Satellite Data Assimilation in Regional Models: Promises and Challenges

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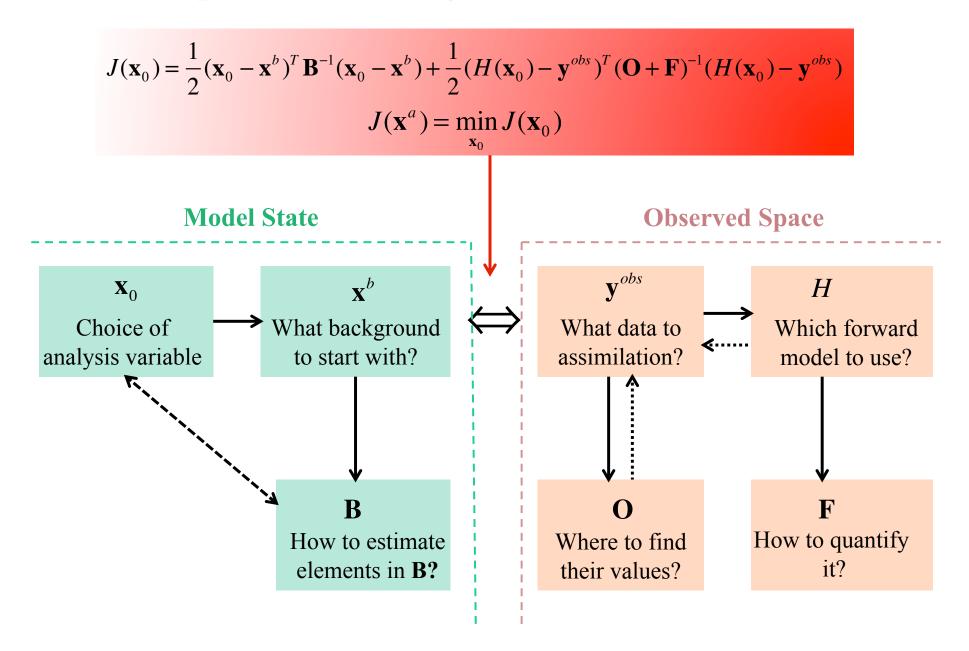
1

September 4, 2013

Outline

- Important Building Blocks for Satellite DA
- GOES Imager Radiance Assimilation in GSI/ARW
- POSE MHS Radiance Assimilation in GSI/ARW
- SNPP ATMS Radiance Assimilation in GSI/HWRF
- Summary & Conclusions

Important Building Blocks for Satellite DA



Part I

An Evaluation of Added Benefits of GOES Imager Radiances to Other Satellite Data Assimilation

GOES Imager Radiance Assimilation in GSI/ARW

1) Comparison of Single Type Satellite Data Assimilation

✓ AMSU-A	✓ HIRS/4	✓ HIRS/3	✓ GSN
✓ AIRS	✓ AIRS	✓ MHS	✓ GOES Imager

2) GOES Imager Added to Different Types of Satellite Data

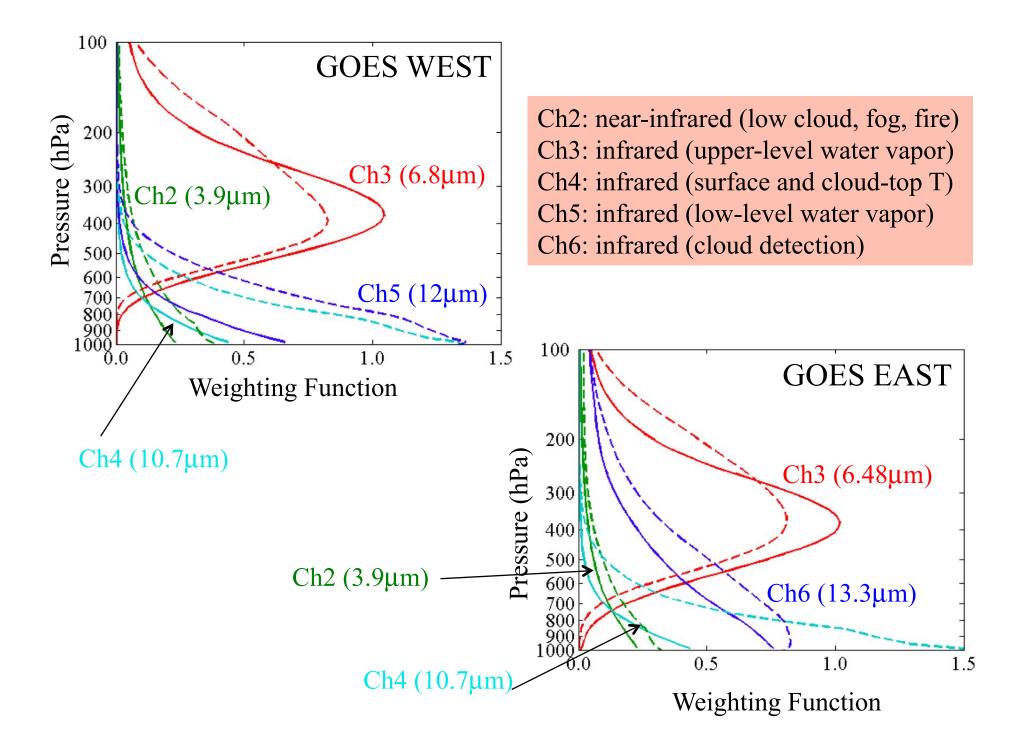
3) GOES Imager Added to All Satellite Data Assimilation

