



Satellite Data Assimilation in the GEOS Atmospheric Data Assimilation System

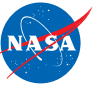
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February 26, 2024

Acknowledgements to: Yanqiu Zhu, Amal El Akkraoui, Ming-Jeong Kim, Wei Gu (NASA GMAO) and Will McCarty (NASA HQ).

GMAO

Global Modeling and Assimilation Office
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Outline

- Brief overview of Goddard Earth Observing System – Atmospheric Data Assimilation System (GEOS-ADAS)
- Microwave radiance assimilation in the GEOS
 - Assimilate AMSR2 data in all-sky conditions
 - Assimilate new ATMS-NOAA21 data
- Explore better assimilation of geostationary infrared observations
- Summary and on-going work



GEOS hybrid 4D-EnVar atmospheric data assimilation system

- Cost function:

$$J(\delta x) = \frac{1}{2} \delta x^T \mathbf{B} \delta x + \frac{1}{2} (y - y^o)^T \mathbf{R}^{-1} (y - y^o) + J_c$$

where,

δx , four dimensional correction (increment) to a background state x^b ;

\mathbf{B} and \mathbf{R} represent the four dimensional extensions of the background and observation error covariance matrices, respectively,

y^o , observations,

$y = \mathbf{H} (x^b + \delta x)$ is the observation-space equivalent of the model state interpolated using the nonlinear observation operator \mathbf{H} ,

J_c encompasses additional constraints such as the dry air conservation imposed on the analysis.

In the hybrid 4D-EnVar, the background error covariance \mathbf{B} is the sum of climatological covariance estimates \mathbf{B}_c and localized ensemble-based covariances \mathbf{B}_e :

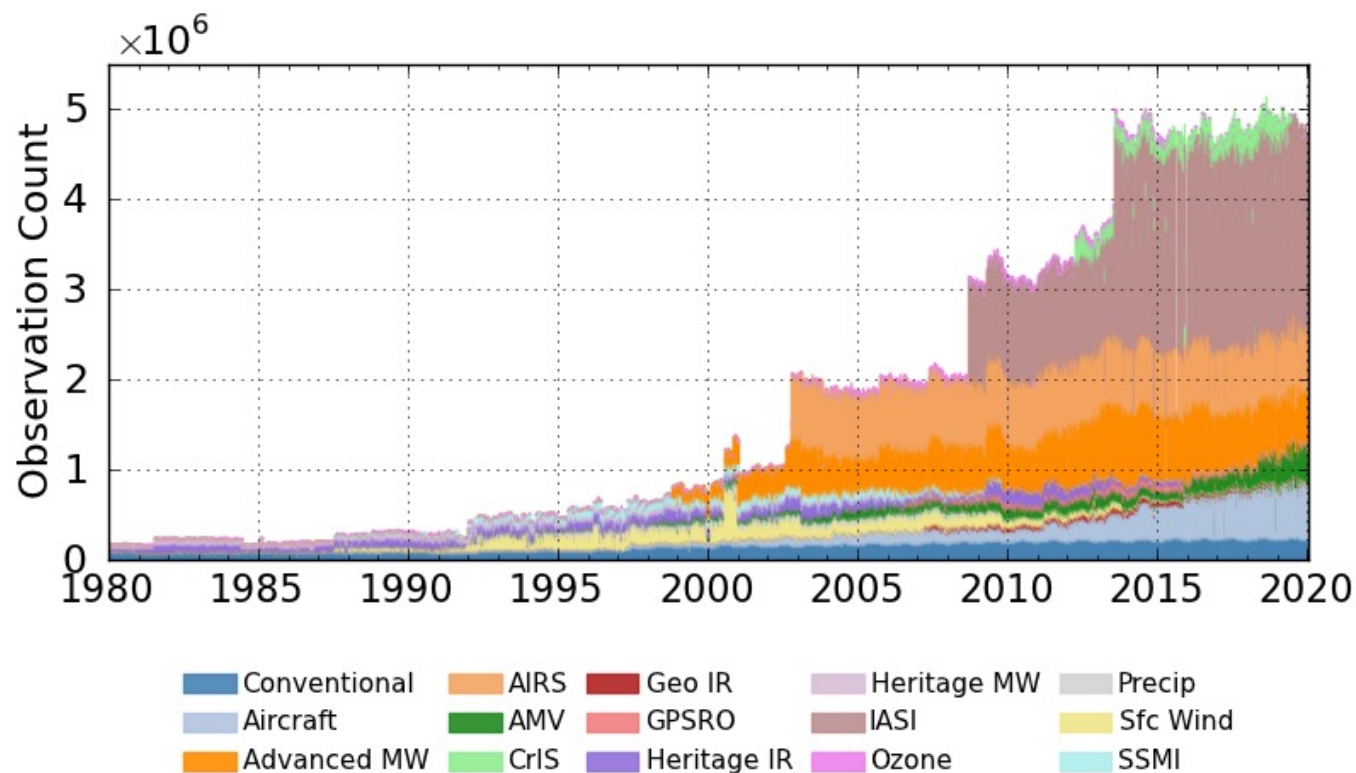
$$\mathbf{B} = \beta_c^2 \mathbf{B}_c + \beta_e^2 \mathbf{B}_e$$

- The near-real-time GEOS analysis is conducted with 32 ensemble members and 1 central analysis which rely on about 50-km horizontal atmospheric states. All atmospheric states are made by the nonhydrostatic cubed-sphere general circulation model.

(Todling and El Akkraoui, 2018; Akkraoui A. E., et al., 2023)

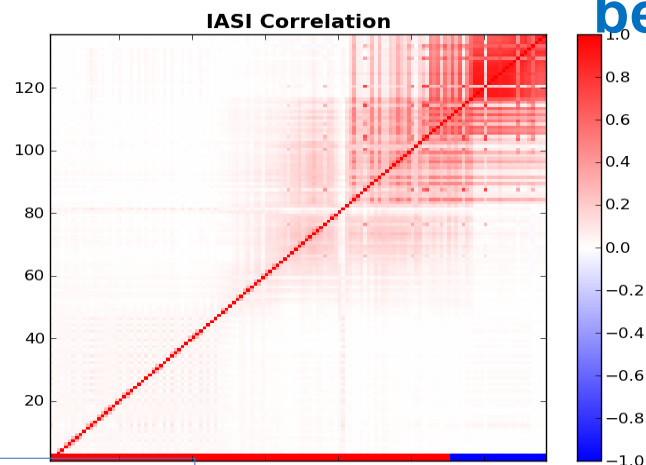


Satellite and conventional observations in MERRA2



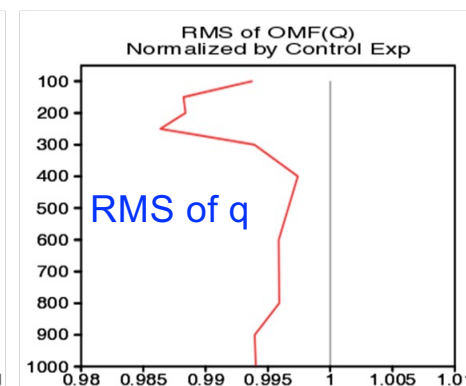
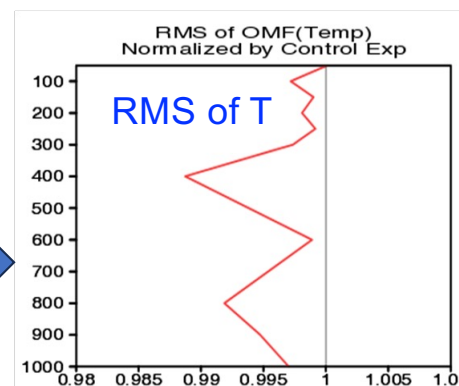
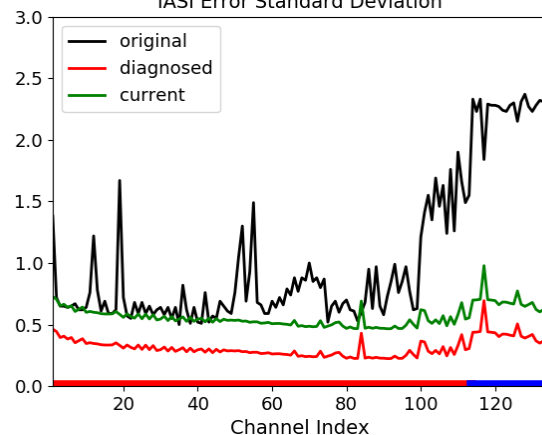
- New data types in GEOS ADAS
 - IASI Metop-C;
 - CRIS-FSR on Suomi-NPP and NOAA 20;
 - GPM/GMI, AMSR2/GCOM-W1;
 - Commercial GNSS-RO data.

Impact of considering correlated observational errors between infrared channels



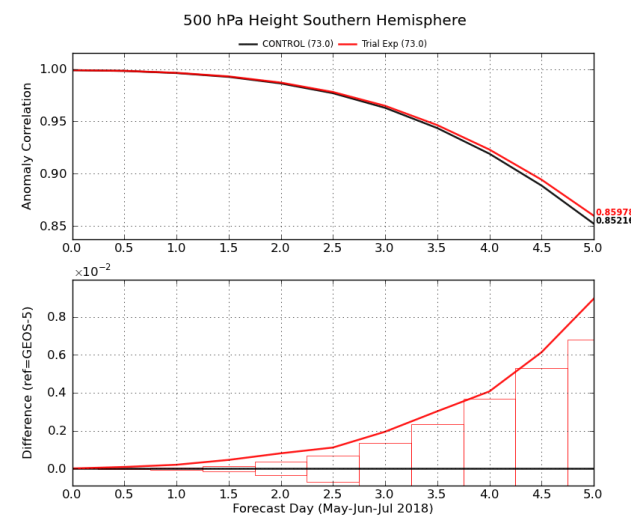
LW Temp

Window



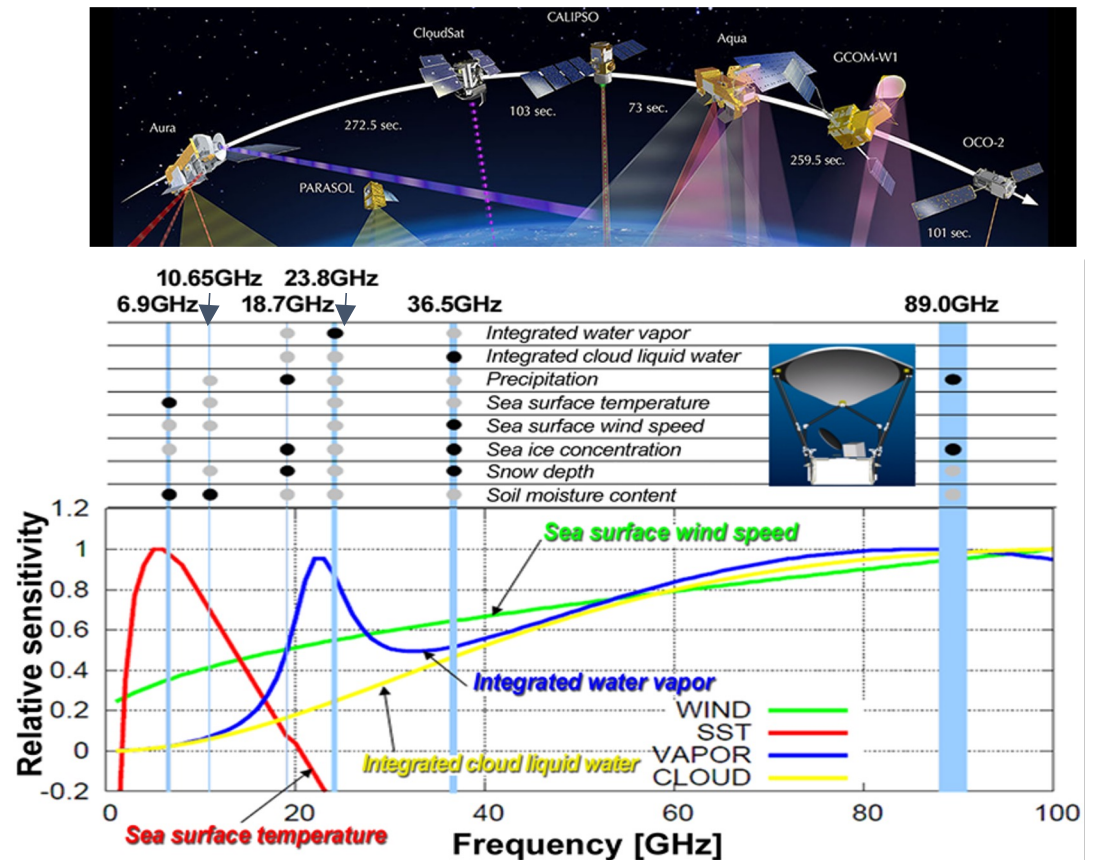
Reduced the RMS of OmFs for rawinsonde T and Q

