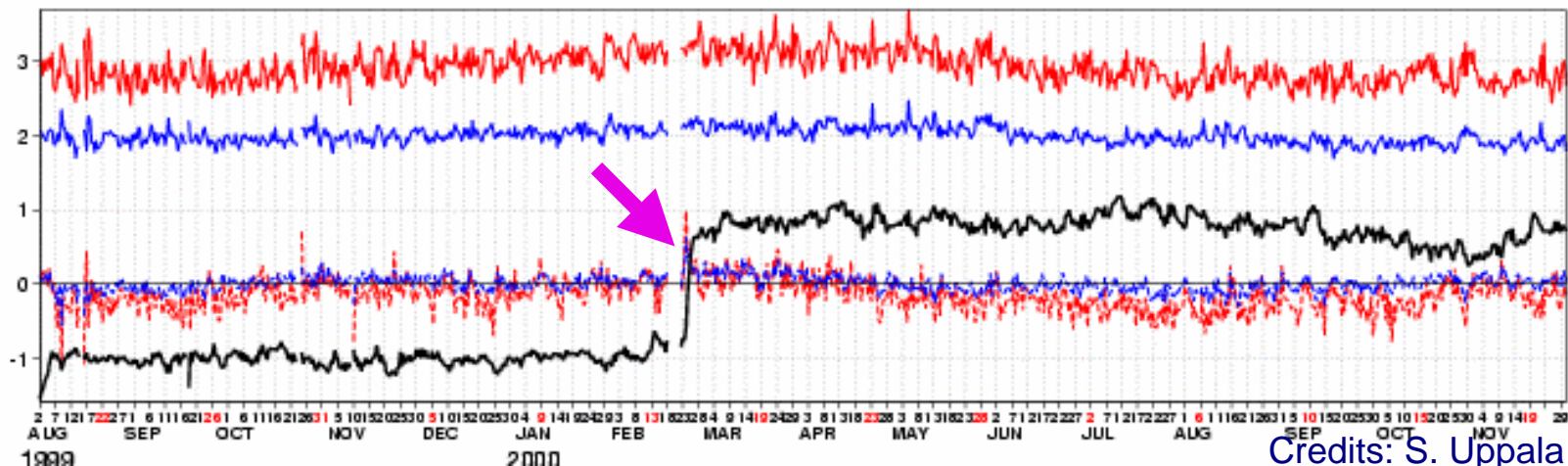
0469 (DA) : SSMI-1C\_dmsp-14\_SSMI\_Tb Ch 3 Southern Hemisphere  
St.dev. and bias (K) OB-FG (red) OB-AN (blue) BIASCOR (black)

First Guess departures  
Analysis departures



Adaptive

0470 (DA) : SSMI-1C\_dmsp-14\_SSMI\_Tb Ch 3 Southern Hemisphere  
St.dev. and bias (K) OB-FG (red) OB-AN (blue) BIASCOR (black)

