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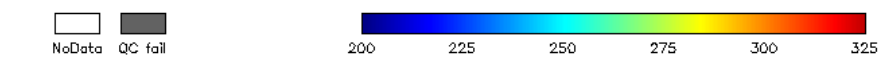
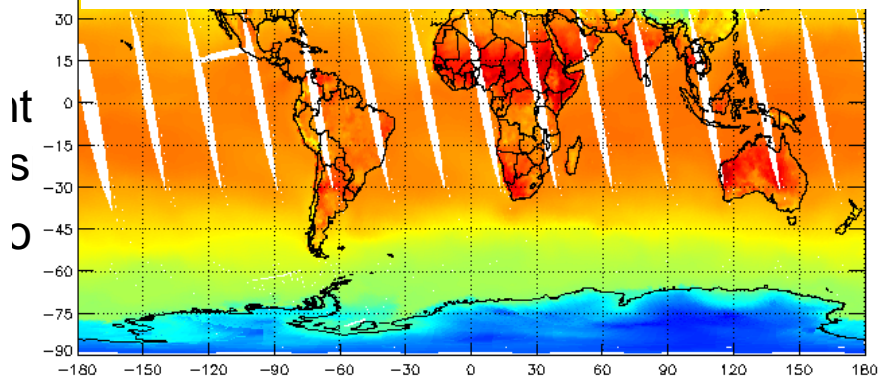
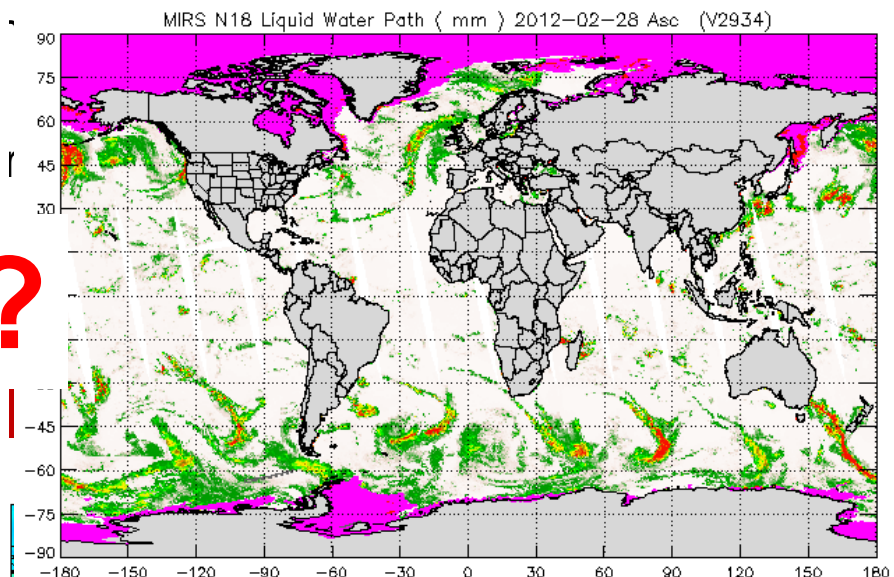
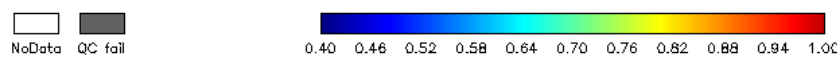
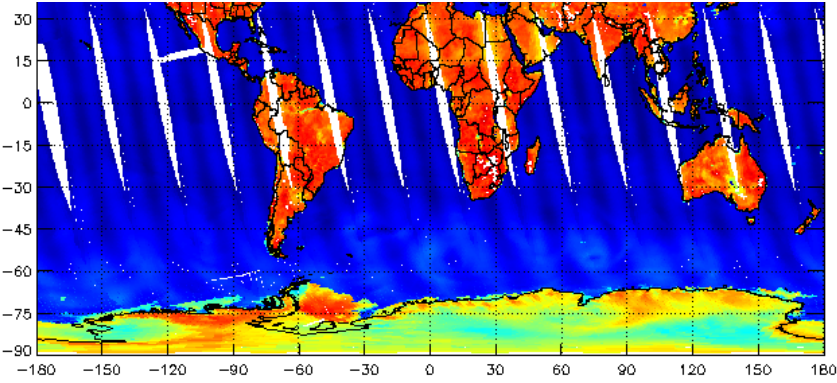
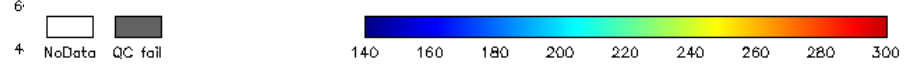
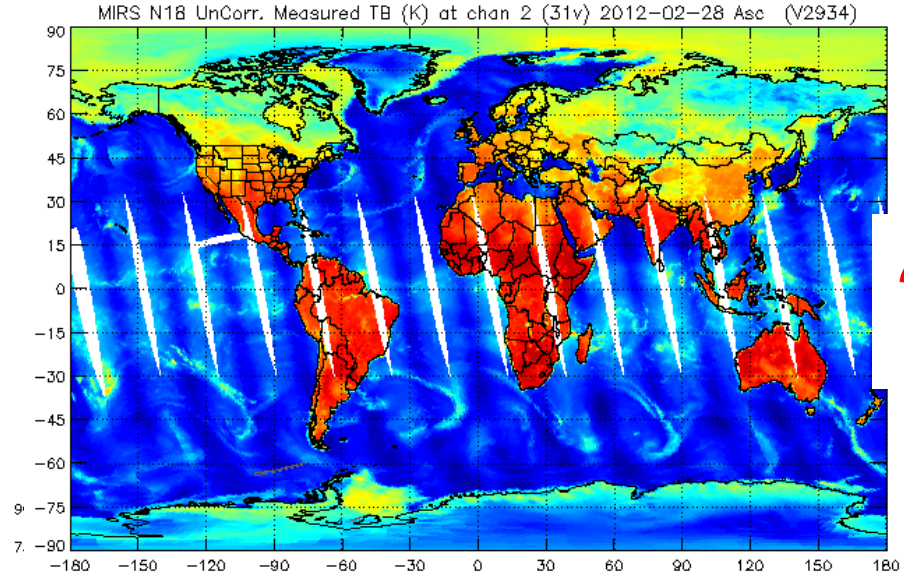
A Variational Approach to NWP Preprocessing and Quality Control