4) Impact of Quality Control on MHS Data Assimilation

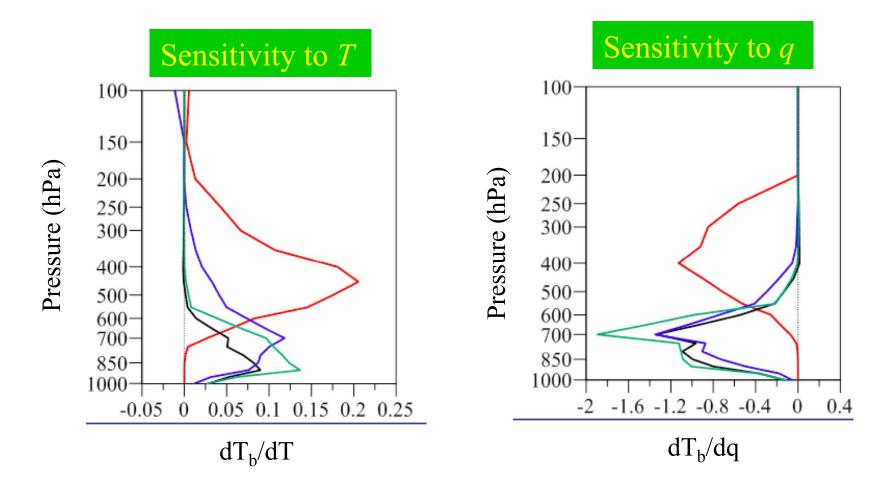
GOES West (11) and GOES East (12)