Improved the overall forecast skills of GEOS



All-sky microwave radiance data assimilation in GEOS

- Advanced Microwave Scanning Radiometer 2 / Global Change Observation Mission 1st – Water
 - Launched in 2012
 - Part of A-Train satellite constellation
 - Lower frequency channels (ch1 - 6, 6.9 - 10.7 GHz) good for sea surface temperature (SST).
 - Ch 7 - 14 (18.7 - 89 GHz) are good for moisture, cloud and rain detection.
- GMAO assimilates its brightness temperature data over ocean.
- Cloud hydrometeors state and control variables** (liquid, ice, rain and snow clouds).
- Cloud analysis increments are updated the 2nd outer loop but are not added in model simulations in the next cycle.

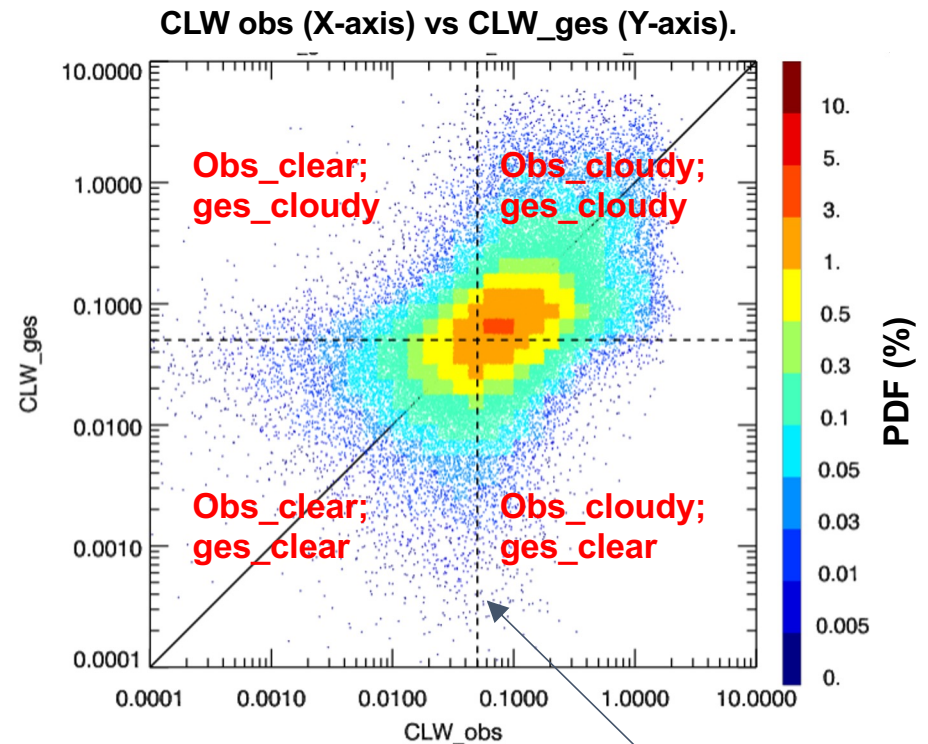


Identify observed and forecasted / guess clouds

Observed clouds (CLW_obs) are retrieved from **observed** brightness temperature (Tb) data, using the retrieval algorithm developed by K. Garrett for AMSR2.

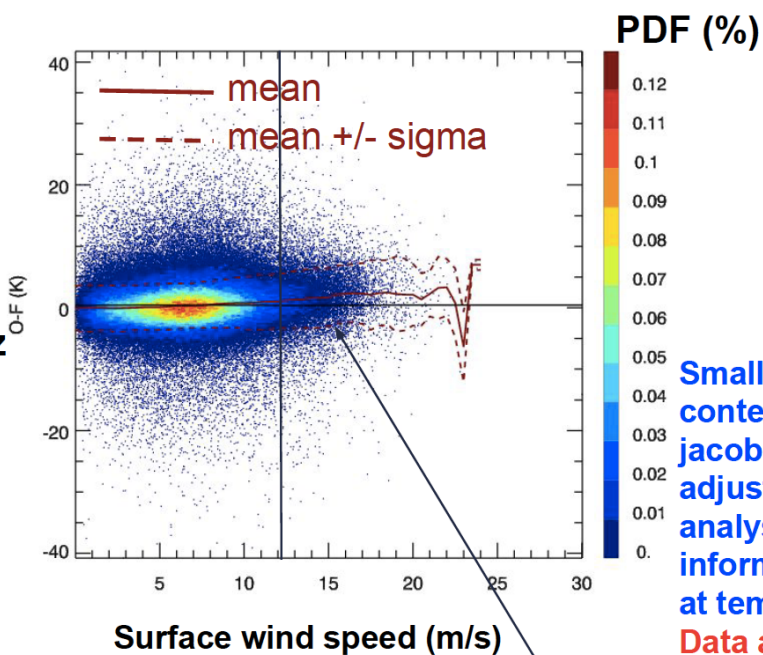
Forecasted / Guess clouds (CLW_ges) are retrieved from **Tb forecasted** by a radiative transfer model (CRTM) with inputs from GEOS-5 profiles including clouds. They are not prognostic clouds in GEOS GCM.

→ “See clouds through the same lens.”



Quality control

Tb O-F,
Ch9
23V GHz

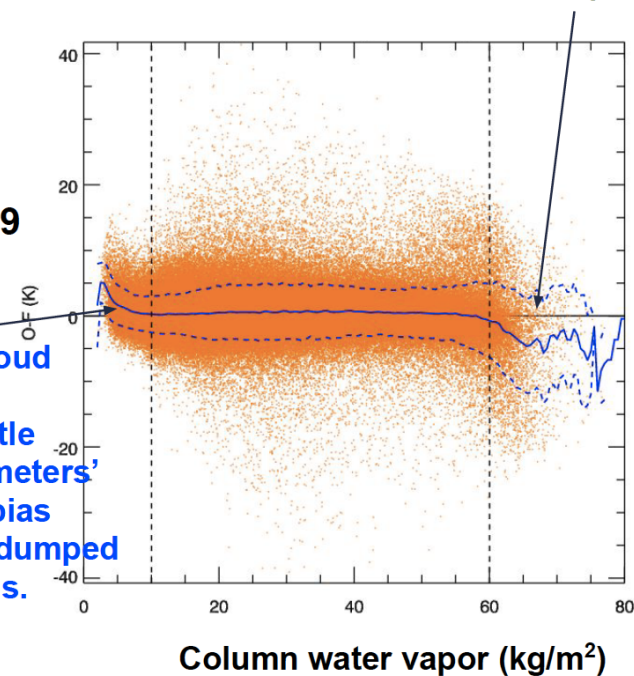


Bias is due to uncertainties in surface emissivity simulation. Data are tossed out.

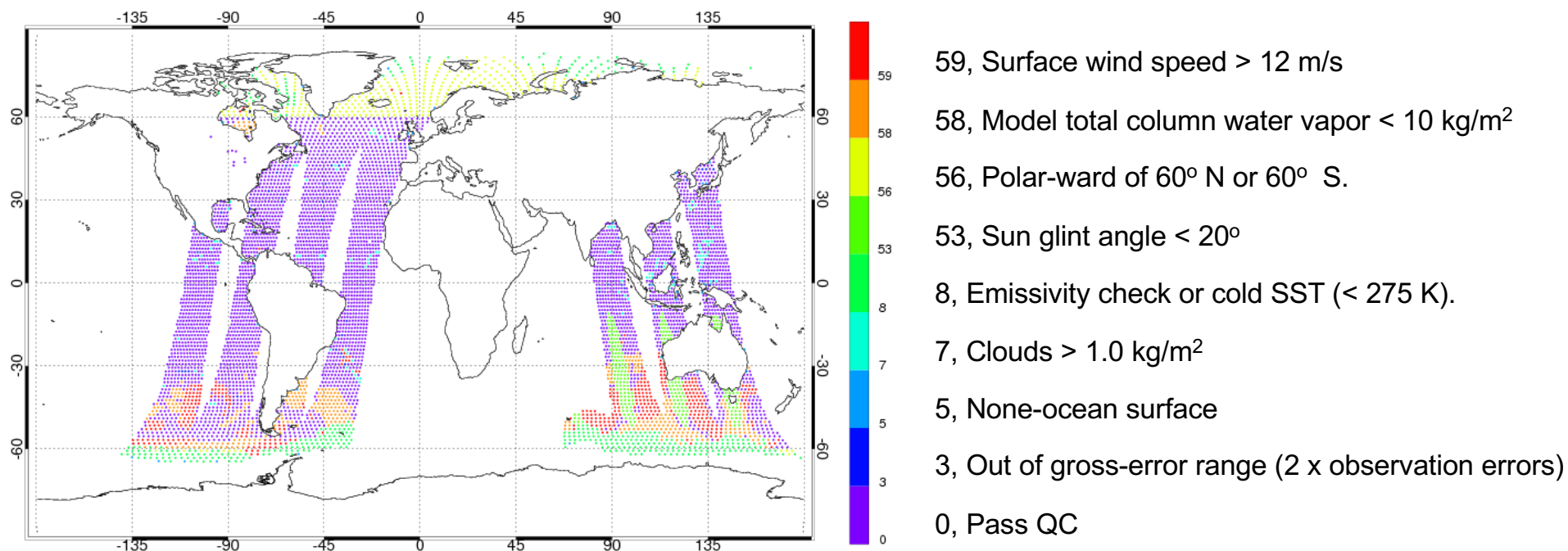
Tb O-F, Ch9
23H GHz

Small moisture and cloud contents lead to small jacobians, and thus little adjustments in hydrometers' analysis. As a result, bias information would be dumped at temperature analysis. Data are tossed out.

Bias is likely caused by moisture and clouds. Cloud analysis hopefully will reduce the O-F here. Data are kept.