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18th International TOVS Study Conference
March 23, 2012

Toulouse, France

Motivation





Outline

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The 1DVAR Algorithm

2

Preprocessor – QC/Geophysical Fields

3

Radiance Assimilation Impact

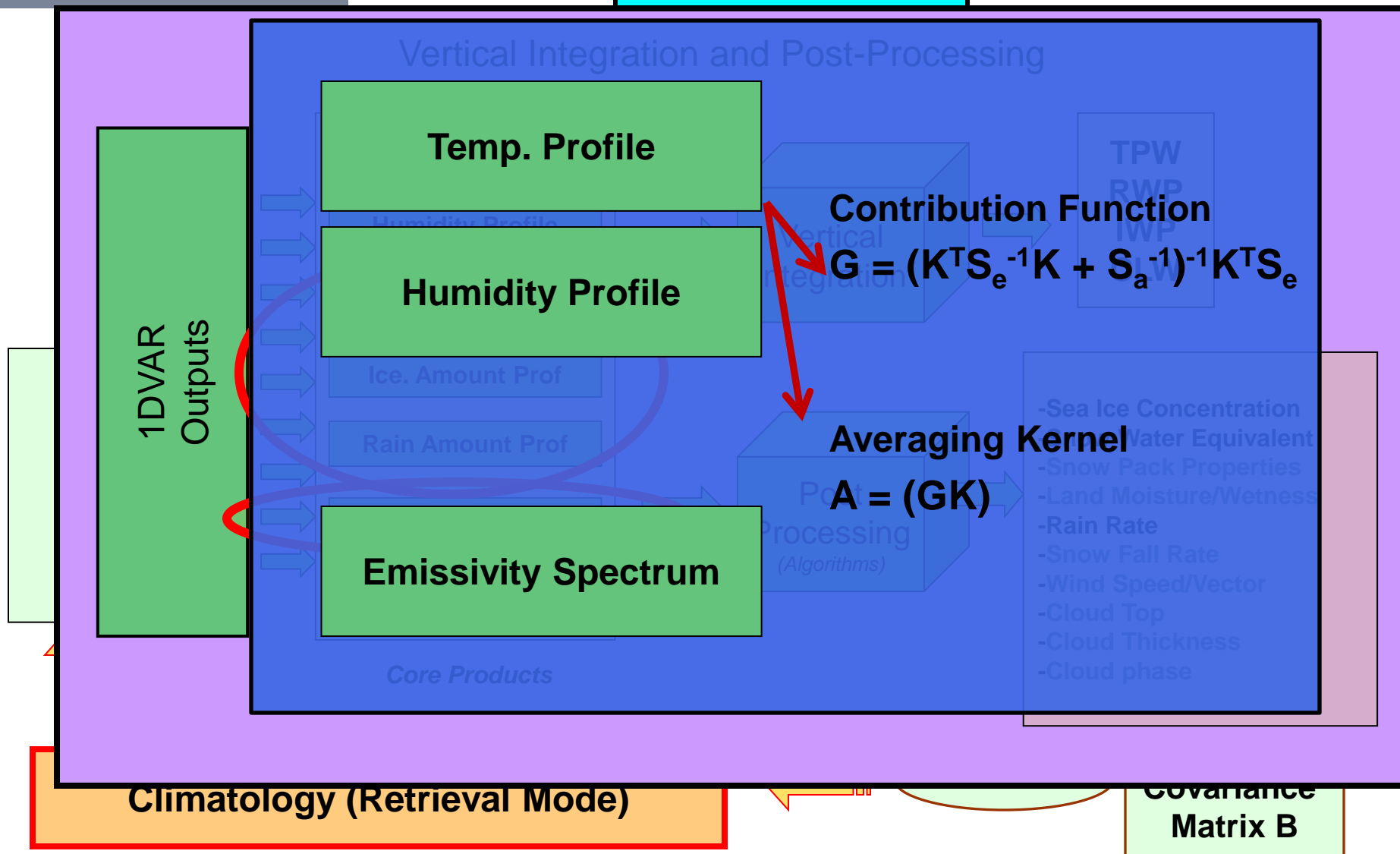
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Summary



- ❖ Physical algorithm (1DVAR) for microwave sensors (MiRS)
- ❖ MiRS applies to imagers, sounders, combination
- ❖ Cost to extend to new sensors greatly reduced
- ❖ MiRS uses the CRTM as forward operator (leverage)
- ❖ Applicable on all surfaces and in all-weather conditions
- ❖ Operational for N18, N19, Metop-A and F16/F18 SSMI/S
- ❖ **On-going / Future:**
 - Extension operations to Metop-B, NPP/ATMS and Megha-Tropiques (MADRAS and SAPHIR)
 - Get ready for the JPSS and GPM sensors.
 - **Extend MiRS to Infrared Remote Sensing (CRTM is already valid)**

The 1DVAR Algorithm





The 1DVAR Algorithm



Preprocessor – QC/Geophysical Fields



Radiance Assimilation Impact



Summary

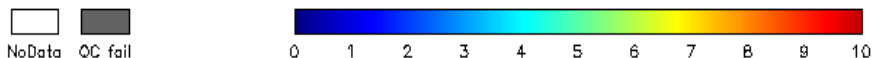
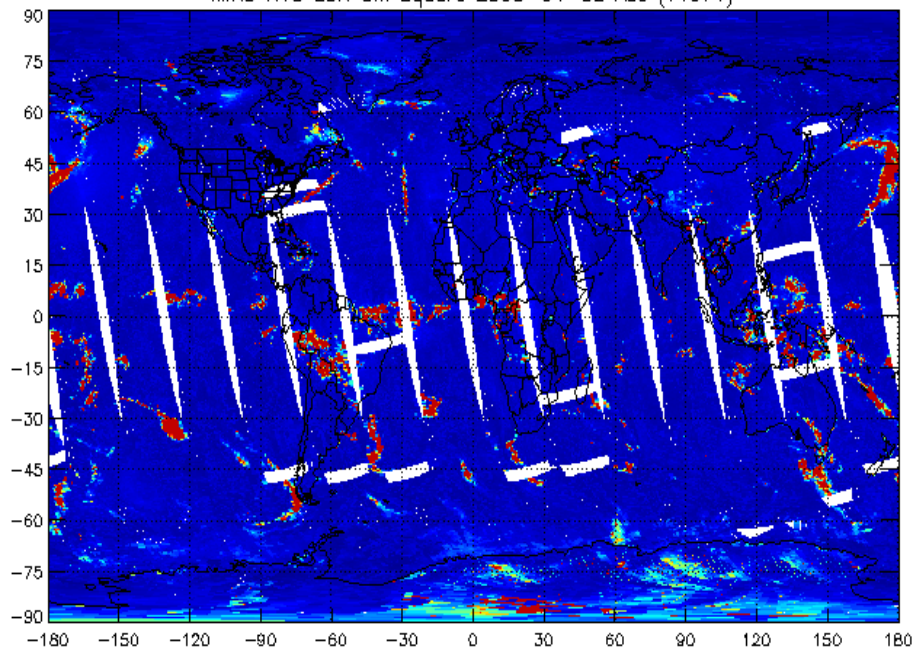
Preprocessor QC