GOES-R Series Imagery Coverage Figure

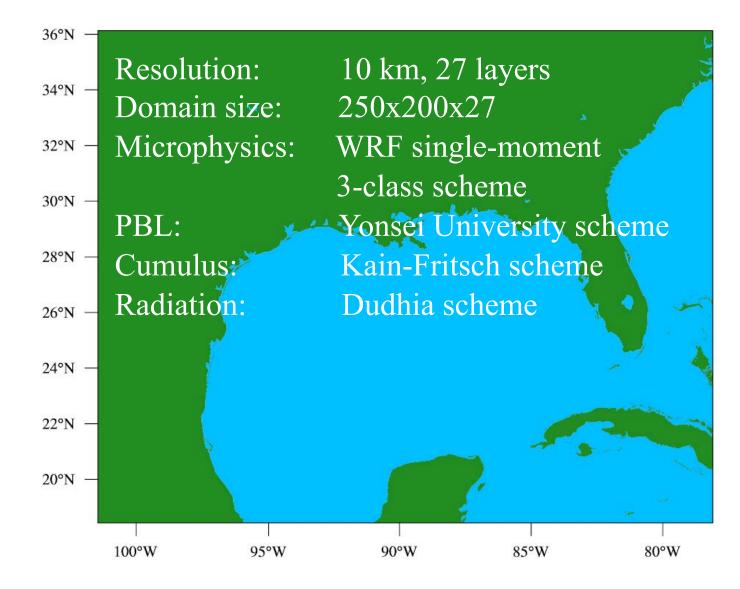


Mean Jacobian of Brightness Temperature



-Ch3 -Ch4 -Ch5 -Ch6

Advanced Research WRF (ARW) Model Domain

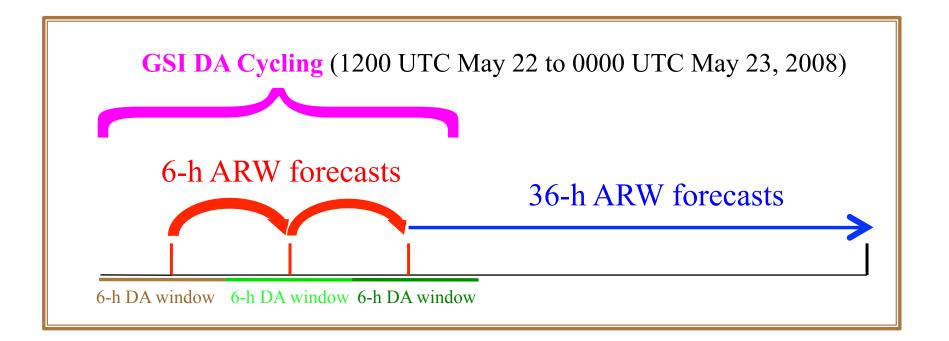


NCEP GSI 3D-Var Data Assimilation System

Assimilation of Different Combinations of Observations

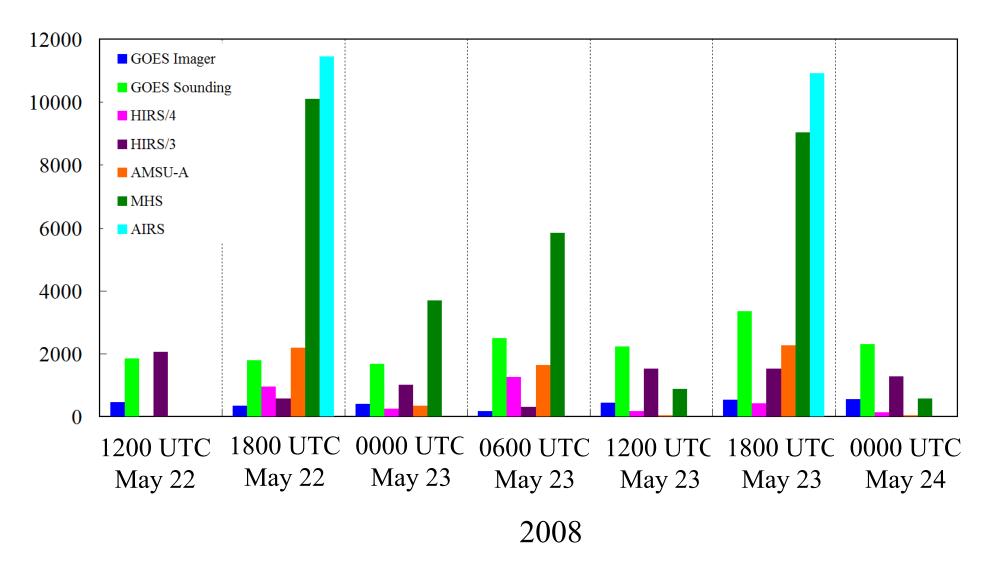
AMSU-A, AIRS, HIRS/3, HIRS/4, MHS, GSN,

GOES imager, Conventional data

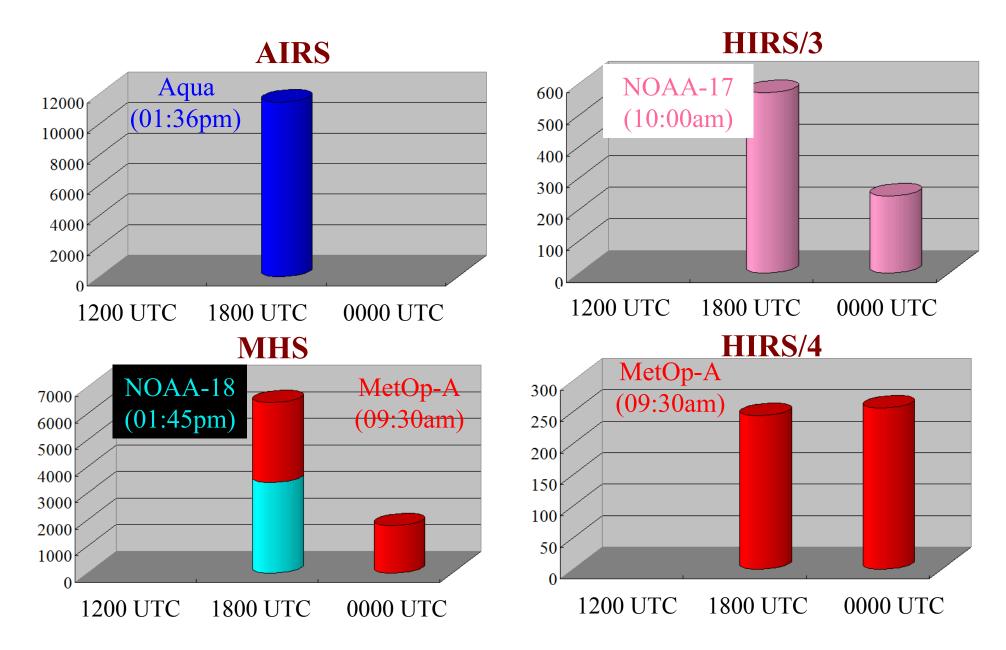


Satellite	Instruments	Satellite	Instruments	Satellite	Instruments
NOAA-14	[HIRS/2] ⁽¹⁾		HIRS/4	GOES-11	(SNDR)
	[MSU]	M-40 A	AMSU-A		Imager
NOAA-15	AMSU-A	MetOp-A	MHS		SNDRD1
	AMSU-B		[IASI]		SNDRD2
NOAA-16	(HIRS/3) ⁽²⁾		AIRS		SNDRD3
	(AMSU-A)	Aqua	(AMSU-A)		SNDRD4
	AMSU-B		(AMSRE)	GOES-12	(SNDR)
	(AVHRR3)	F13	(SSMI)		Imager
NOAA-17	HIRS/3	F14	(SSMI)		SNDRD1
	(AMSU-A)	F15	(SSMI)		SNDRD2
	AMSU-B	F16	(SSMIS)		SNDRD3
	(AVHRR3)				SNDRD4
NOAA-18	(HIRS/4)				(SNDR)
	AMSU-A				(Imager)
	MHS			COES 12	(SNDRD1)
	(AVHRR3)			GOES-13	(SNDRD2)
⁽¹⁾ Data not ava	uilable for this case.		(SNDRD3)		
⁽²⁾ Instruments	removed from ope		(SNDRD4)		

