

Leslie Moy<sup>1</sup>, Sid-Ahmed Boukabara<sup>2</sup>, Kevin Garrett<sup>1</sup>, Christopher Grassotti<sup>1</sup>, Flavio Iturbide-Sanchez<sup>1</sup>, Wanchun Chen<sup>3</sup> and Fuzhong Weng<sup>2</sup> 1. I.M. Systems Group at NOAA/NESDIS/Center for Satellite Applications and Research, Camp Springs, MD 20746. 2. NOAA/NESDIS/Center for Satellite Applications and Research, Camp Springs, MD 20746.

3. P.S.G.S. at NOAA/NESDIS/Center for Satellite Applications and Research, Camp Springs, MD 20746.



5. Current and Future Work

Emisivity at 23 GHz

May 4, 2007 (b

# 1. Introduction

## 3. Direct Assessment of Rain Rate

Comparisons to TRMM 2A12

The Microwave Integrated Retrieval System (MiRS) algorithm has been applied successfully in the past, and its modular design allows for a timely and efficient extension to the GPM and future sensors. MiRS is currently being applied operationally at NOAA to NOAA-18, 19, Metop-A, DMSP-F16 SSMIS, TRMM, and experimentally for AMSR-E (i.e a daily processing is in place but not in real-time). MiRS also runs for NPP/ATMS and GPM using proxy data. MiRS will expand to include other sensors in the GPM constellation.



## 2. Description of the Algorithm

NOAA/NESDIS/STAR has developed MiRS, a flexible, physical algorithm:

- · Can be applied to multiple microwave imagers and sounders.
- · 1DVAR approach using CRTM as forward and jacobian operator.

· Retrieves sounding and surface parameters simultaneously, including hydrometeor profiles, rainfall rate, and surface emissivity.

Applicable over all surfaces and in all-weather conditions.

· Run operationally at NOAA OSDPD (and integrated at NDE for NPP/JPSS future processing)



Schematic of the MiRS retrieval algorithm iterative process. The initial state vector is a regression algorithm applied on observed brightness the temperatures, but could also come from a climatological background or NWP model

To reach the iterative solution, the algorithm seeks to minimize the cost function

$$J(X) = \left[\frac{1}{2}\left(X - X_0\right)^T \times B^{-1} \times \left(X - X_0\right)\right] + \left[\frac{1}{2}\left(Y^m - Y(X)\right)^T \times E^{-1} \times \left(Y^m - Y(X)\right)\right]$$

where X in the 1st term on the right is the retrieved state vector, and the term itself represents the penalty for departing from the background  $X_0$ , weighted by the error covariance matrix B. The 2<sup>nd</sup> term represents the penalty for the simulated radiances Y departing from the observed radiances Y<sup>m</sup>, weighted by instrument and modeling errors E. This leads to the iterative solution

$$\Delta X_{n+1} = \left\{ BK_n^T \left( K_n BK_n^T + E \right)^{-1} \right\} \left[ \left( Y^m - Y(X_n) \right) + K_n \Delta X_n \right]$$

where  $\Delta X$  is the updated state vector at iteration n+1, and K is the matrix of Jacobians which contain the sensitivity of X (parameters to retrieve) to the radiances

Two retrieval attempts are possible for each scene. The first attempt assumes a clear or cloudy scene (radiometric signal is due to atmospheric emission). If the attempt is nonconvergent (Ym not fit by Y), precipitation is assumed, scattering is turned on and rain and ice water profiles are retrieved along with sounding and surface products.



## TPW Assessment : MiRS vs. ECMWF



similar to F16 SSMIS (noisy sensor) but less than those of F18 or N18. Metop (global coverage) This could be due to: Lack of WV sounding channels or more

noisiness in sensor Limited coverage (should be tropics only) Might indicate that bias needs fine tuning

### Emissivity Assessment : MiRS vs. ECMWF





Performances (2.4% standard deviation) are lower than SSMIS (around 1.6%).

Could be due to higher differences in MiRS/TMI Tskin (due to penetration depth issues at 10, 65 GHz)

# Emissivity MiRS vs. GDAS at a number of sites

6. Summary