- ❖ Convergence is reached everywhere: all surfaces, all weather conditions including precipitating, icy conditions
- ❖ A radiometric solution (whole state vector) is found even when precip/ice present. With CRTM physical constraints.

$$\phi^2 = (\mathbf{Y}^m - \mathbf{Y}(\mathbf{X}))^T \times \mathbf{E}^{-1} \times (\mathbf{Y}^m - \mathbf{Y}(\mathbf{X}))$$

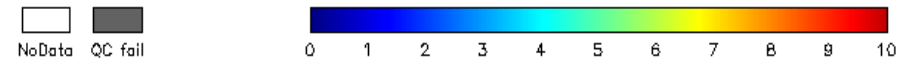
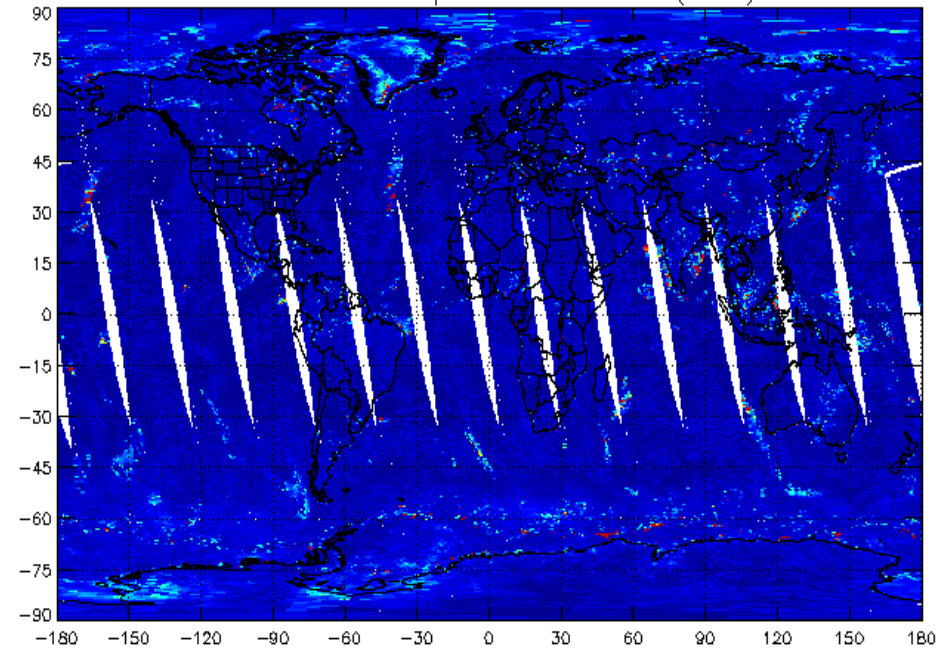
Previous version (1 attempt)
(assume non-scattering atmosphere)

MIRS N18 EDR Chi Square 2008-04-02 Asc (V1071)



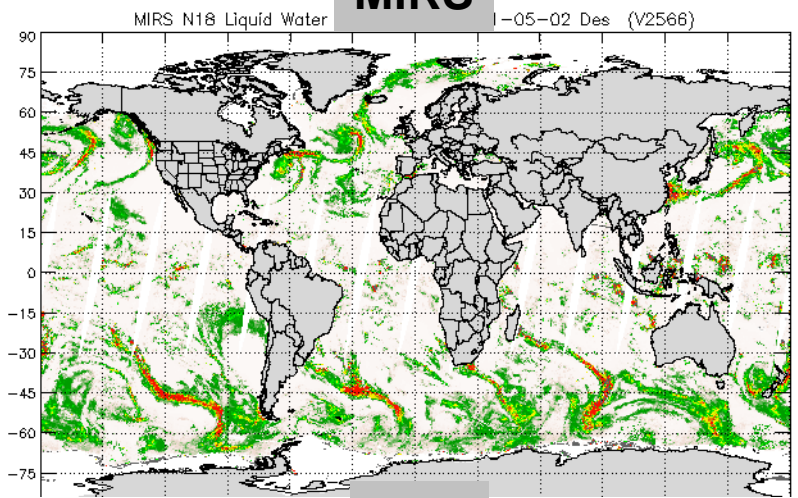
Current version (2 attempts)
(assume scattering from precip)

MIRS N18 EDR Chi Square 2008-06-08 Asc (V1316)