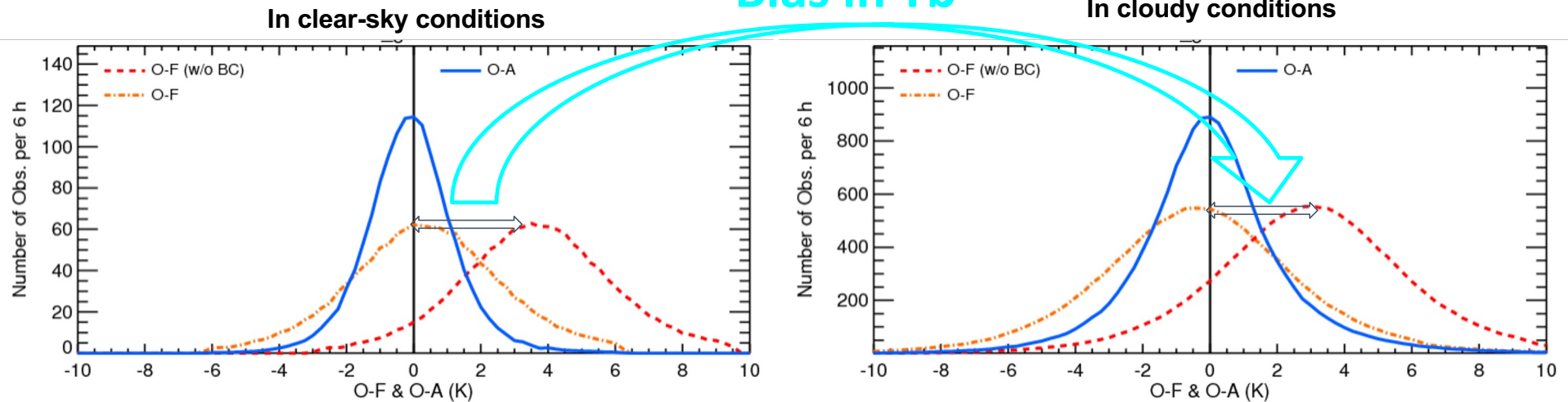
AMSR2 data quality control: Examples



**An example of data screening criteria between 03 hr – 09 hr, Aug 01, 2018 .
Data over land, ice or snow surfaces are not processed.**

Bias corrections in the all-sky assimilation framework

Bias in Tb

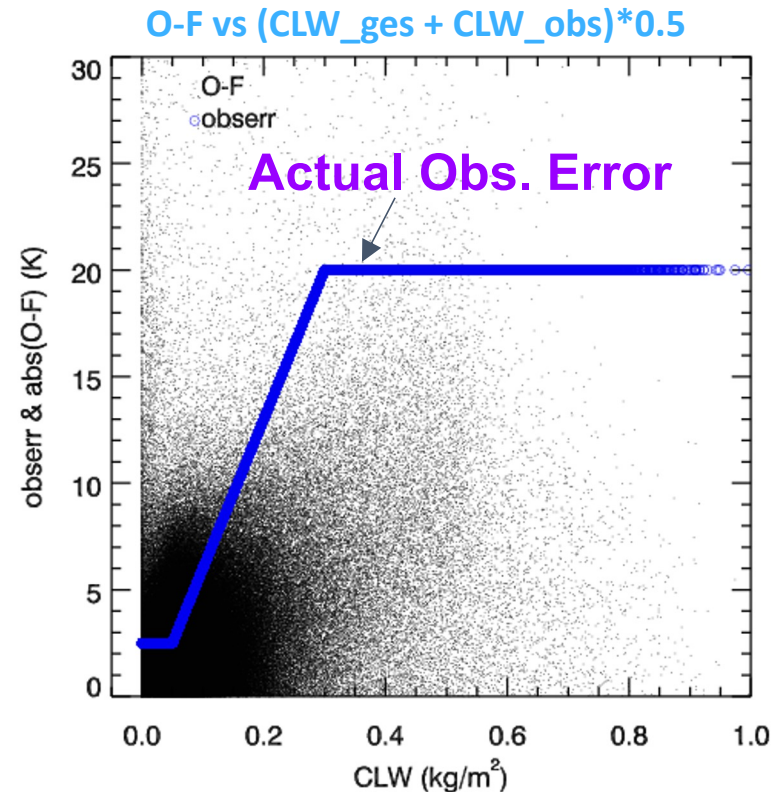
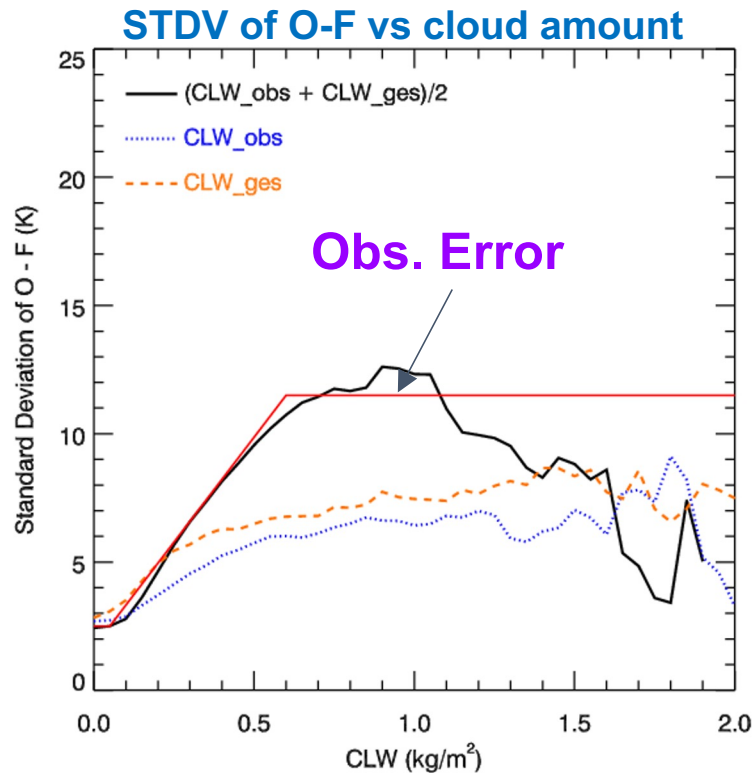


Histograms of Tb departure, O-F, before and after bias correction, and Tb O-A (ch 9, example)

Use the bias correction information from data in clear-sky conditions to correct data in both clear-sky and all-sky conditions.

Note: Cloud amount is not a bias correction predictor within the all-sky assimilation framework.

Observation error: A function of symmetric clouds

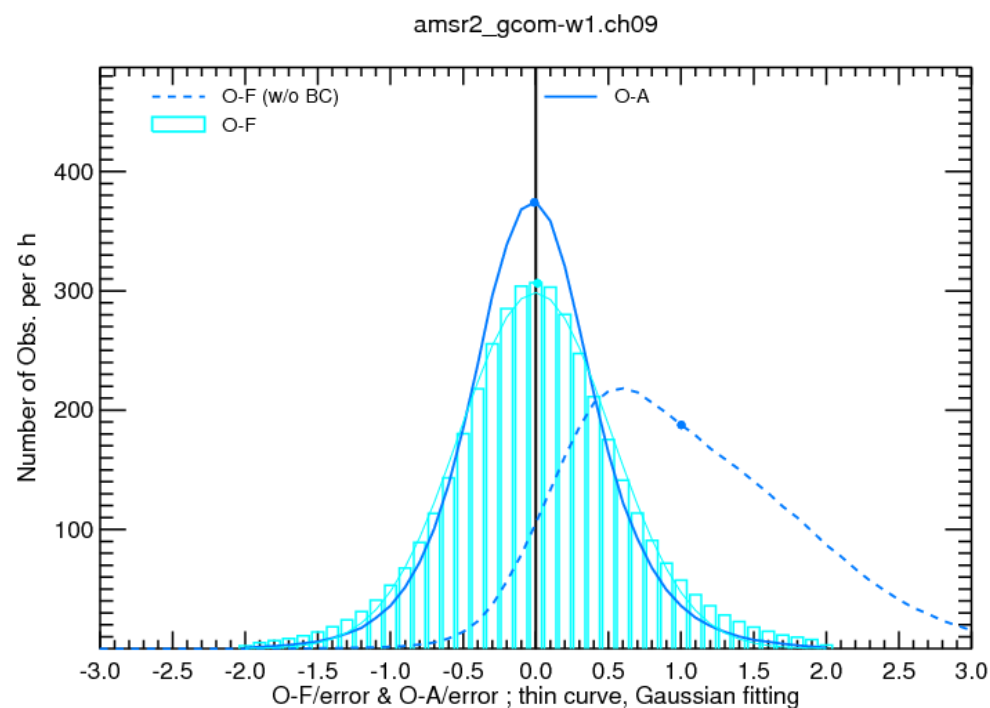


- **Observational errors have to be largely inflated** because there is a large inter-channel correlation between observational error covariances (not shown).
- **Inflation may be not large enough at locations where there are small mean CLW between mis-matched clouds.**



An experiment to assimilate AMSR.

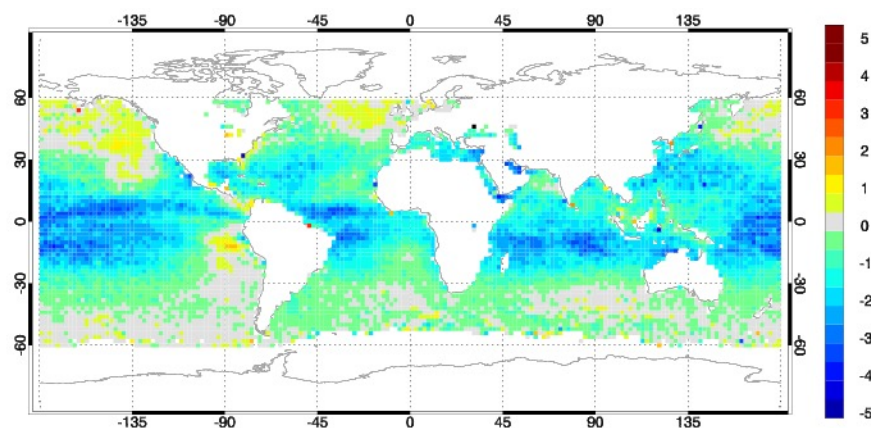
- Assimilate AMSR2 along with other observational data sets within GEOS in June-Aug 2018.
- An Ensemble-replay experiment:** Use the ensemble in the control and only update central analysis.
Background errors for hydrometeors are derived from the ensemble analysis only.
- Only assimilate AMSR2 channel 9 (23.8 GHz) and 11 (36.5 GHz).