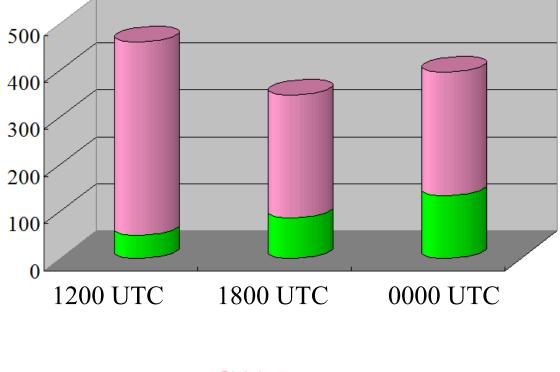
Total Data Count



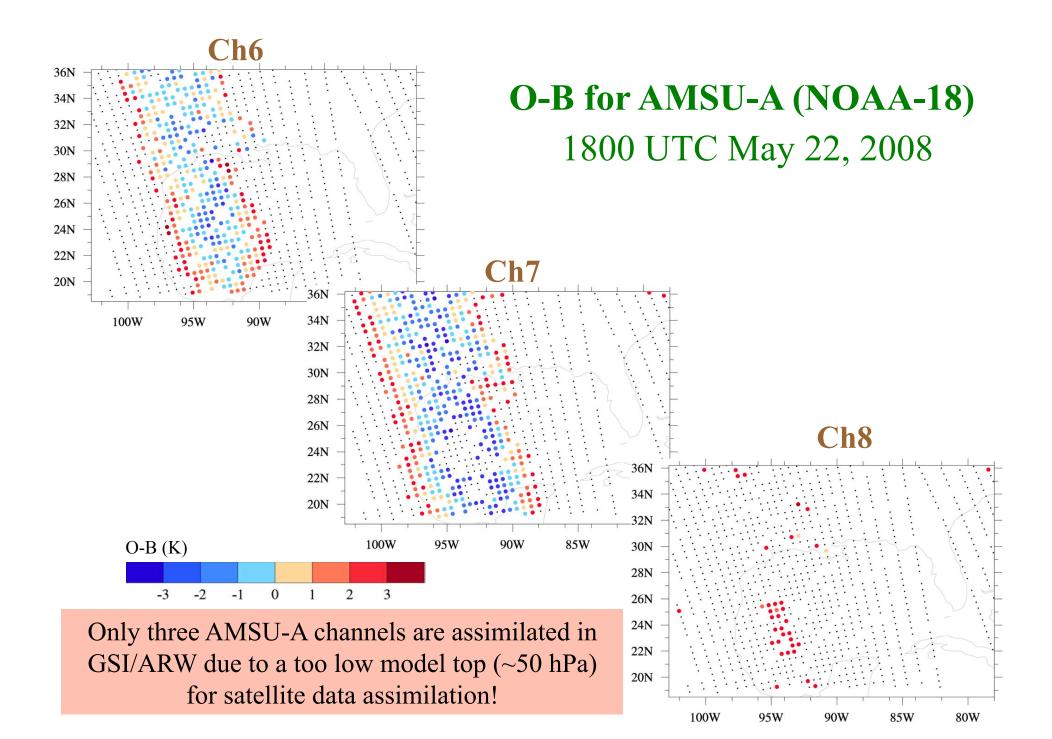
UTC Dependence of POES Data Count



GOES Imager Data Count



G11-Imager G12-Imager



Observed BT of GOES-11 Ch5 on May 23, 2008

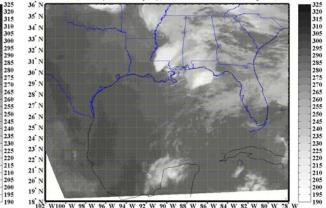
0300-0306 UTC

GOES-11,ch.5 (λ = 12.0 μm), 0300-0306 UTC, May 23,2008

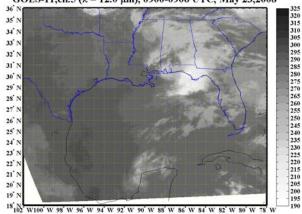
36 35 34 33 1 32° M 31'1 30 29'1 28 N 27 1 250 26 25 -240 235 24 230 23 1 220 22'1 -215 21 -210 20 19 18 N W100 W 98 W 96 W 94 W 92 W 90 W 84 W 82 W

0600-0606 UTC

GOES-11,ch.5 (λ = 12.0 µm), 0600-0606 UTC, May 23,2008



0900-0906 UTC GOES-11,ch.5 (λ = 12.0 μm), 0900-0906 UTC, May 23,2008



1200-1206 UTC GOES-11,ch.5 (λ = 12.0 μm), 1200-1206 UTC, May 23,2008

205

190

320

265 260

-255 250

-240

- 235

-230

- 225

220

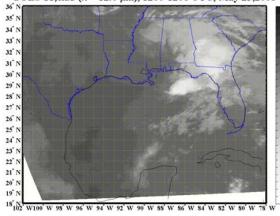
215

205

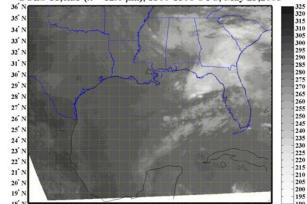
200 195

190

W 84 W 82 W 80 W



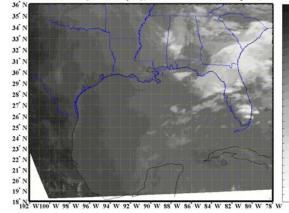
1500-1506 UTC GOES-11,ch.5 (λ = 12.0 μm), 1500-1506 UTC, May 23,2008