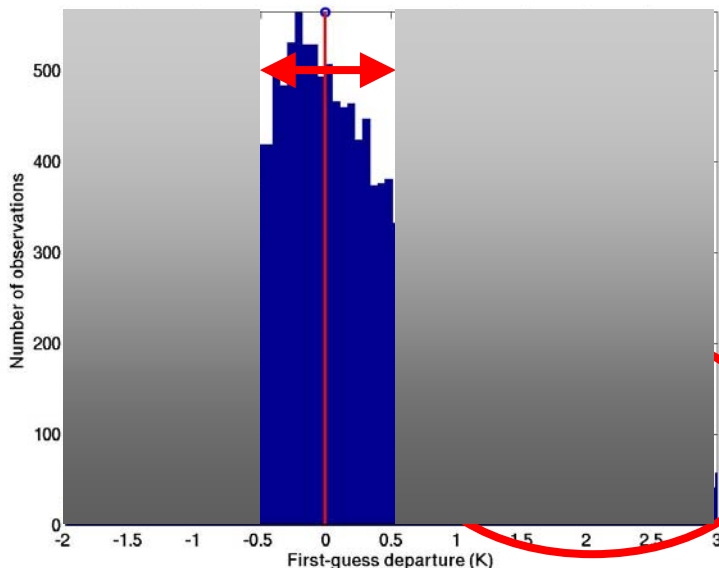


Credits: S. Uppala



# Adaptive bias correction (and quality control)

Aqua  
AMSUA-4  
over sea  
2005/03/01

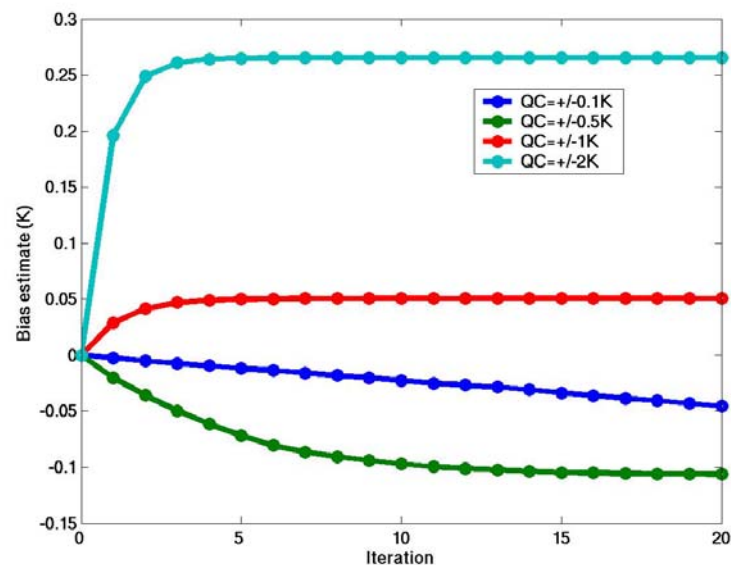


A typical distribution of departures has a cold / warm tail (IR / MW) due to cloud contamination. Quality Control based on departures is often applied to remove the tail and outliers (bad quality data) BEFORE estimating the bias.

Single (static bias correction) and successive (adaptive) applications lead to different estimates.

The value of the estimate and speed of convergence depend on the size of the boxcar window QC.

We are evaluating the use of the MODE for bias estimation as opposed to the mean.







# Variational bias correction

VarBC = adaptive bias correction INSIDE the assimilation system

bias parameters  $\beta_i$  (*i.e.* coefficients for the bias model) become part of the 4DVar control variable

→  $\underline{H}(x, \beta) = H(x) + \sum \beta_i \cdot P_i(x)$  (H: observation operator,  $P_i$ : predictors)

## **Cons:**

(Small) overhead of computer calculation during NWP assimilation.  
Data used for QC but not assimilated must go through minimisation to estimate the bias.

## **Pros:**

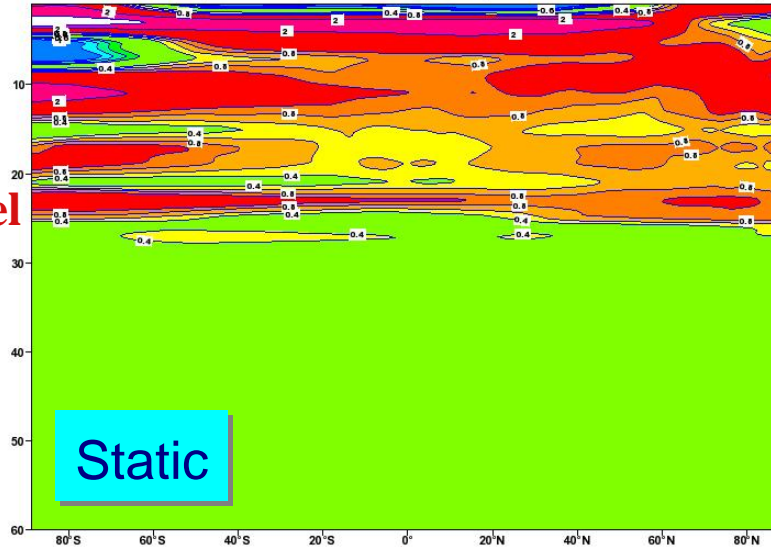
Estimation is constrained by other information inside the analysis, *i.e.* model, other data (Radiosondes, aircraft, surface, ...).