Normalized O-F and O-A departures in July-Aug 2018

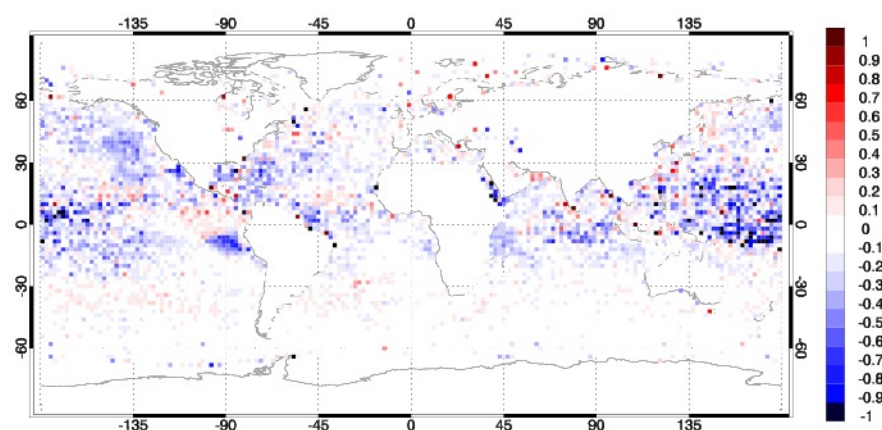
Impact on GEOS column water vapor analysis by AMSR2 data

GEOS Exp. – RSS, (kg/m^2)



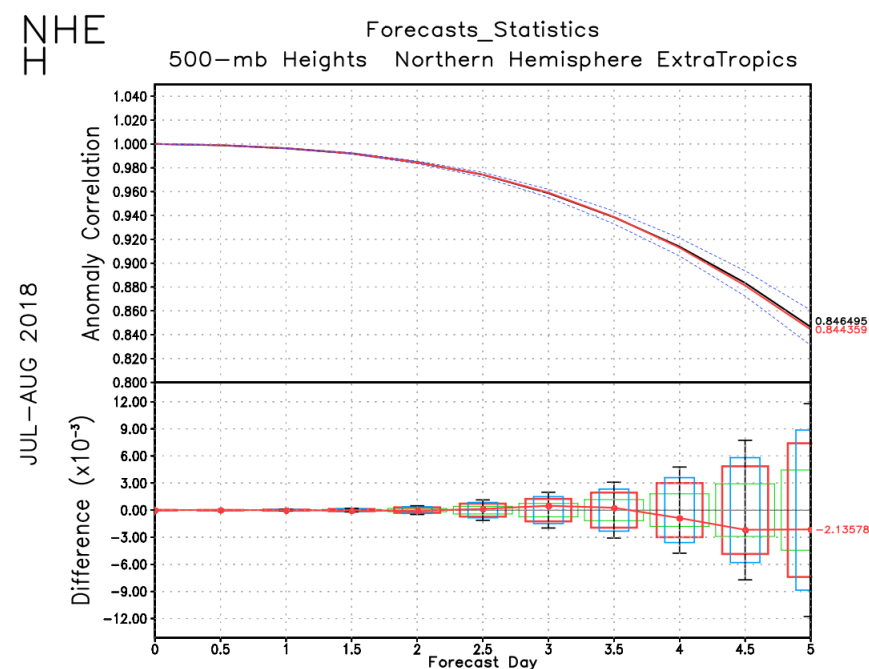
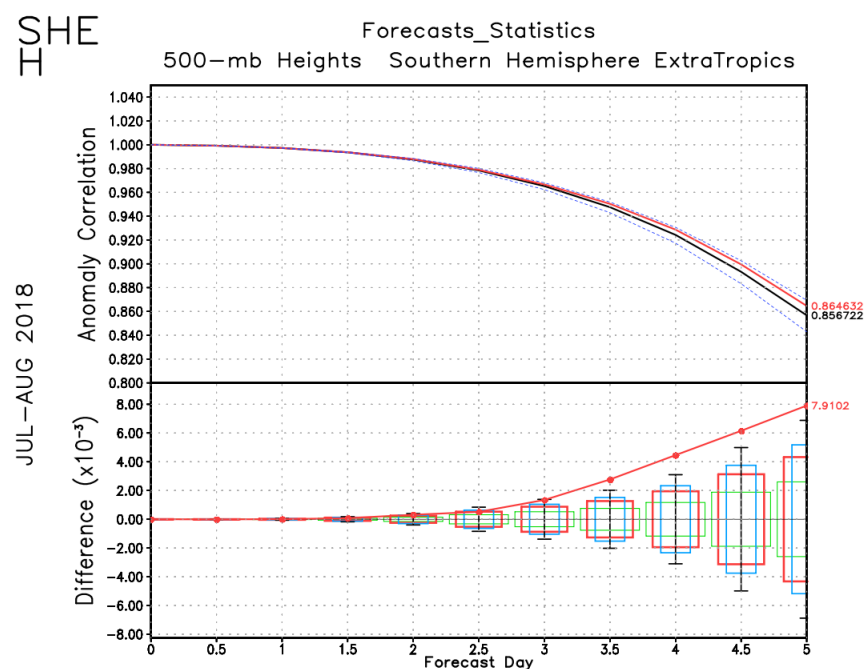
- Total column water vapor in GEOS analysis is generally smaller than retrievals made by Remote Sensing Systems, Inc. (RSS) from AMSR2 data during July-August 2018.

$|\text{Exp.} - \text{RSS}| - |\text{Control} - \text{RSS}|$, (kg/m^2)



- Assimilation of AMSR2 data brings the GEOS analysis closer to the retrievals in tropics and the east Pacific.

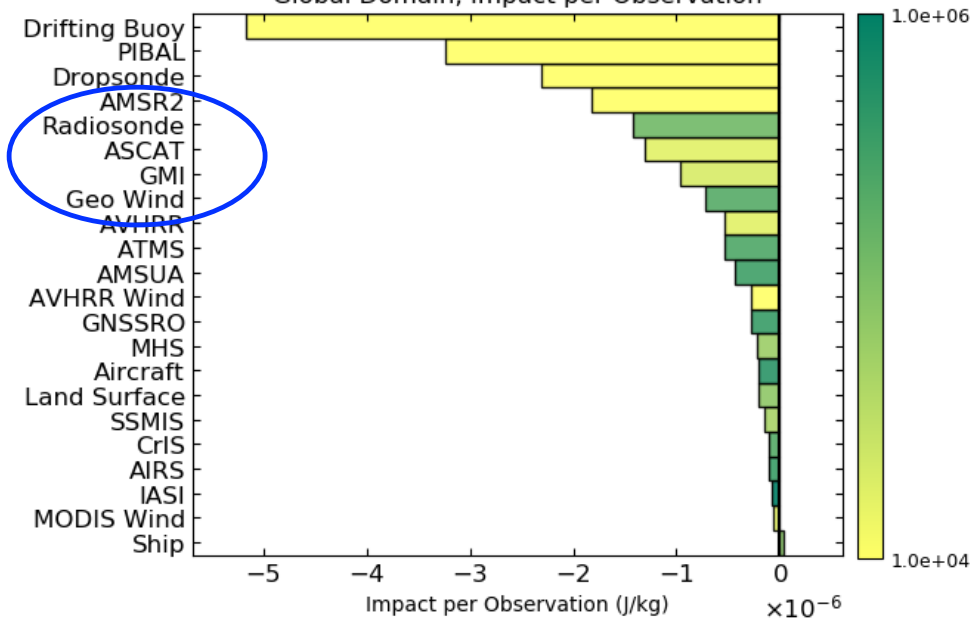
Impact of assimilating AMSR2 data in GEOS forecasts



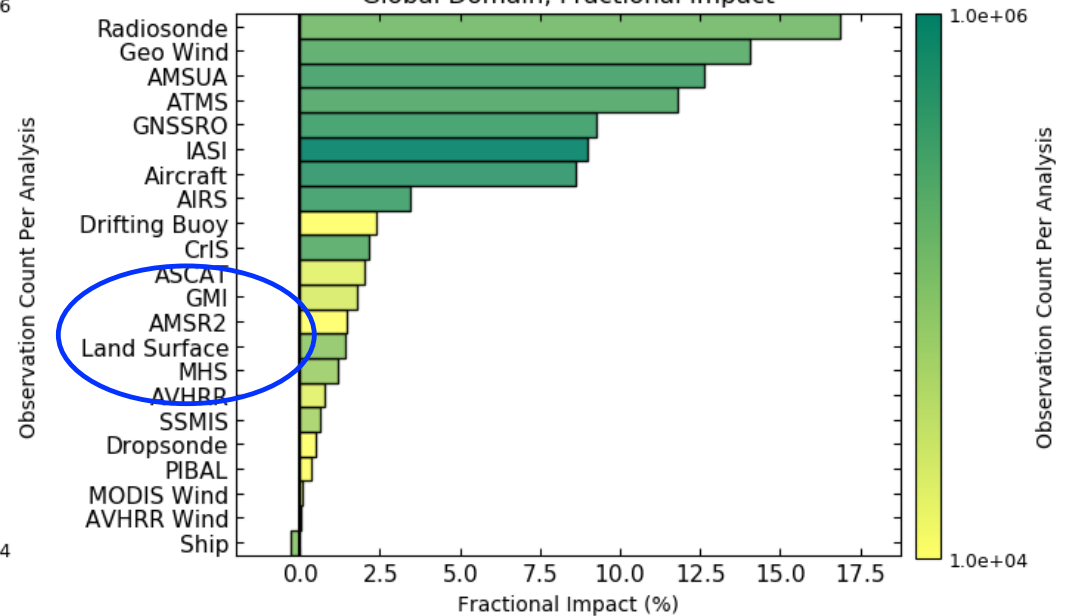
- Assimilating AMSR2 data improves GEOS forecasts in the southern hemisphere during July-August 2018 as shown in the left panel, but it has a neutral impact in the forecasts in the northern hemisphere as shown in the panel.

Impacts of assimilating GMI and AMSR2 in current GEOS

GEOS 24h Observation Impact Summary
26 Nov 2023-24 Feb 2024 00z
Global Domain, Impact per Observation



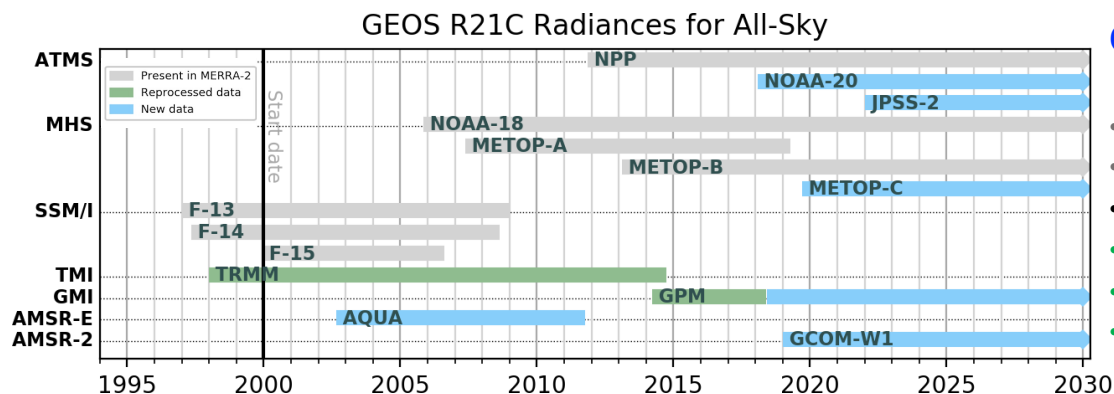
GEOS 24h Observation Impact Summary
26 Nov 2023-24 Feb 2024 00z
Global Domain, Fractional Impact



- AMSR2 and GMI have largest impact per observations among satellite radiance data but have overall smaller impact than many other MW or IR sounder **because of limited data are assimilated.**
- **Among these MW data, AMSR2 and GMI brightness temperature data at 23 GHz have larger impacts than data at other channels do.**