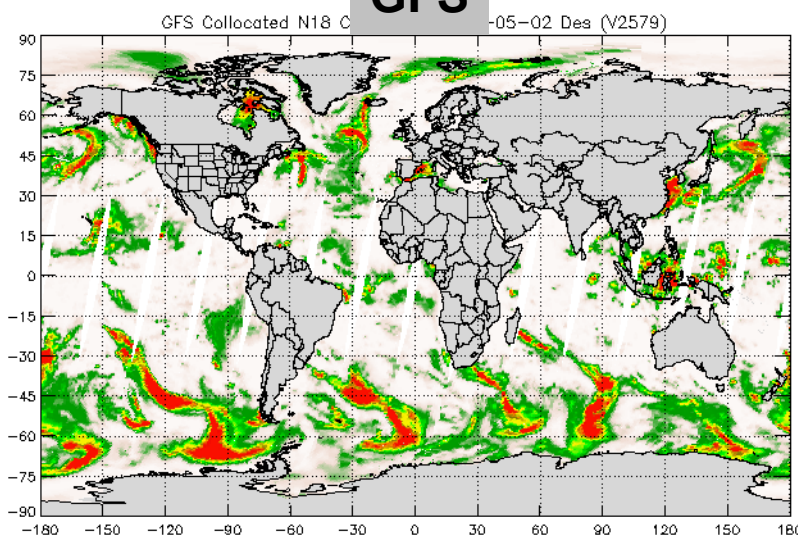


Preprocessor-CLW

MiRS

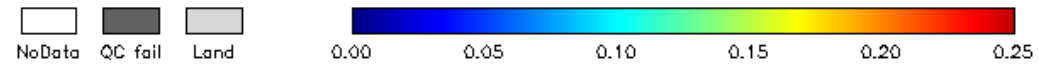
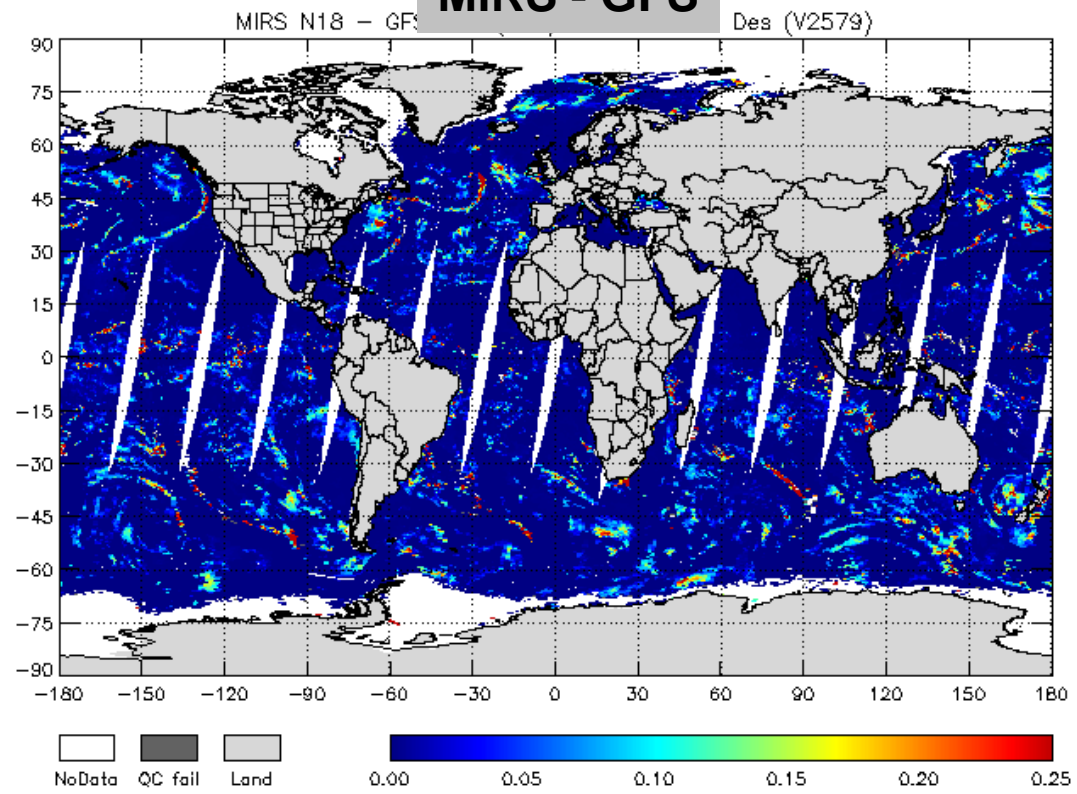


GFS



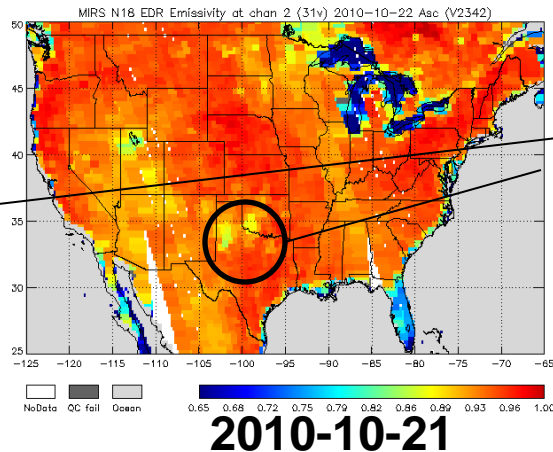
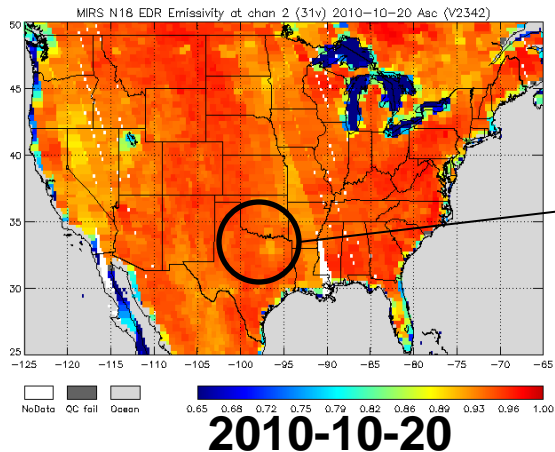
Cloud Liquid Water