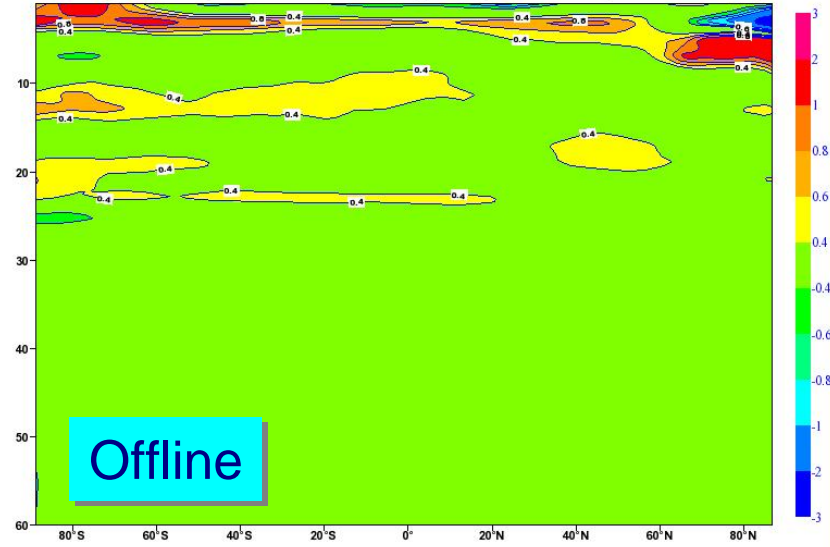
# Variational bias correction

Model  
level

Average of temp 20050228 2100 step 0 Expver emh3 (177.1W-177.1E)

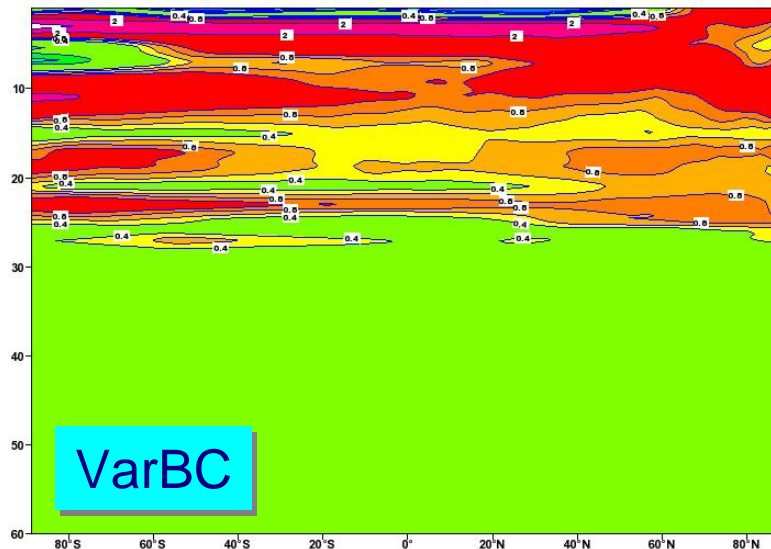


Average of temp 20050228 2100 step 0 Expver emh5 (177.1W-177.1E)



Latitude

Average of temp 20050228 2100 step 0 Expver emh4 (177.1W-177.1E)