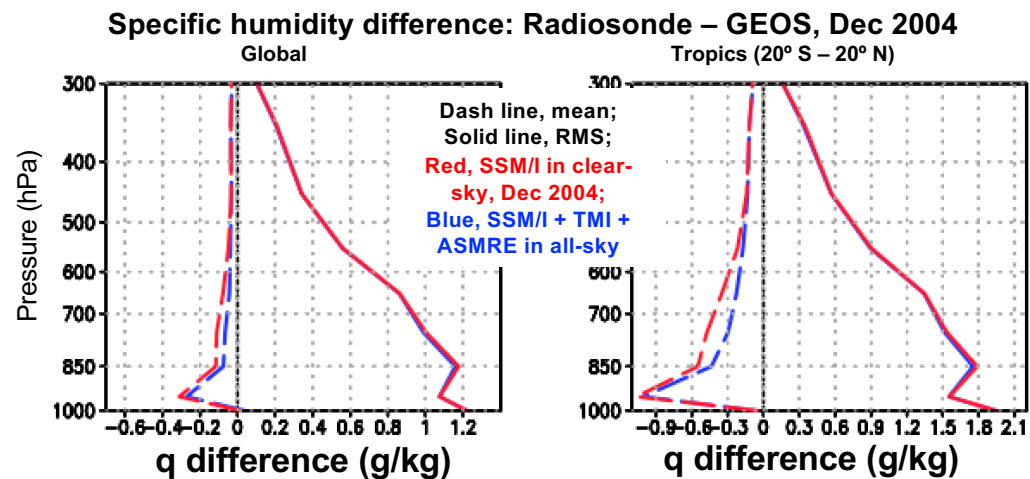


Microwave radiance all-sky data assimilation in GEOS-R21C



Changes from MERRA-2

- MHS: clear-sky => all-sky
- SSM/I: clear-sky => all-sky
- TRMM: rain-rate => all-sky
- AMSR-E: not used => all-sky
- AMSR-2: not used => all-sky
- GMI: not used => all-sky

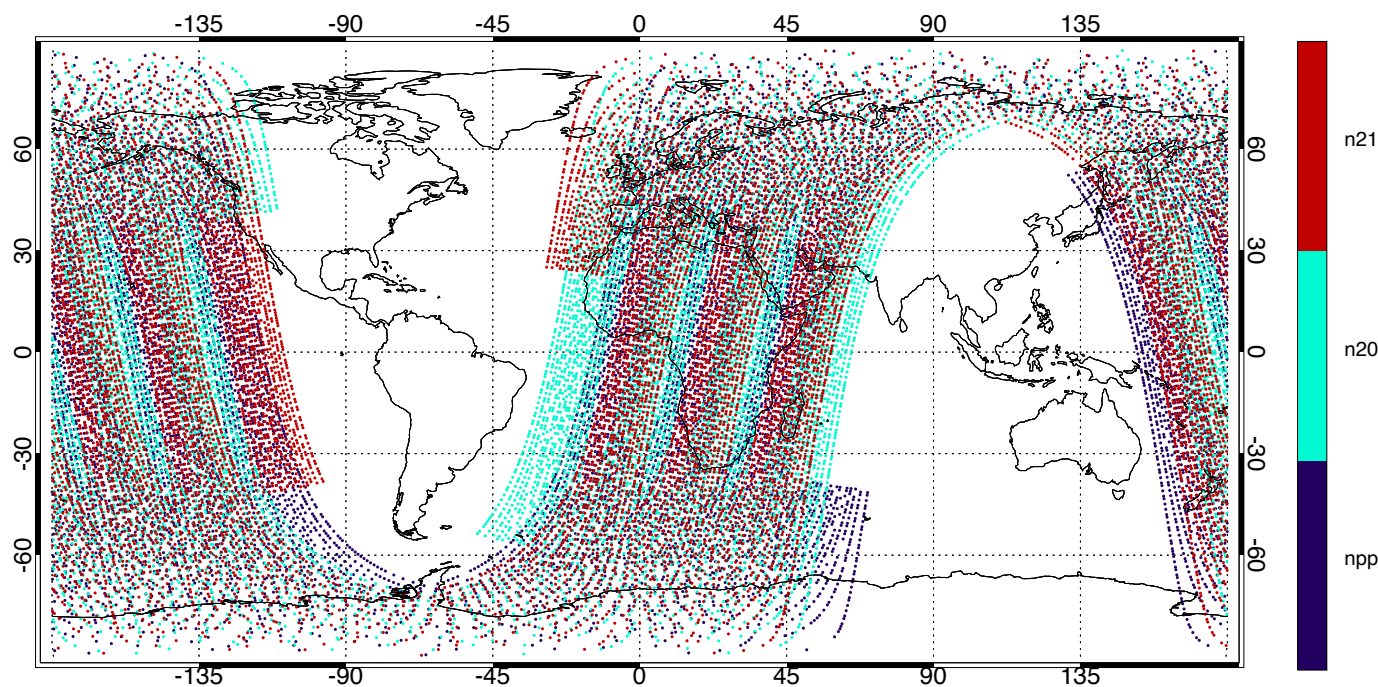


- GEOS humidity analysis is closer to radiosonde observations in tropical convection regions.



Assimilate New ATMS_NOAA 21 Observations

ATMS Suomi-NPP, NOAA-20, and NOAA-21 observation locations within a 6-hour window centered at 00hr 07/01/2023.



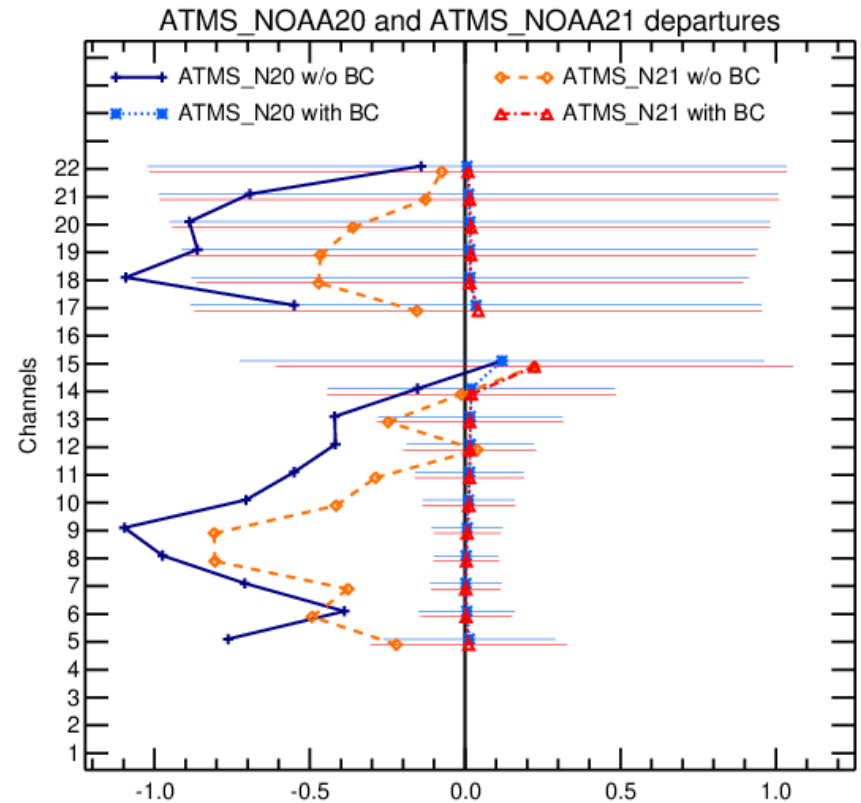
Each satellite crosses the equator at 1:30 PM (local time) in northern-bound direction at Sun-synchronous orbits.

The NOAA-21 is followed for about half of an orbit by NOAA-20, and Suomi-NPP is flying between them.



Data Selection

- Assimilate temperature sensitive channels 5-15, and moisture sensitive channels 17-22.
- Data are tossed over ice, snow, or mixed surface types. IFOV shapes are not considered when surface properties are interpolated. Data are also tossed at scan edges.
- ATMS data are assimilated in clear-sky and thin-cloud conditions within the clear-sky assimilation framework. Cloud effect is bias corrected.
- An experiment is conducted between mid-July – Aug 2023.



ATMS_NOAA20 and ATMS_NOO21 mean O-F w/o and with bias corrections in Aug 2023 (K). Error bars are standard deviations of O-F with BC.



Forecast impact

Forecast score card
based on statistics
verified against ERA5
in August 2023

Legend

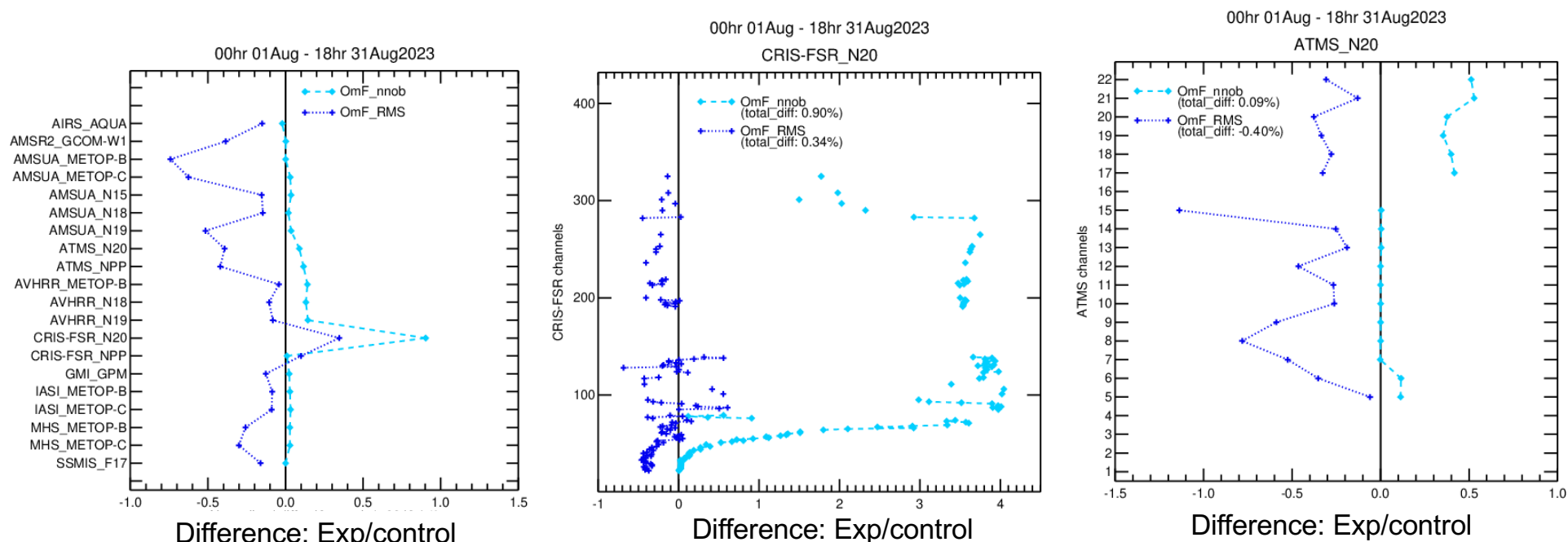
- ▲ far better, significant (99.99% confidence)
- △ better, significant (99% confidence)
- ⬢ slightly better, significant (95% confidence)
- no significant difference
- ⬢ slightly worse, significant (95% confidence)
- ▼ worse, significant (99% confidence)
- ▼ far worse, significant (99.99% confidence)

Northern Hemisphere											
Variable	Pressure Level	COR					RMS				
Geopotential Height	Forecast Day	1	2	3	4	5	1	2	3	4	5
	10						▼	▼	▼	▼	▼
	70										
	100										
	250										
	500										
	700										
SLP	850										
	1000										
Specific Humidity	10										
	70										
	100										
	250										
	500										
	700										
Temperature	850										
	1000										
	10										
	70										
	100										
	250										
U-Wind	500										
	700										
	850										
	1000										
	10										
	70										
V-Wind	100										
	250										
	500										
	700										
	850										
	1000										