MiRS - GFS

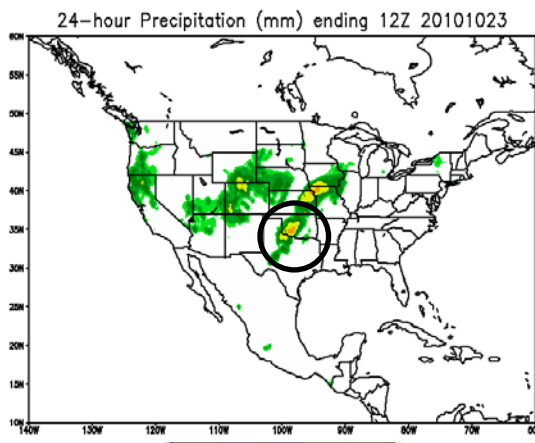
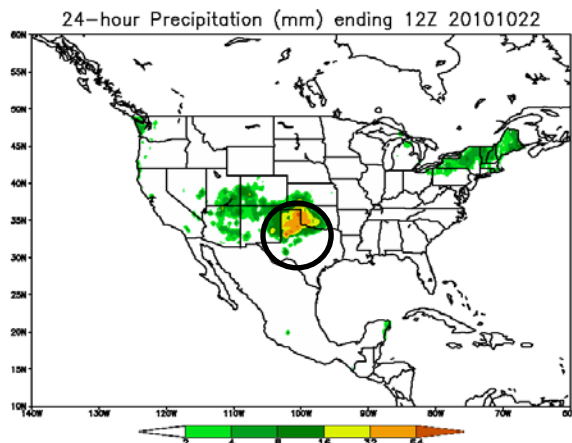
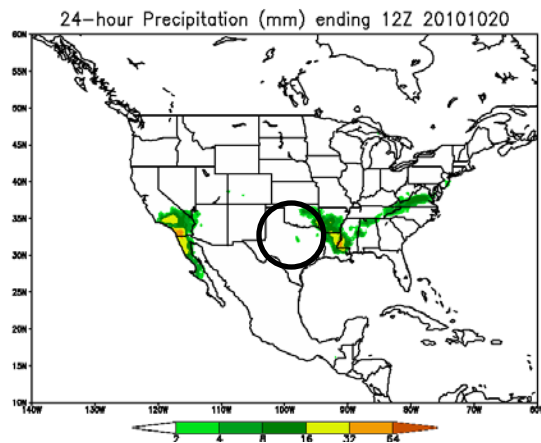
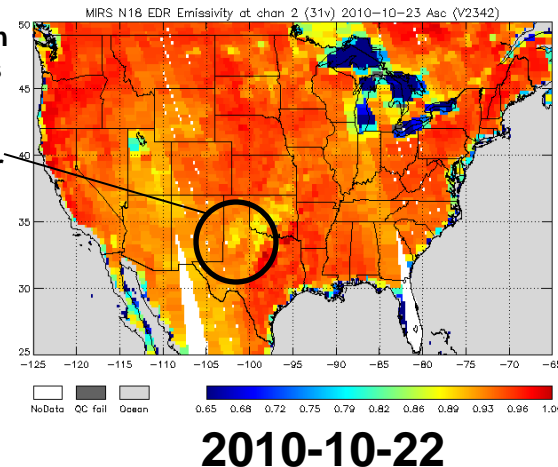


Preprocessor-Emissivity

MiRS N18 retrieved emissivity at 31 GHz ascending node



Evolution of emiss before, during and after rain



CPC real-time 24-hour precipitation



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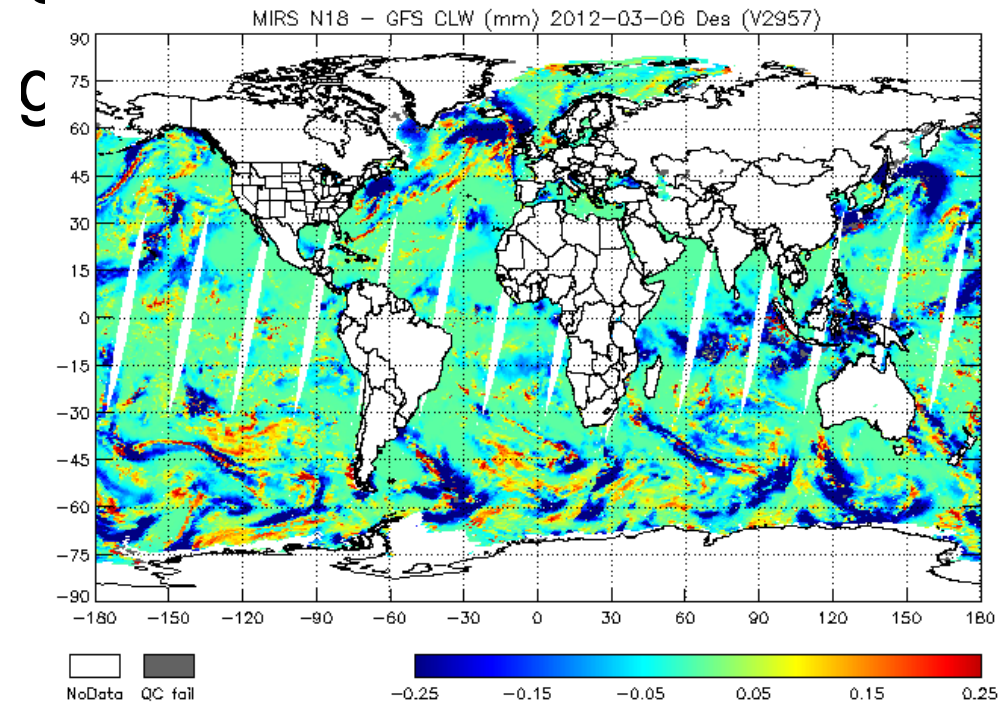
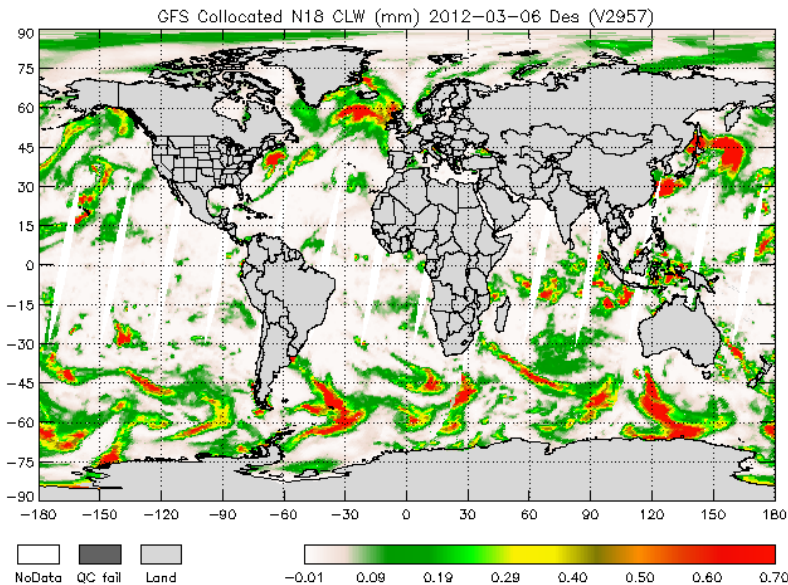
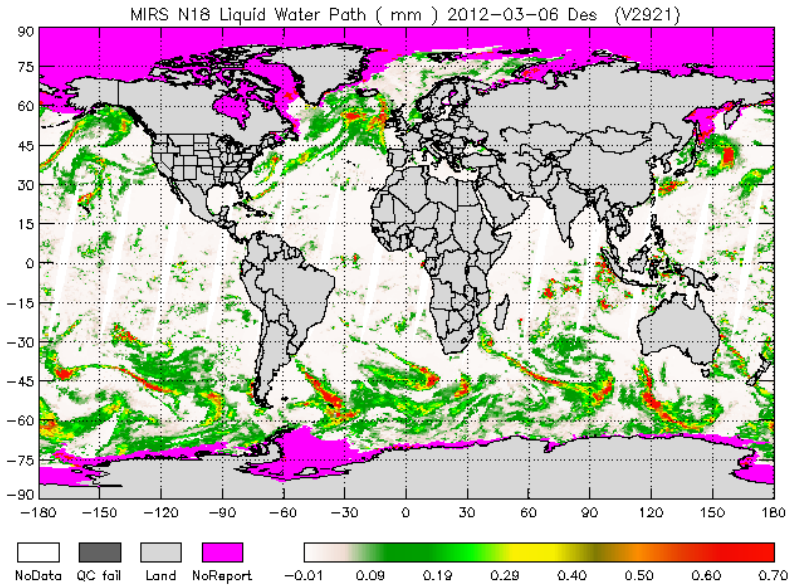
Radiance Assimilation Impact