190

18 N W100 W 98 W 96 W 94 W 92 W 90 W 88 86 W 84 W 82 W 80

1800-1806 UTC GOES-11,ch.5 (λ = 12.0 μm), 1800-1806 UTC, May 23,2008



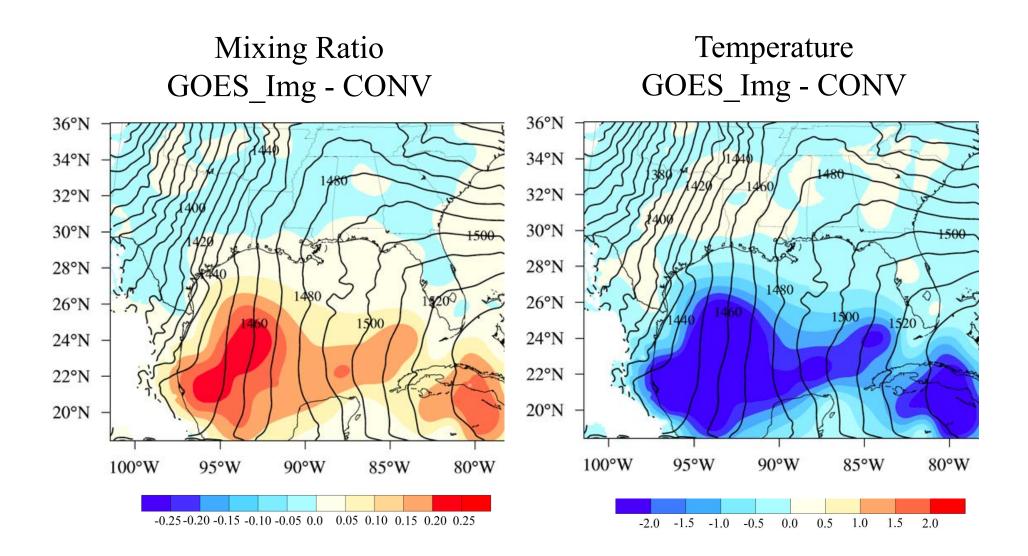
320 315

190

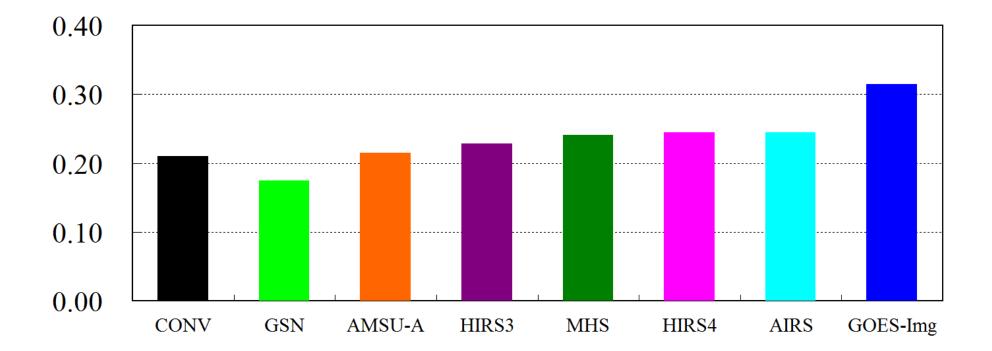
320 315

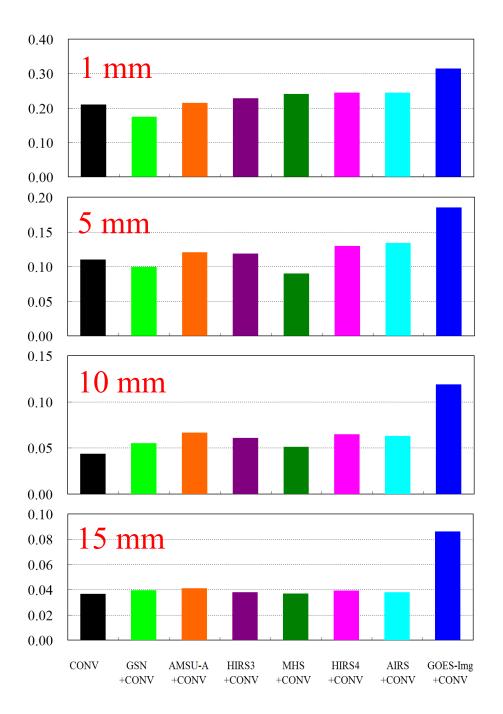
190

12-h Forecast Differences at 850 hPa 1200 UTC May 23 2008



Threat Scores of 3-h Accumulative Rainfall at 1mm thresholds Averaged over 24 Hours

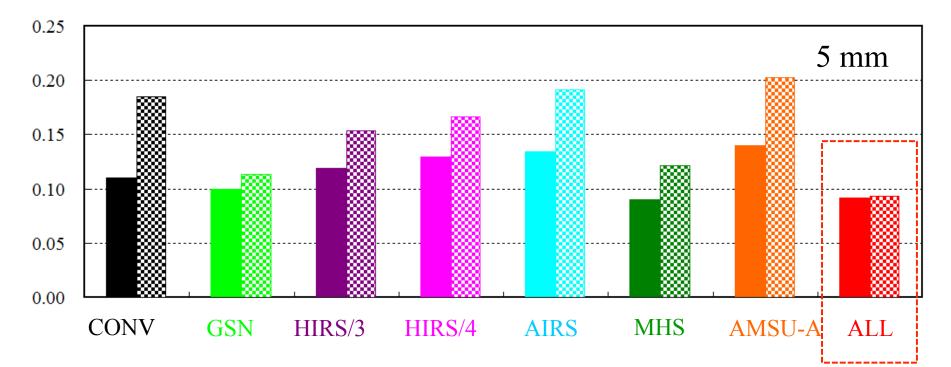




Threat Scores of 3-h Accumulative Rainfall Averaged over 24 Hours

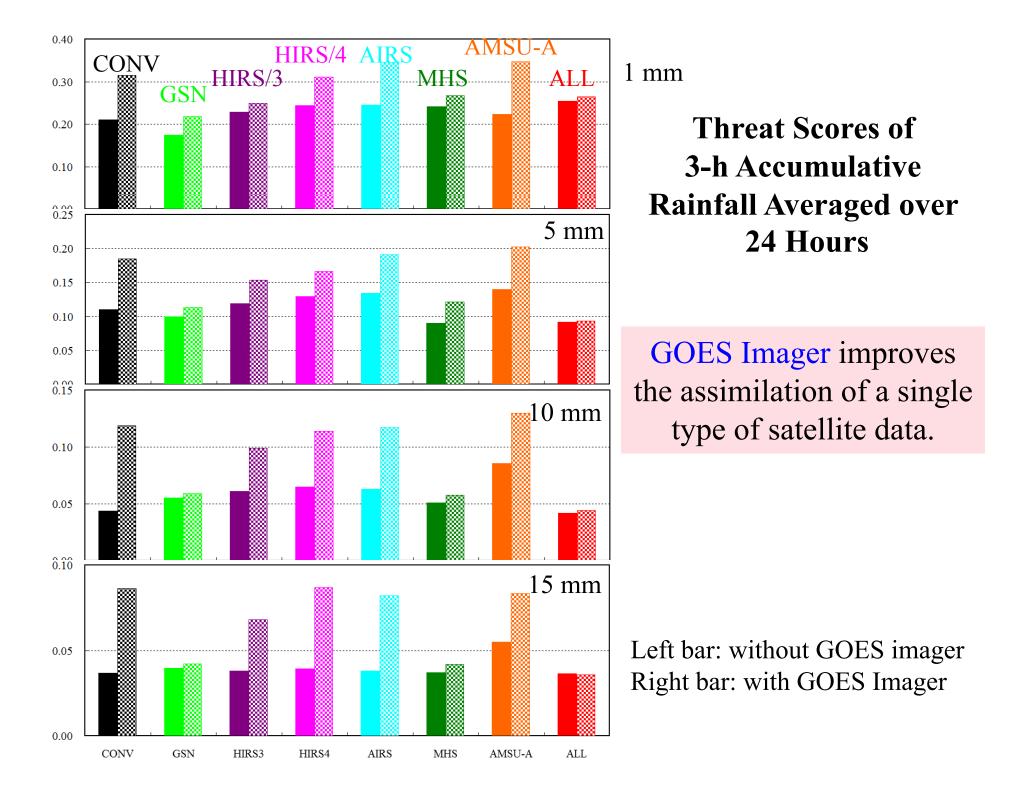
Assimilation of a Single Type of Satellite Observations