Southern Hemisphere											
Variable	Pressure Level	COR					RMS				
Forecast Day		1	2	3	4	5	1	2	3	4	5
Geopotential Height	10										
	70										
	100										
	250										
	500										
	700										
SLP	850										
	1000										
Specific Humidity	10										
	70										
	100										
	250										
	500										
	700										
Temperature	850										
	1000										
	10										
	70										
	100										
	250										
U-Wind	500										
	700										
	850										
	1000										
	10										
	70										
V-Wind	100										
	250										
	500										
	700										
	850										
	1000										

Tropics											
Variable	Pressure Level	COR					RMS				
Forecast Day		1	2	3	4	5	1	2	3	4	5
Geopotential Height	10						▼	▼	▼	▼	▼
	70										
	100										
	250				✱						
	500										
	700										
SLP	850			✱							
	1000										
Specific Humidity	10										
	70						▲	▲	▲	▲	▲
	100						▲	▲	▲	▲	▲
	250					✱					
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Temperature	850										
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U-Wind	700										
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	10										
	70										
	100										
	250										
V-Wind	500										
	700										
	850										
	10										
	70										
	100										

Compare O-F rms values and number of used data

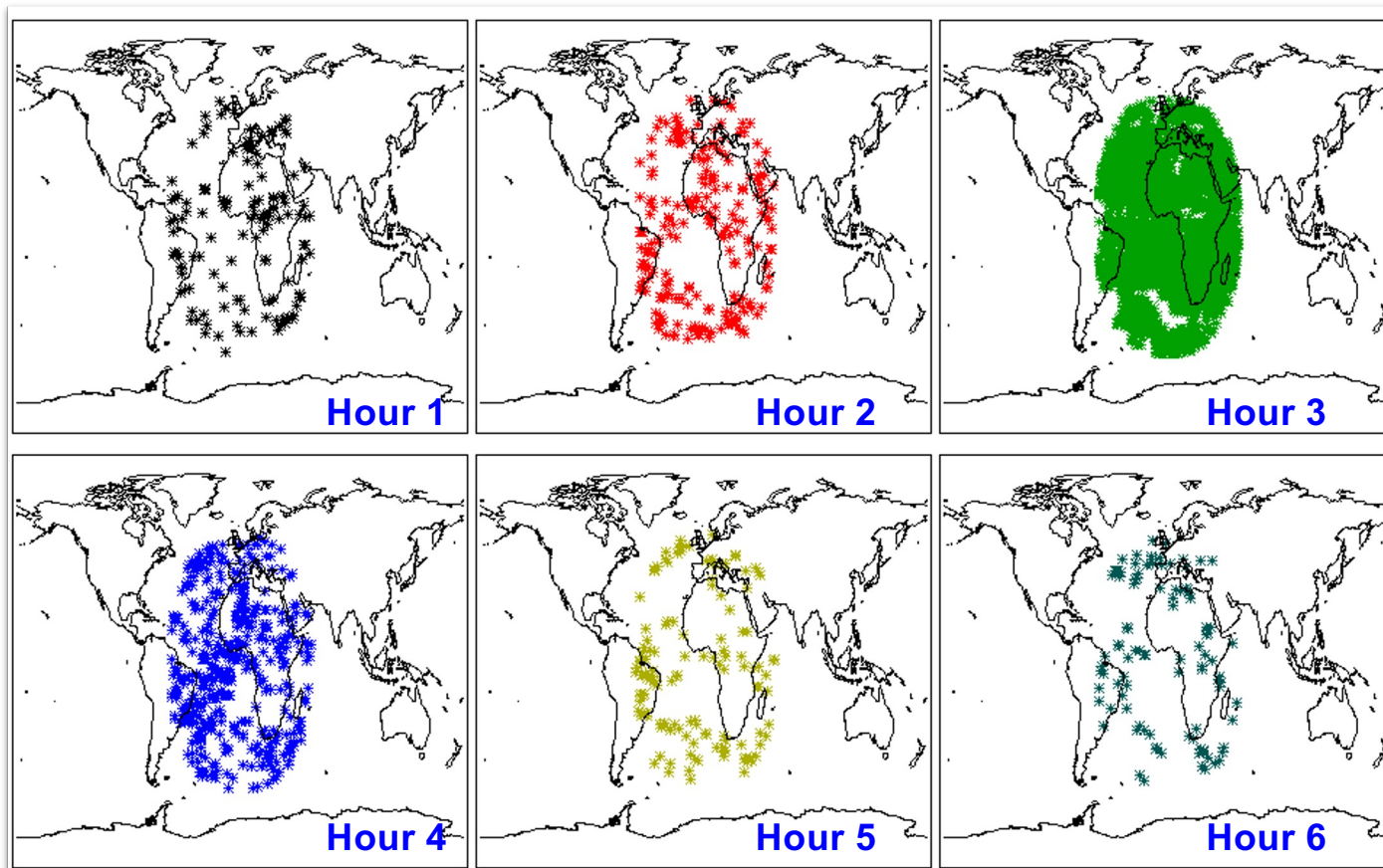




Improve Geostationary Radiance Data Selection

- Geostationary radiance data near the center of six-hour time-window were given preferences **during thinning** in a 3-D assimilation system.
- In current 4-D system, No time information is used during thinning. Therefore, observations are selected per the “first-come, first-selected” manner when they meet other data selection criteria. **More data are selected within the 1st hour because satellite data are saved and read sequentially.**
- New two-hourly-thinning for geostationary radiance data: Preference is given to observations at the beginning of the 1st two-hour period, the center time of the 2nd two-hour period, and the end of the 3rd two-hour period.

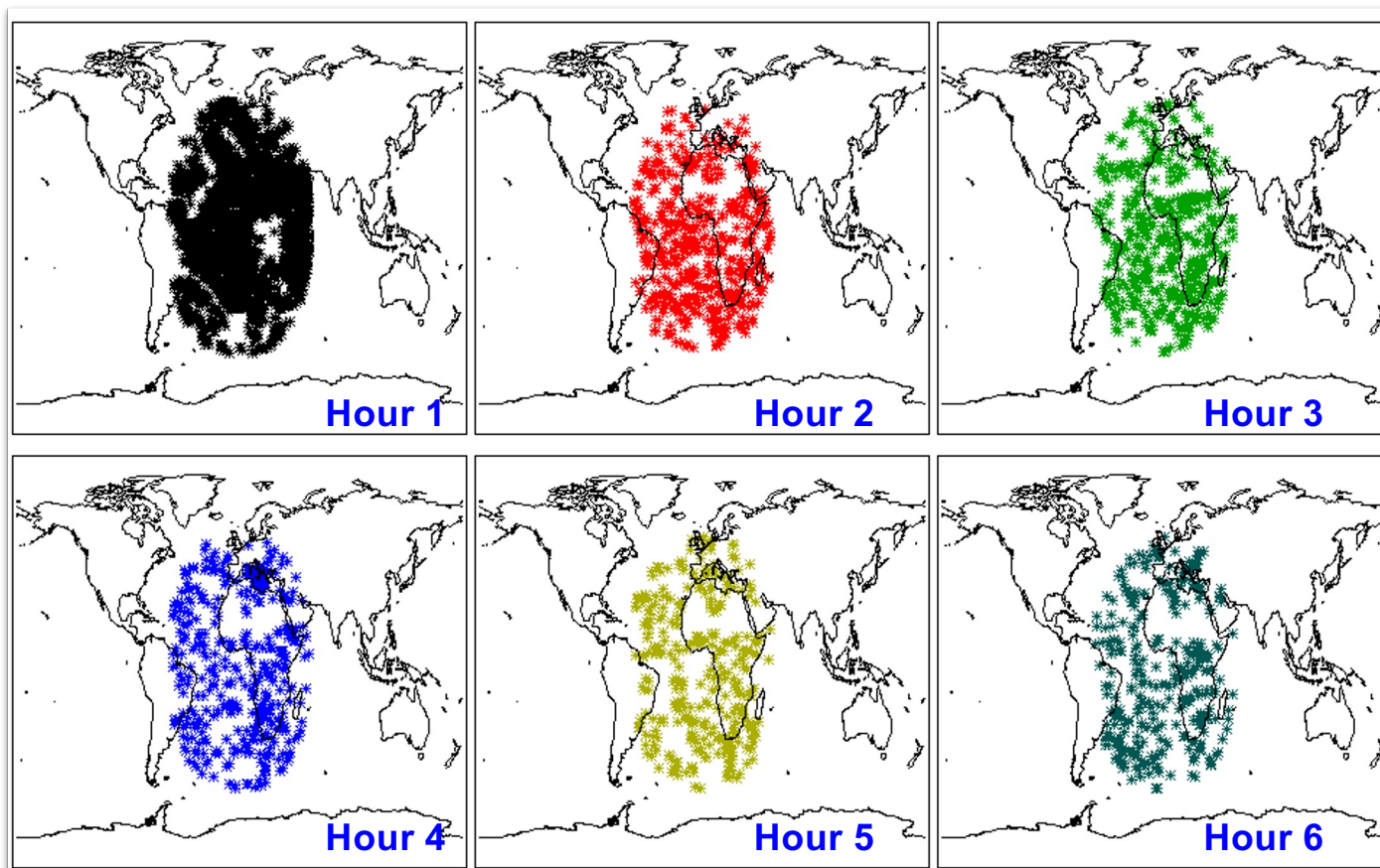
Maps of SEVIRI data after thinning within a 3-D assimilation framework



SEVIRI_M10 at
18 UTC, Jun 1,
2017.

More
observations are
selected within
hour 3 than other
hours.

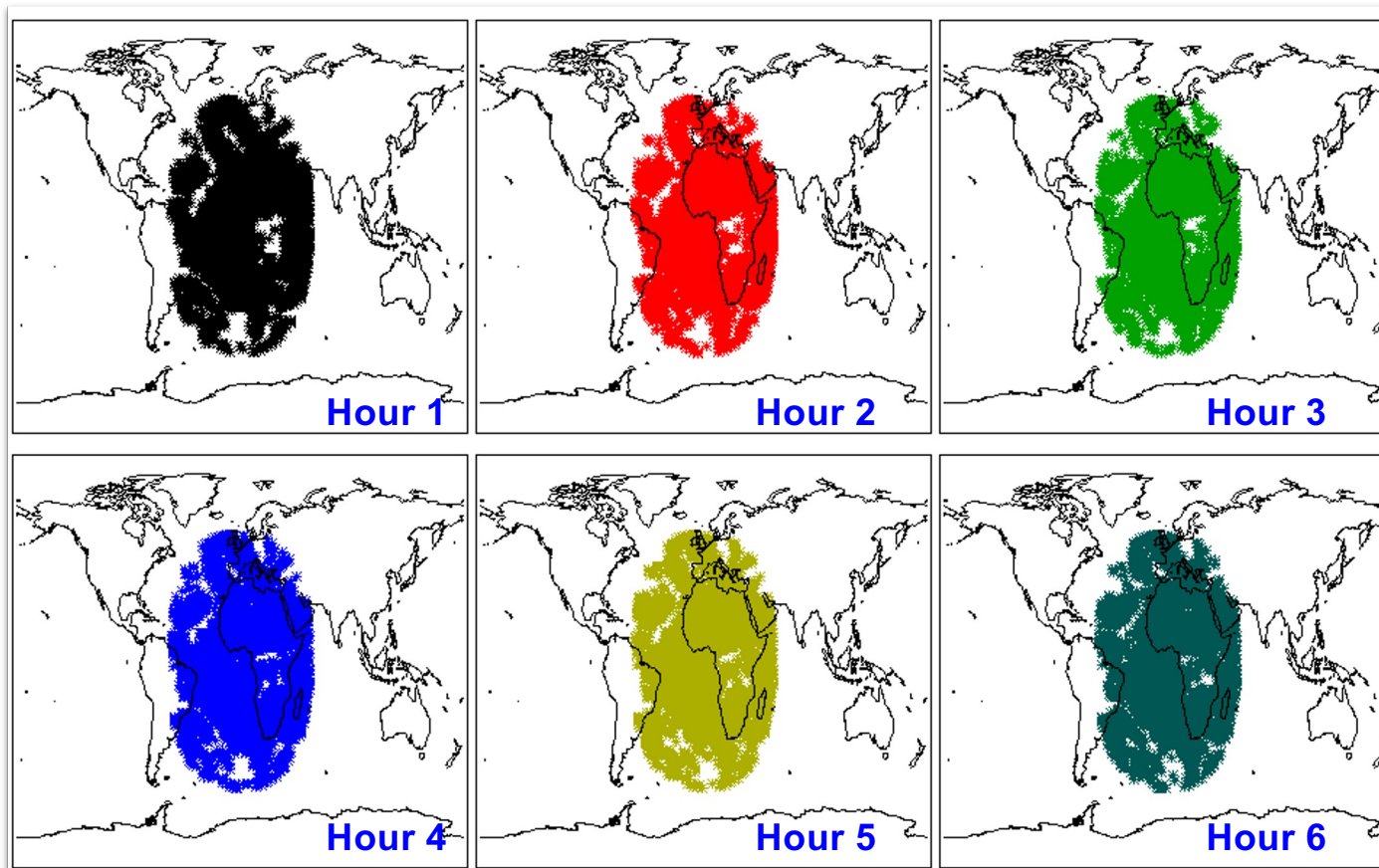
Maps of SEVIRI data after thinning in current 4-D system.