4

Summary

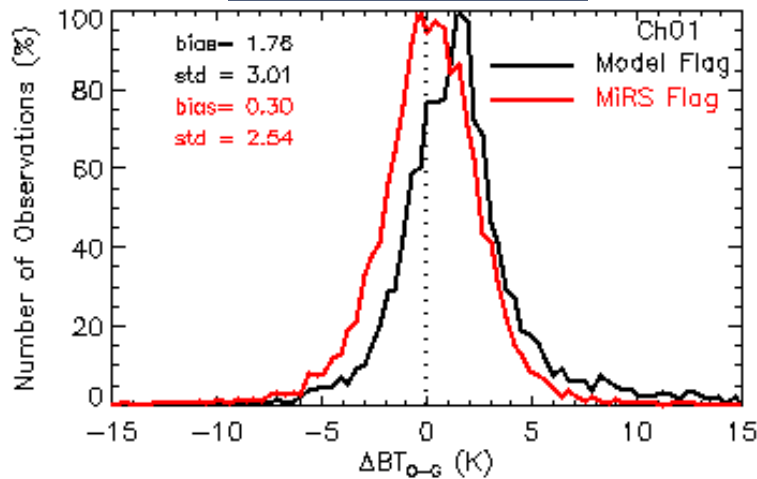
Assimilation Impact

Using MiRS retrieved CLW

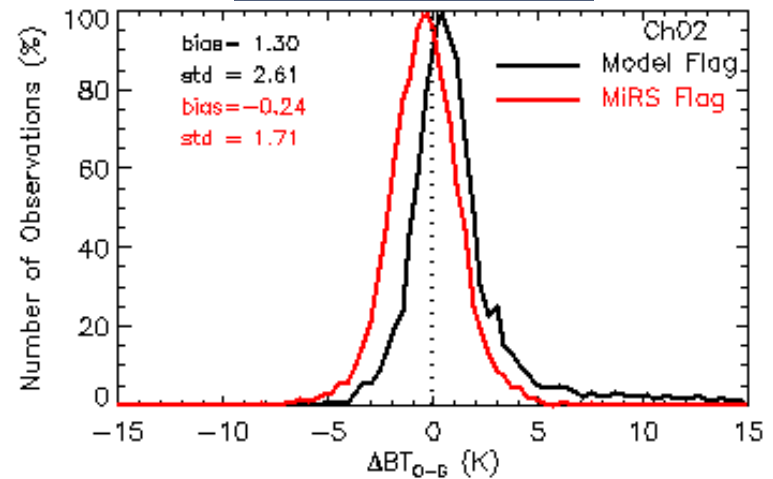


Assimilation Impact (O-B)