Threat Scores of 3-h Accumulative Rainfall at 5mm thresholds Averaged over 24 Hours

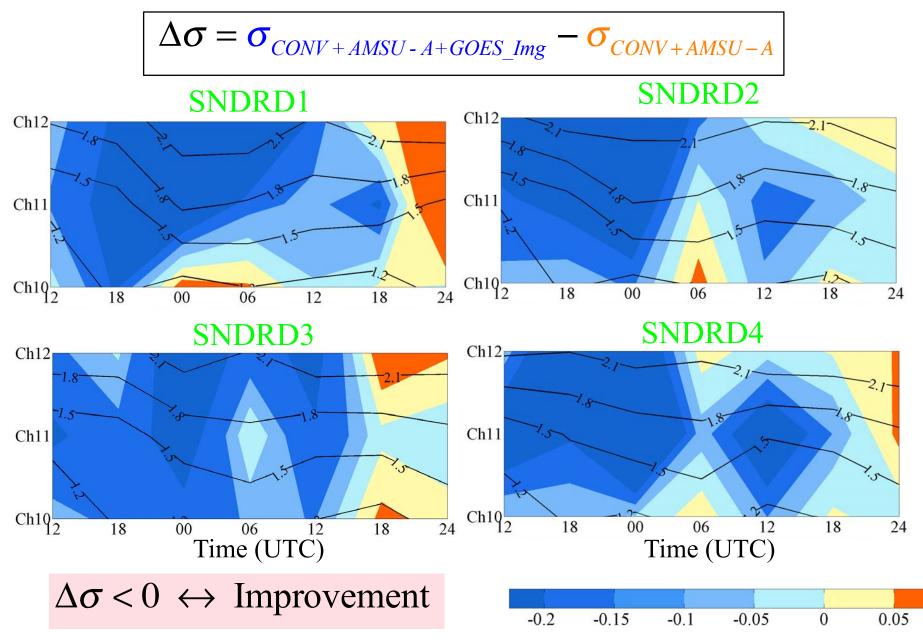


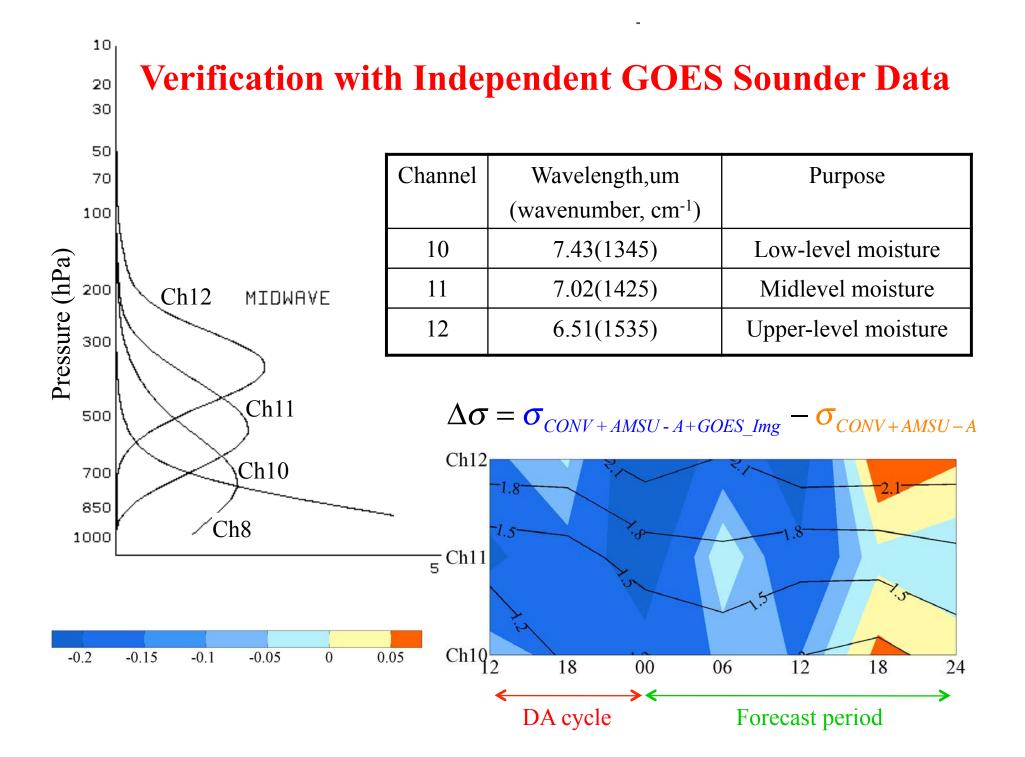
Left bar: without GOES Imager data

Right bar: with GOES Imager data