SEVIRI_M10 at
18 UTC, Jun 1,
2017.

More
observations are
selected within
hour 1 than other
hours.

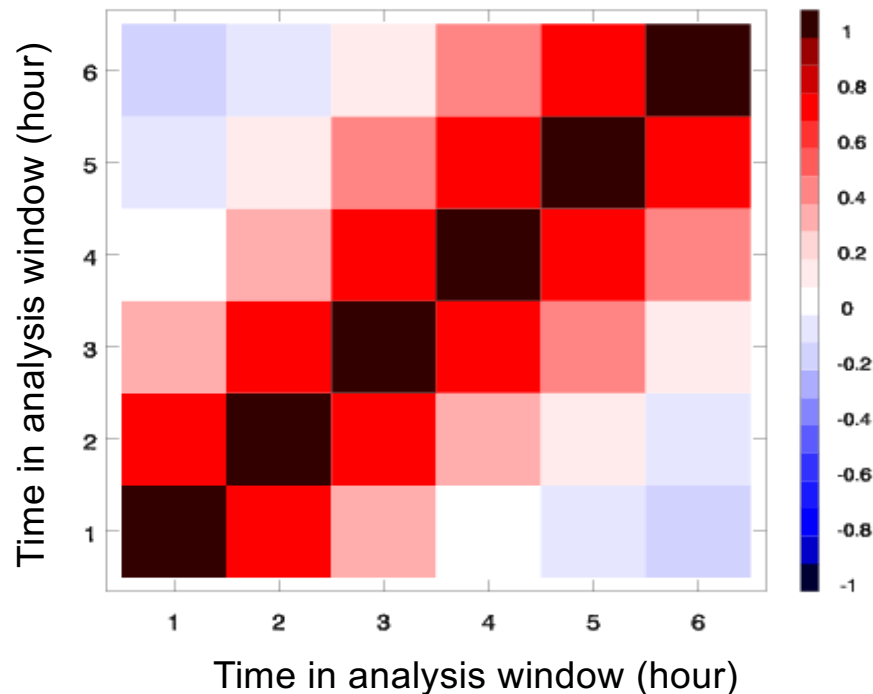
First attempt to fix the issue: Hourly-thinning.



SEVIRI_M10 at 18 UTC, Jun 1, 2017.

Set thinning time-window to be one hour. Nearly same number of data is selected within each hour after thinning.

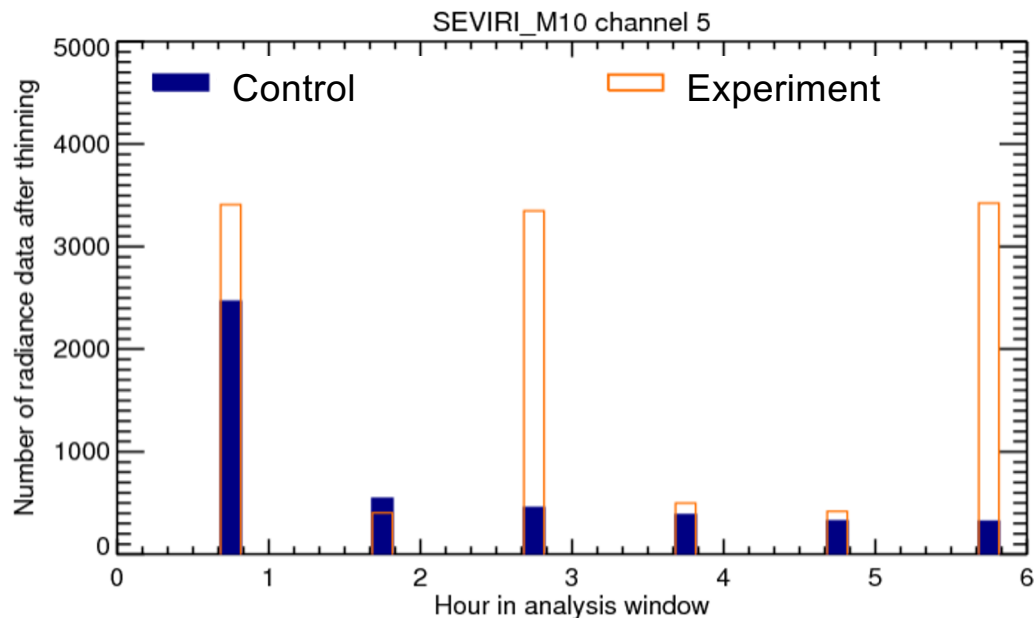
Correlations of observational error SEVEIRI_M10, cha 5.



- Hourly-thinning has a **slightly negative impact** in GEOS-5's forecast skills.
- Time correlations of SEVIRI_M10 channel 5 (6.25 μm) observation error in Dec 2016.
- These correlations are derived from O-F and O-A covariance within the same six-hour analysis cycles.
- This figure **shows large observation error correlations within 2-3 hours.**



New solution: Two-hourly-thinning

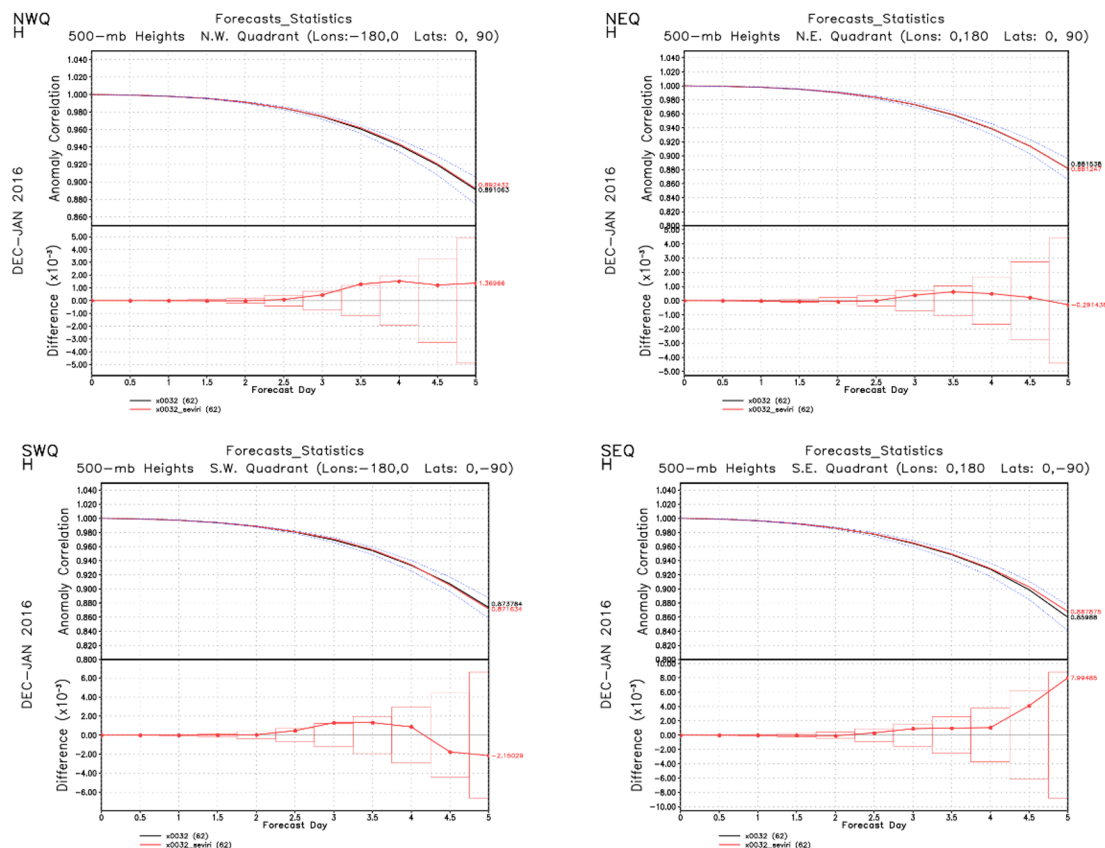


Number SEVIRI_M10 data in control and the experiment (two-hourly-thinning) after thinning in Dec 2016

- **New two-hourly-thinning:** Preference is given to observations at the **beginning of the 1st two-hour period, the center time of the 2nd two-hour period, and the end of the 3rd two-hour period.**
- As a result, we can have nearly the same number of observations at the beginning, middle, and end of the six-hour time window for geostationary satellite observations.



Forecast impact: Anomaly correlation of forecasted GPH.



Experiment: Control + two-hourly-thing of SEVIRI data

- Slightly positive impact on forecasts in the Southeast Quadrant than in other regions. where most of SEVIRI_M10 observations are located in the tropical Atlantic Ocean.



Summary and on-going work

- All-sky microwave radiance data assimilation
 - Very few low frequency and surface sensitive microwave radiance are assimilated over land, ice or snow surfaces.
 - Explore “all-sky-all-surface” assimilation of microwave radiance observations to improve planetary boundary analysis.
- Assimilate polar orbit infrared observations in all-sky conditions.
- Geostationary radiance
 - No geostationary radiance data are assimilated in GEOS. Will investigate the impact of assimilate those data over ocean.
- New NASA missions.

