

Improving Cloud and Moisture Representation by Assimilating GOES Sounder Products into Analyses for NWP



Jordan J. Gerth

Author Affiliations at the University of Wisconsin – Madison:
 Cooperative Institute for Meteorological Satellite Studies (CIMSS)
 Space Science and Engineering Center (SSEC)
 Department of Atmospheric and Oceanic Sciences (AOS)

Experiment Design

Objective: Quantify NWP response to GOES-13 Sounder-adjusted moisture concentrations.

Three Advanced Research Weather Research and Forecast (WRF-ARW) simulations are run twice daily (00/12Z):

WRFX – Initial conditions and boundary conditions from previous (06/18Z) GFS run

WRFY – Initial conditions and boundary conditions from initial hour CRAS20MKX run

WRFZ – Initial conditions of previous (06/18Z) GFS run modified with GOES-13 Sounder retrievals and GFS boundary conditions

Dynamics	Non-Hydrostatic with Gravity Wave Drag
Cumulus Scheme	Kain-Fritsch
Microphysics Scheme	WSM Single-Moment 5-Class
PBL Scheme	Yonsei University
Land Surface Scheme	Noah 4-Layer LSM
Surface Layer Physics	Monin-Obukhov with heat and moisture surface fluxes
Long Wave Radiation	RRTM
Short Wave Radiation	Dudhia Scheme
Time-Integration Scheme	Runge-Kutta 3 rd Order
Damping	Rayleigh

Kain-Fritsch Convective Scheme

The WRF simulations in this experiment utilize the Kain-Fritsch (KF) convective parameterization, which –is a mass flux scheme –requires an adjusted response based on the grid scaling

The closure for the KF scheme is convective available potential energy (CAPE). This is an important source for latent heat release and accumulated convective precipitation.

It has been shown in Kain and Fritsch (1990) that the normalized vertical mass flux varies significantly by a factor of two in the upper troposphere for changes of relative humidity between 50% and 90%.

This sensitivity is critical because, for cold temperatures, the amount of water vapor mixing ratio to adjust to the relative humidity is not particularly substantial.

Assimilating Three-Layer Precipitable Water from GOES

r_t = GOES total precipitable water
 r_b = Background total precipitable water
 $w(\sigma)$ = background mixing ratio
 $w(\sigma')$ = background mixing ratio perturbation
 w_s = surface mixing ratio
 $w(T)$ = saturation mixing ratio
 $w(\sigma'')$ = final mixing ratio
 Let,
 $r_t = \int_{p_s}^{p_t} w(\sigma) d\sigma$

Precipitable water is defined as: $r = \int_{p_s}^{p_t} w(\sigma) d\sigma$

Define a mean mixing ratio profile: $w(\sigma) = w(\sigma') + w(\sigma'')$ such that $\int_{p_s}^{p_t} w(\sigma) d\sigma$ is a minimum and $1.0 < \lambda < 3.5$ following Smith, 1966.

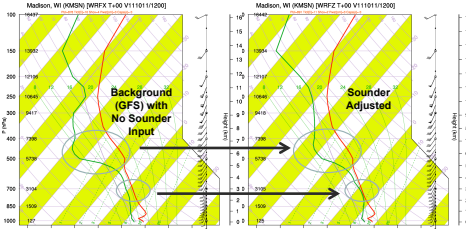
Solve for $\lambda = \lambda'$ such that:
 $r_t = \int_{p_s}^{p_t} w(\sigma') + w(\sigma'') d\sigma$ with $[\lambda w(\sigma') + w(\sigma'')] < w(T)$

The final adjusted mixing ratio is:
 $w(\sigma) = \lambda w(\sigma') + w(\sigma'')$

GOES-13 Sounder Moisture Correction

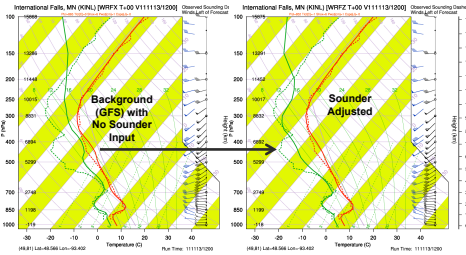
Madison, WI; 11 October 2011, 12 UTC

This example shows how moisture is added to the background analysis ahead of approaching precipitation while the distribution is maintained.

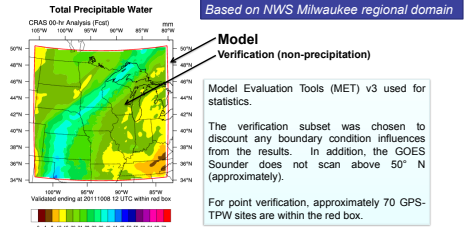


International Falls, MN; 13 November 2011, 12 UTC

This example shows the improvement to the background (left) by the GOES Sounder retrieval (right), compared to a radiosonde (dashed).