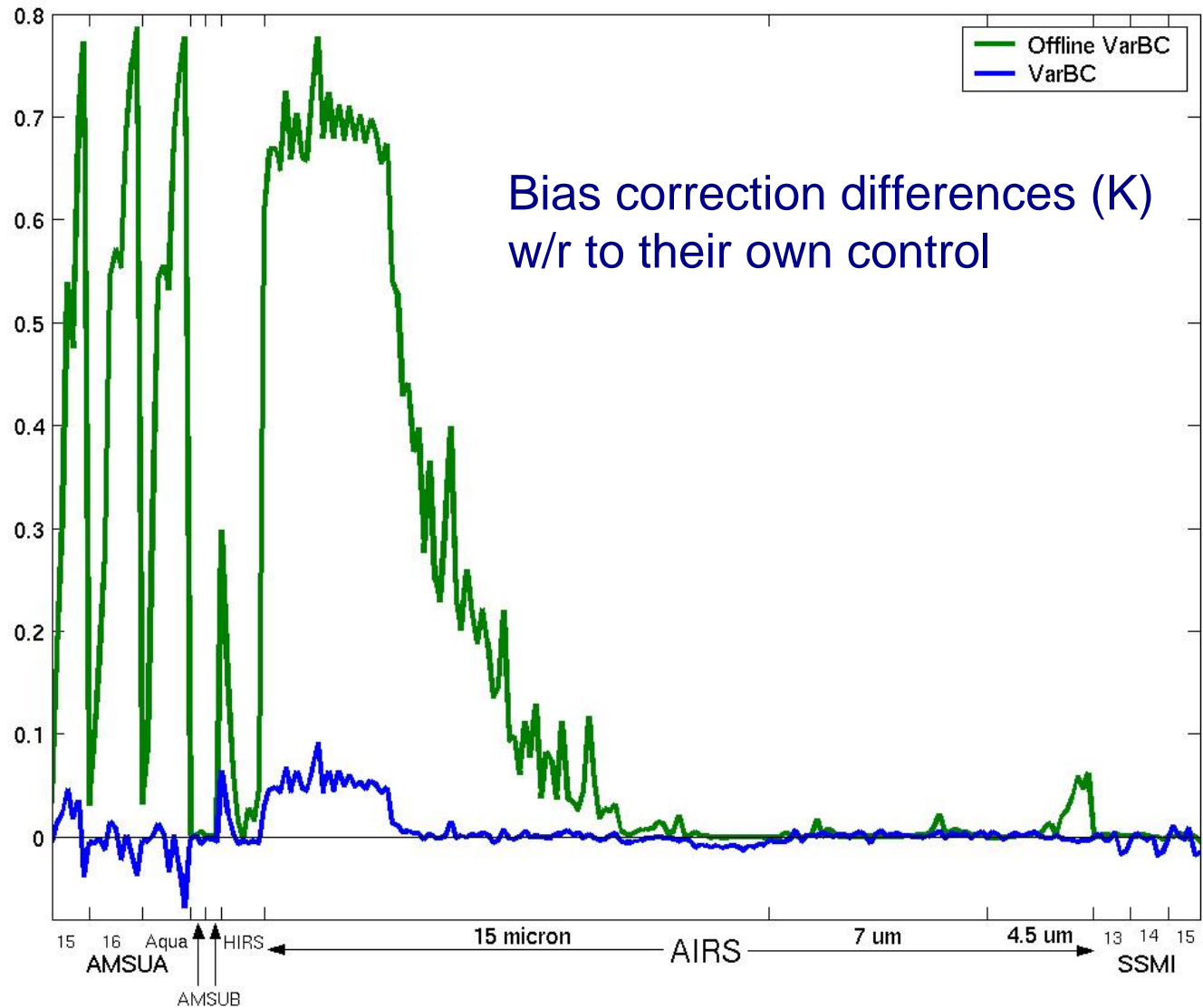
Flat artificial model perturbation:

-1K for levels 1 → 25 (above 100hPa)

Zonal means of temperature analysis differences for experiments with different bias correction schemes w/r to their own control (*i.e.* same scheme without perturbation)



# Variational bias correction





# Conclusion

$[\gamma, \delta]$  bias model used operationally for AIRS and AMSUA.

Technical and scientific advantages of adaptive bias correction.

Mapping of NWP error into bias estimate is greatly reduced with VarBC, due to the constraint of other data.

Feedback process b/w QC (first-guess check, cloud detection) defining the active population and adaptive bias correction modifying next cycle's departures.

Investigation on the use of the mode of the departures distribution as bias estimate.

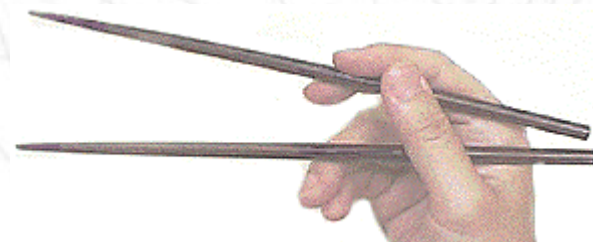
Potential benefit of GPS Radio Occultation data.





Thank you for your attention

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International TOVS Study Conference, 14<sup>th</sup>, ITSC-14, Beijing, China, 25-31 May 2005.  
Madison, WI, University of Wisconsin-Madison, Space Science and Engineering Center,  
Cooperative Institute for Meteorological Satellite Studies, 2005.