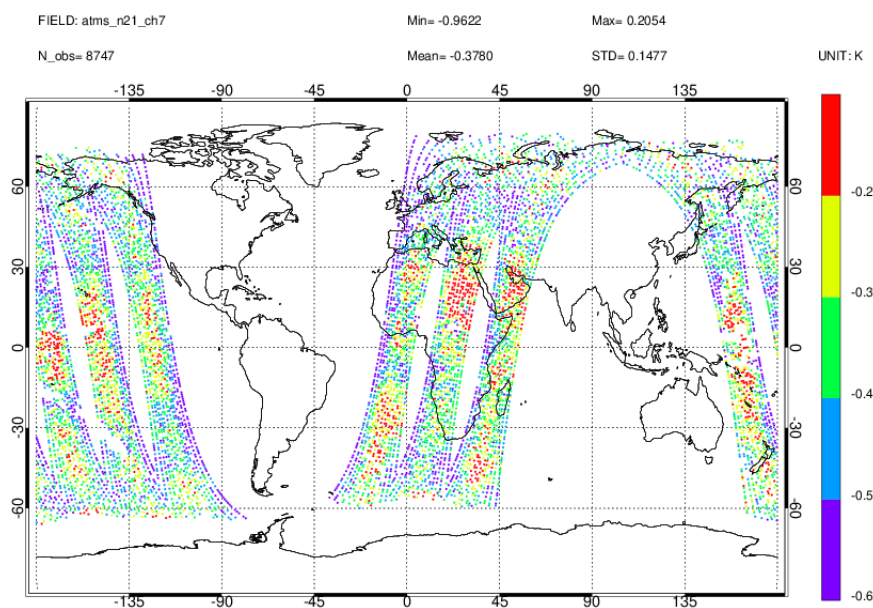
Thank you!



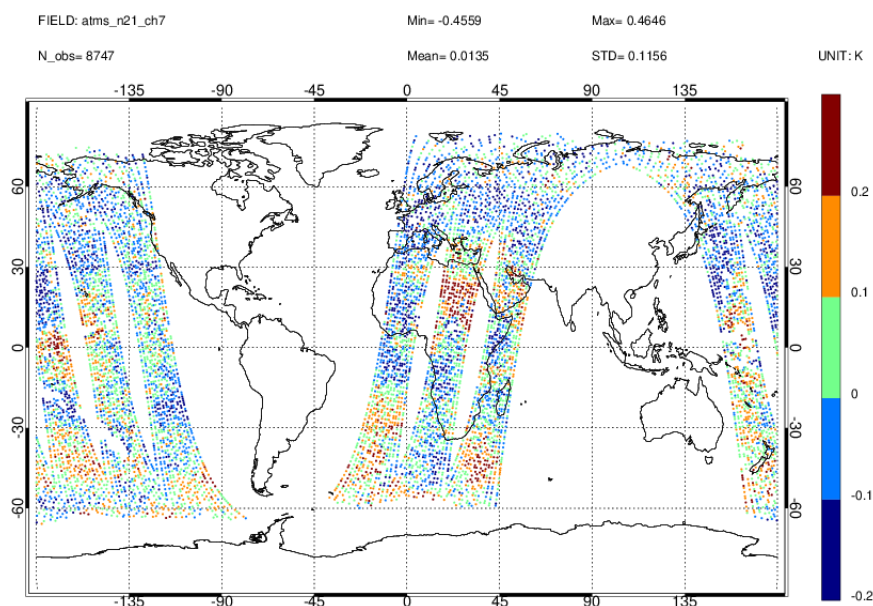
The end

ATMS NOAA21 ch7 O-F before and after bias correction

OmF_nbc_20230801_00z.20230801_00z.ch07

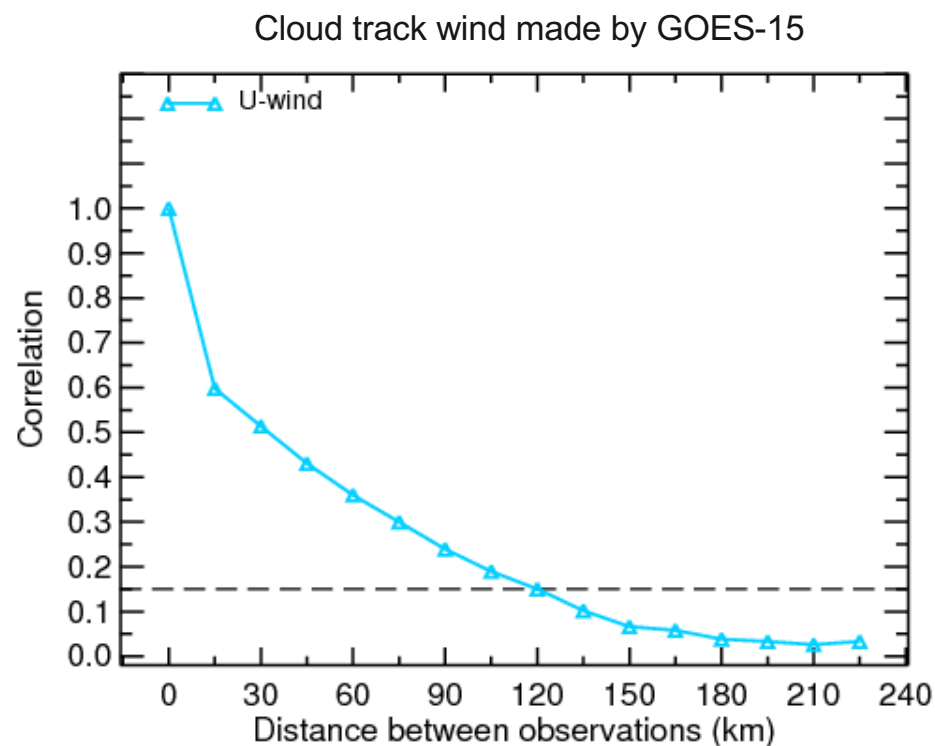


OmF_20230801_00z.20230801_00z.ch07



Horizontal Correlations In Wind Observational Errors

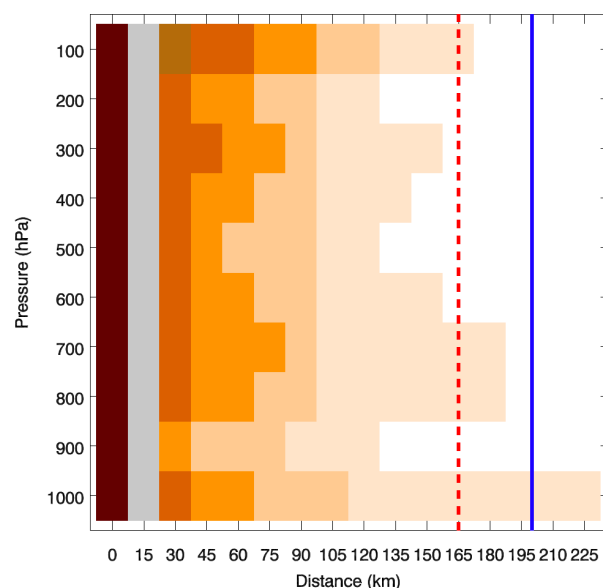
- In a DA system, observational error covariance can be (not perfectly) diagnosed as
covariance (Obs - Analysis, Obs - First guess)
- Correlation in observational errors, derived from the above covariance, decreases as distances between observations increase.
- We chose the horizontal distance where correlation is 0.15 as the new thinning grid size for AMV data.



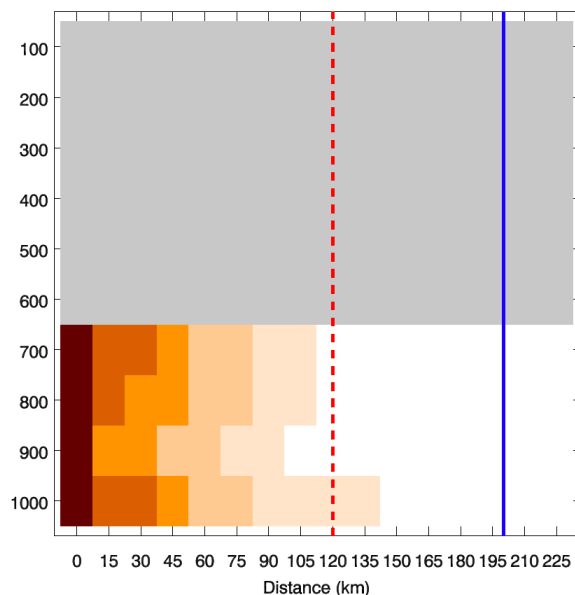


Horizontal Correlations In Observational Errors

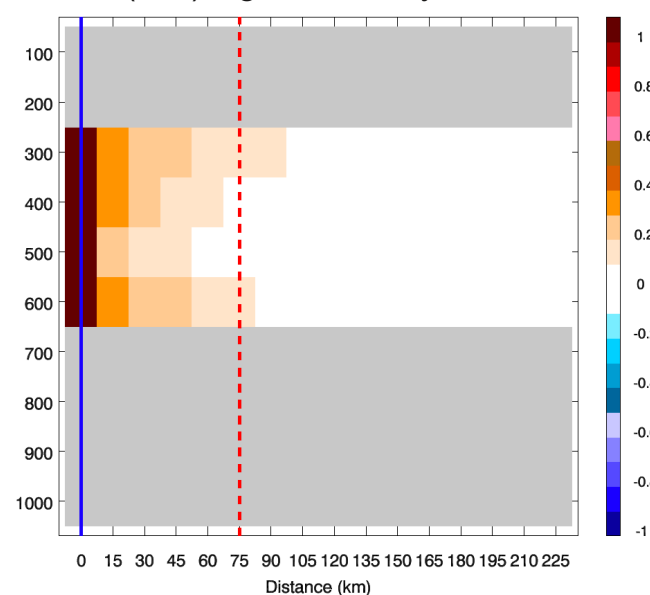
For (IR) cloud track U-wind by Himawari



For (visible) U-wind by Meteosat-11



For (WV) high U-wind by GOES-R

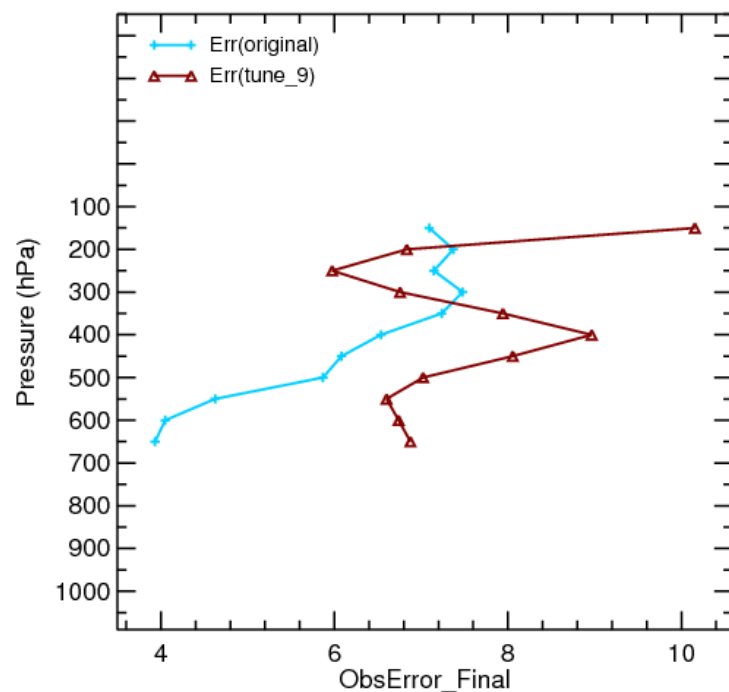


- Distances at **blue lines**: thinning grid sizes in the original configuration, 200 km or zero.
- Distances at **red-dash lines**: maximum distance where horizontal correlations is 0.15, **proposed as new thinning grid sizes**.
- Gray: No co-located observation.
- **AMV data are thinned in a scientifically consistent way.**

Tune Observational Error Inputs

Adjust observational error inputs in order to give similar weights (penalties) of AMV at different altitudes.

Observational errors for WV cloud top AMV by Meteosat-11.



Penalties for WV cloud top AMV by Meteosat-11.