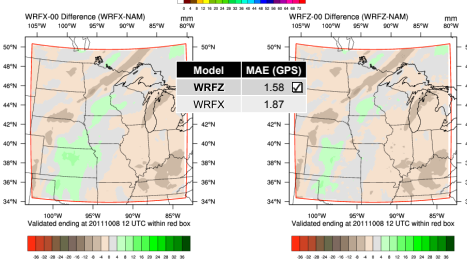
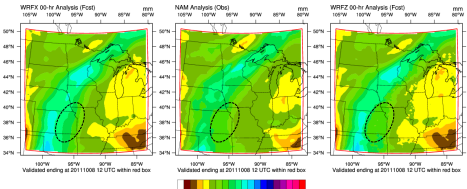


Experiment Domain

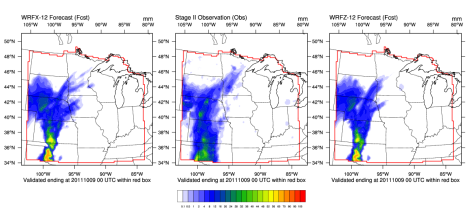


PW Analysis: WRFX vs. WRFZ Comparison for 8 October 2011, 12 UTC

WRFX started with PW up to 8 mm too moist over eastern Kansas, whereas the WRFZ exhibited less bias.



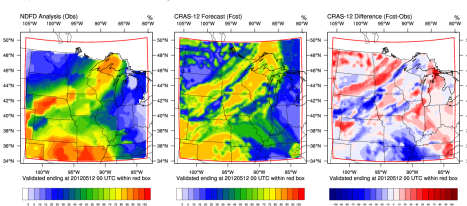
WRFX and WRFZ Precipitation Comparison with Stage II
 The WRFX and WRFZ produced more precipitation than observed over south central Kansas. However, the WRFZ was a drier solution compared to the WRFX. The WRFX and WRFZ predicted similar precipitation coverage patterns.



Using Linear Programming to Optimize Sky Cover Output

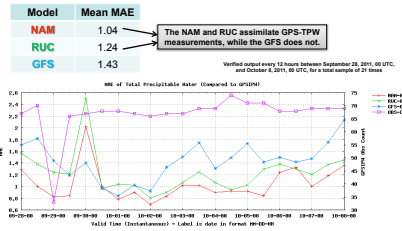
The process employed for staging and using background data was as follows:
 -Initialize the dynamical model at time $t = -12$ (hours).
 -Obtain the 12-hour forecast from the model initialized at $t = -12$.
 -Obtain the analysis from $t = 0$.
 -Run linear program to minimize the objective function comparing the 12-hour forecast and analysis both valid at $t = 0$.
 -Initialize the dynamical model at time $t = 0$.
 -Apply fixed variable values from the linear program (a coefficient and scalar) to dynamical model output at $t = 12, t = 24$, and $t = 36$. Calculate skill.

For this case, the absolute error (post-execution objective function value) from the linear program was 127075.8 (%) using the CRAS (mean absolute error of 13.0% per grid point) and 138651.3 (%) using the WRFX (mean absolute error of 14.1% per grid point).

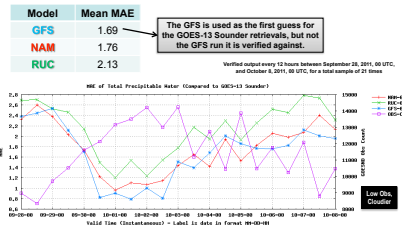


Experiment Results

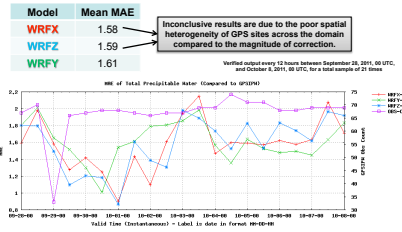
Total PW Mean Absolute Error Analyses verified against GPS-TPW



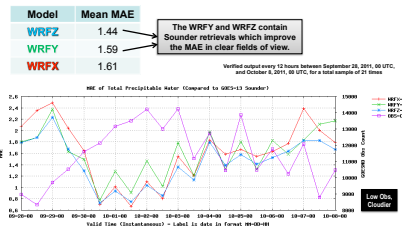
Total PW Mean Absolute Error Analyses verified against GOES-13 Sounder (Ma retrievals)



Total PW Mean Absolute Error Analyses verified against GPS-TPW



Total PW Mean Absolute Error Analyses verified against GOES-13 Sounder (Ma retrievals)



Total PW Mean Absolute Error Forecasts verified against GPS-TPW