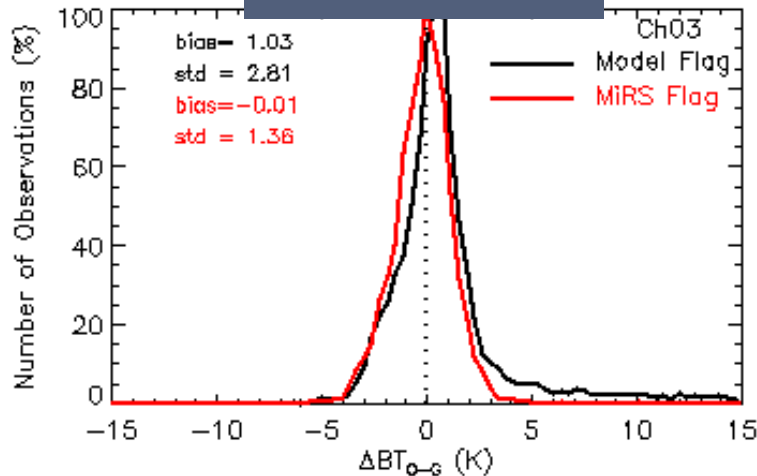
AMSU 23 GHz



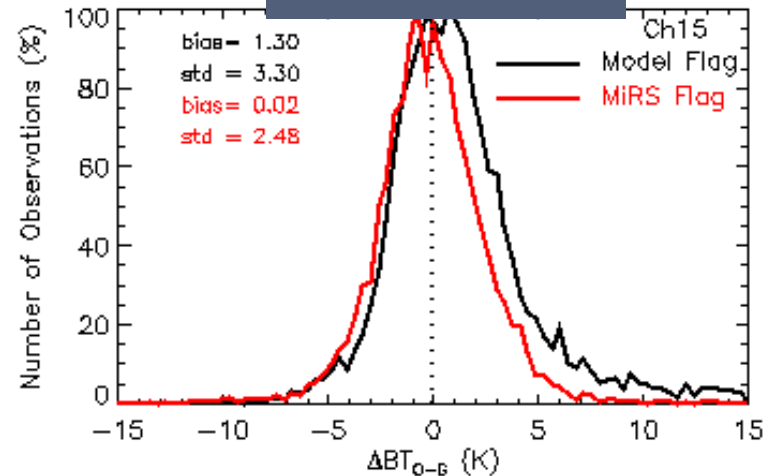
AMSU 31 GHz



AMSU 50 GHz

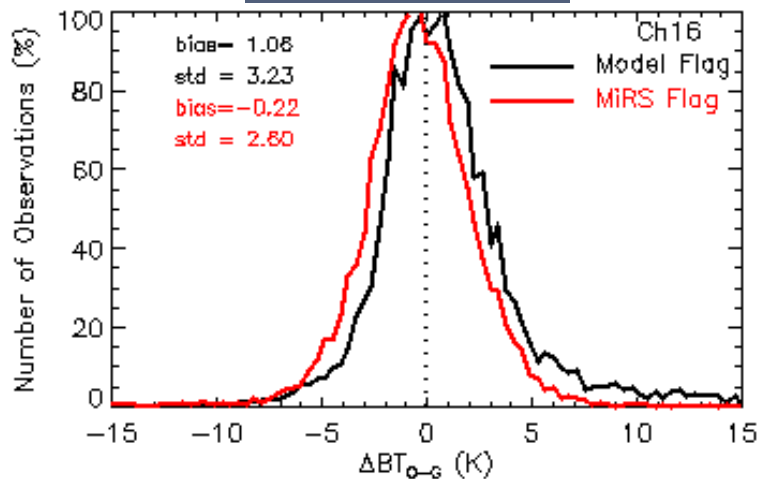


AMSU 89 GHz

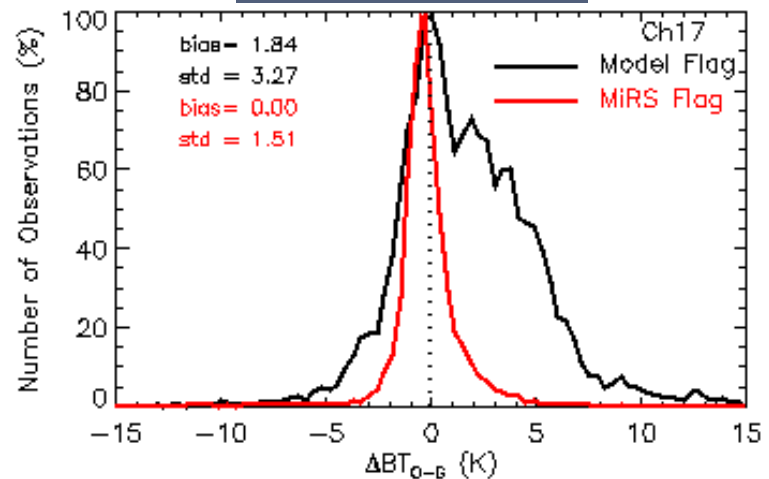


Assimilation Impact (O-B)

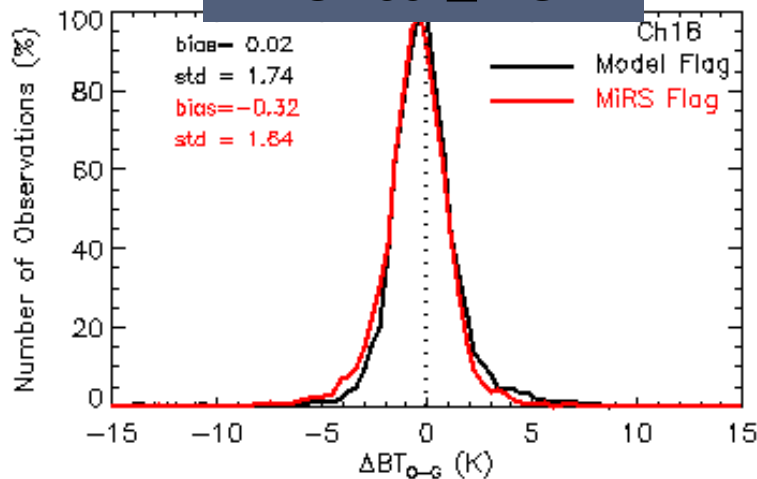
MHS 89 GHz



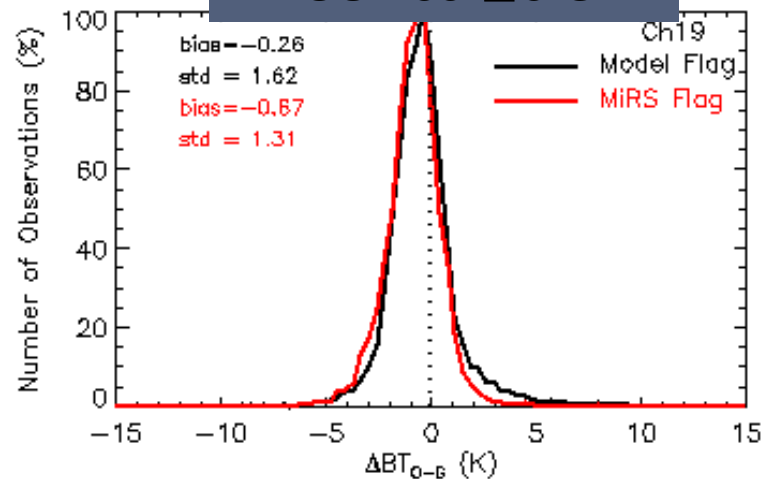
MHS 157 GHz



MHS 183 ±1 GHz



AMSU 183 ±3 GHz





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Impact on Radiance Assimilation