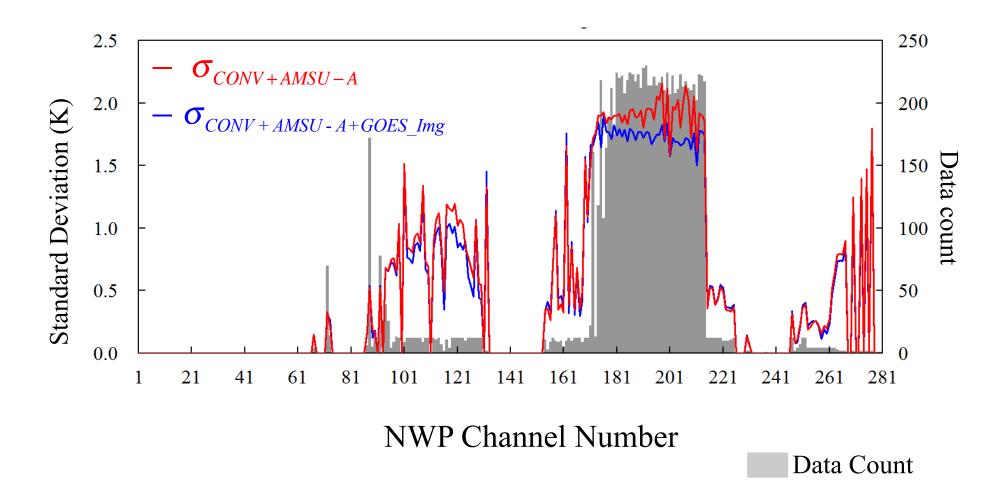


Verification with GOES Sounder

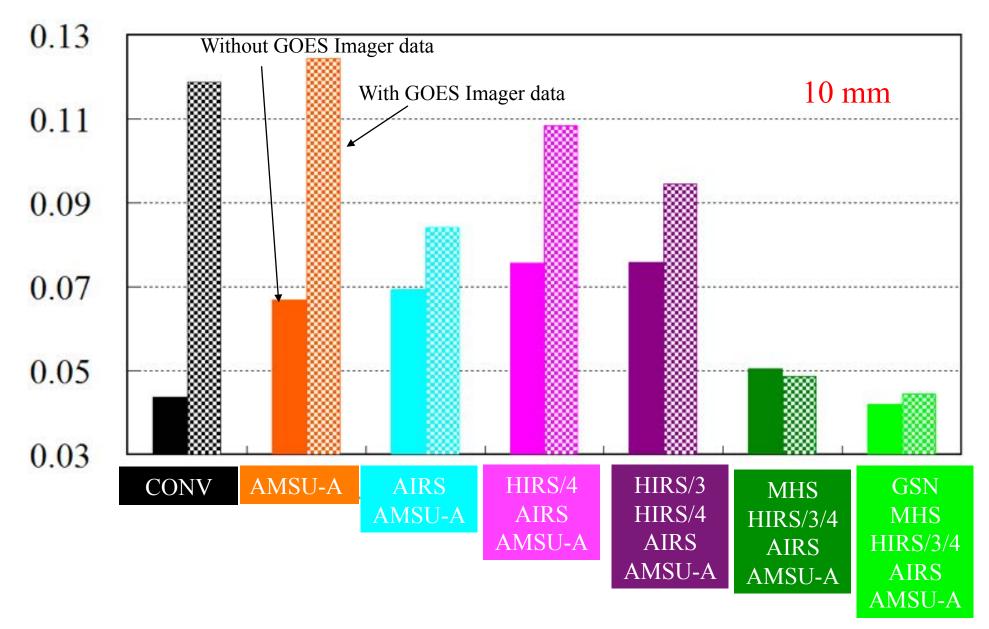




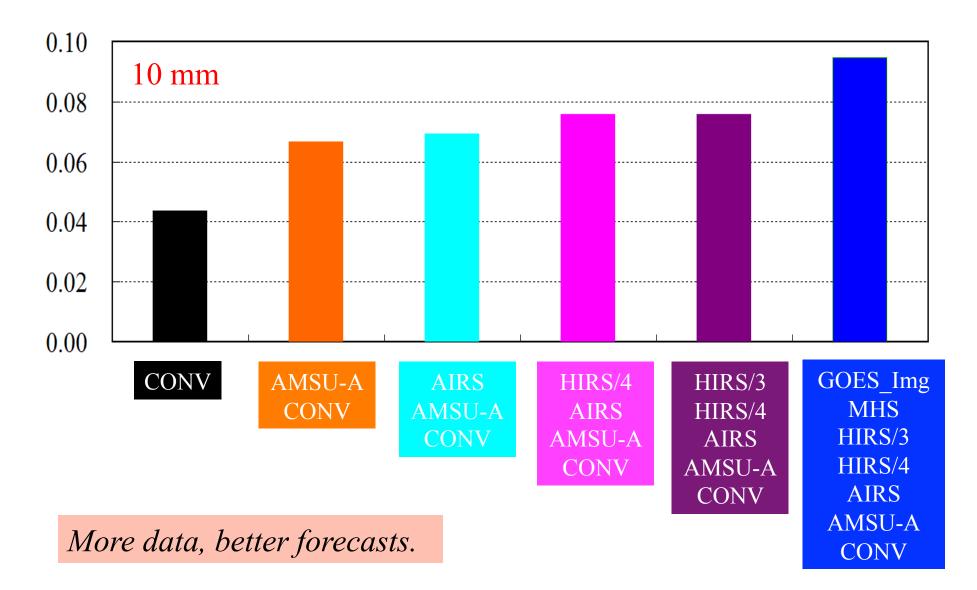
12-h Forecast Verification with AIRS



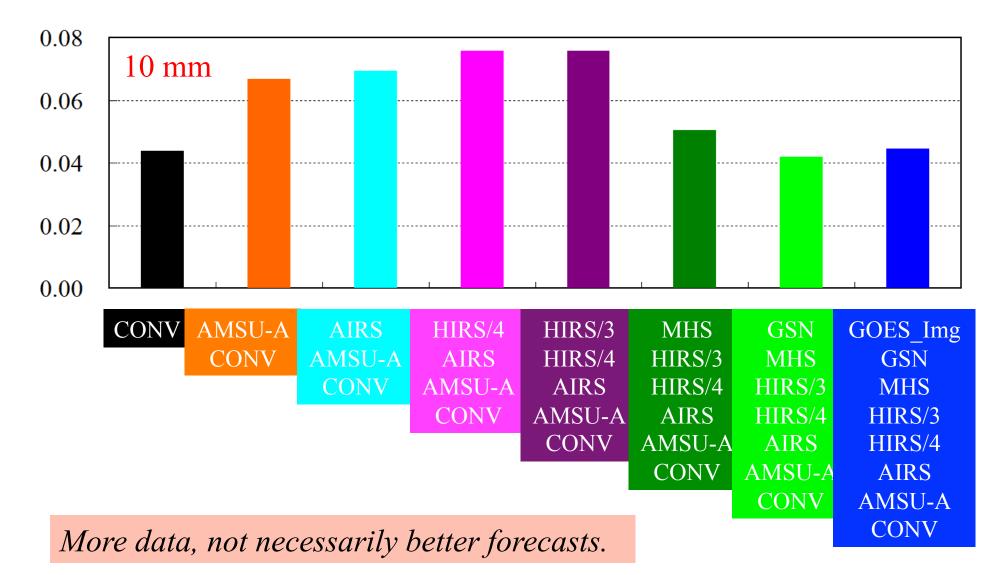
Threat Scores of 3-h Accumulative Rainfall