5

Summary



Summary

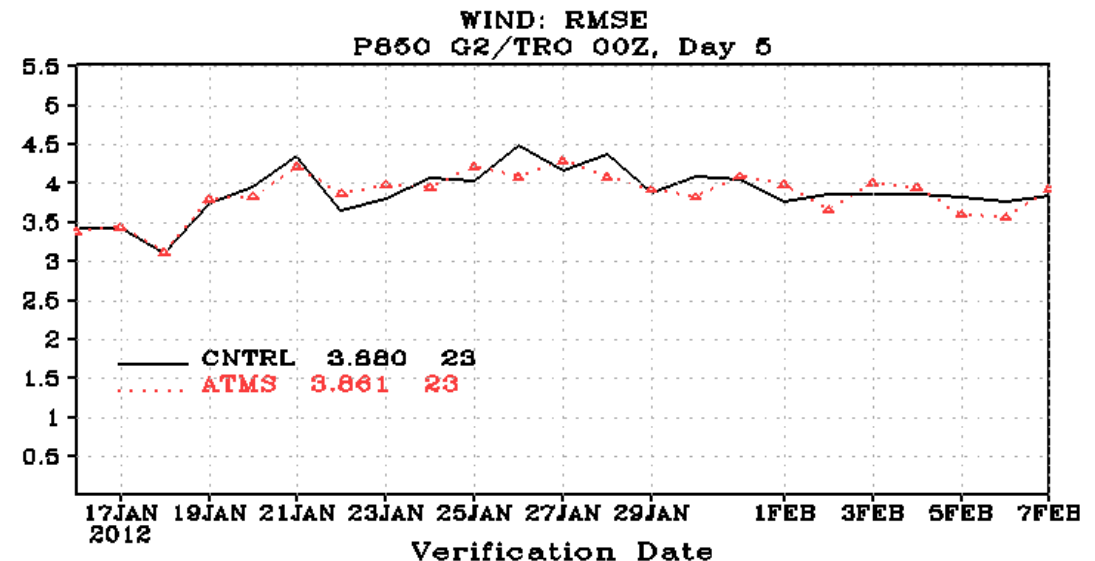
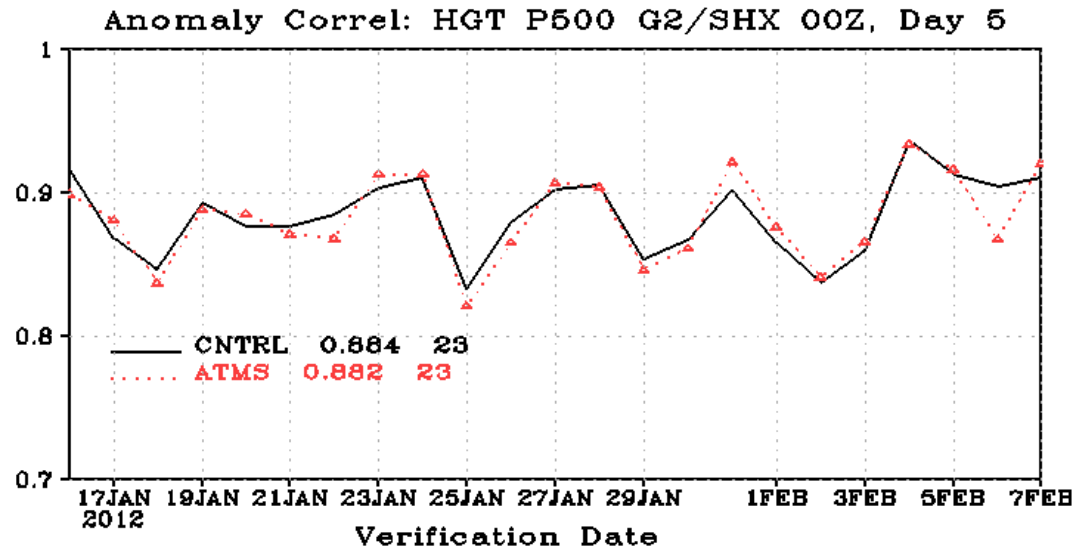
- ❖ MiRS is a generic retrieval/assimilation system (N18, N19, Metop-A, DMSP F16/18 SSMIS)
- ❖ In retrieval mode, MiRS can be used as an NWP (assimilation) preprocessor
- ❖ MiRS LWP has shown to produce reasonable O-B (improvement over using guess fields)
- ❖ Future work will include investigating the use of other QC metrics and retrieved fields to filter/parameterize DA
- ❖ For more detailed information about the MiRS project, visit: mirs.nesdis.noaa.gov (*more validation data, publication list and software package*)



BACKUP SLIDES

ATMS Impact Experiment

**GDAS (Hybrid ENKF)
high resolution
ATMS experiment**



Acknowledgements:

A. Collard

J. Jung

E. M. Devaliere

The 1DVAR Algorithm

❖ Cost Function to Minimize:

$$J(\mathbf{X}) = \left[\frac{1}{2} (\mathbf{X} - \mathbf{X}_0)^T \times \mathbf{B}^{-1} \times (\mathbf{X} - \mathbf{X}_0) \right] + \left[\text{Jacobians \& Radiance Simulation from Forward Operator: CRTM} \right]$$

Jacobians & Radiance Simulation from Forward Operator: CRTM

- ❖ To find the optimal solution, solve for: $\frac{\partial J(\mathbf{X})}{\partial \mathbf{X}} = \mathbf{J}'(\mathbf{X}) = 0$
- ❖ Assuming Linearity $y(\mathbf{x}) = y(\mathbf{x}_0) + \mathbf{K}[\mathbf{x} - \mathbf{x}_0]$
- ❖ This leads to iterative solution:

$$\mathbf{X}_{n+1}^{\Delta} = \left\{ \left(\mathbf{B}^{-1} + \mathbf{K}_n^T \mathbf{E}^{-1} \mathbf{K}_n \right)^{-1} \mathbf{K}_n^T \mathbf{E}^{-1} \right\} \left[\left(\mathbf{Y}^m - \mathbf{Y}(\mathbf{X}_n) \right) + \mathbf{K}_n \mathbf{X}_n^{\Delta} \right]$$

$$\mathbf{X}_{n+1}^{\Delta} = \left\{ \mathbf{B} \mathbf{K}_n^T \left(\mathbf{K}_n \mathbf{B} \mathbf{K}_n^T + \mathbf{E} \right)^{-1} \right\} \left[\left(\mathbf{Y}^m - \mathbf{Y}(\mathbf{X}_n) \right) + \mathbf{K}_n \mathbf{X}_n^{\Delta} \right]$$

More efficient
(1 inversion)

Preferred when $n\text{Chan} \ll n\text{Params}$ (MW)