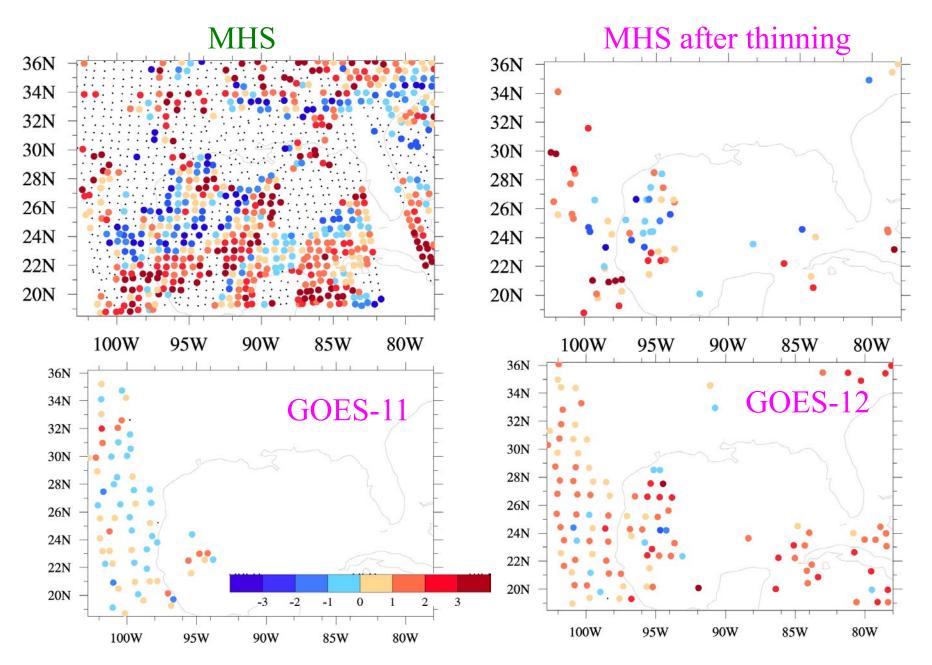
Threat Scores of 3-h Accumulative Rainfall



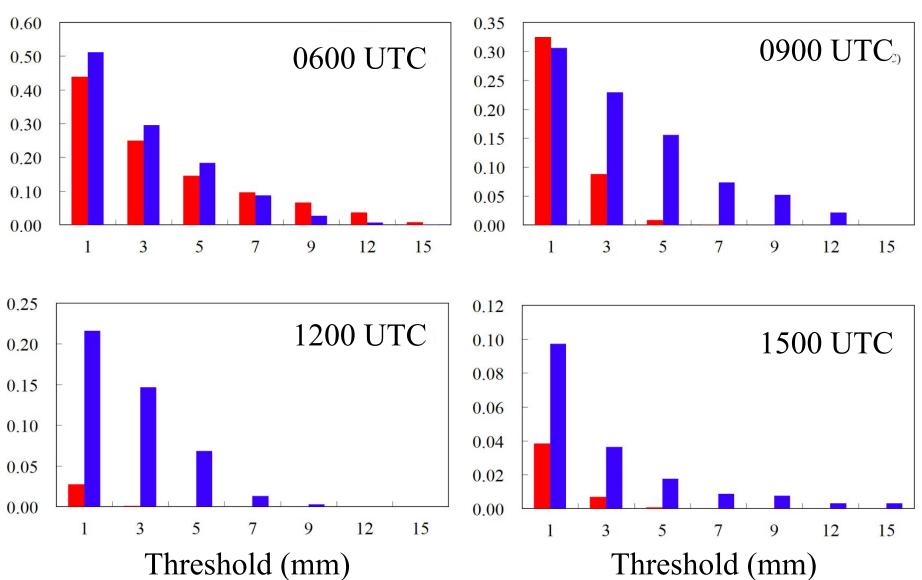
Threat Scores of 3-h Accumulative Rainfall



O-B (MHS Channel 3 at 1800 UTC 05/22/08)



Threat Scores (May 23, 2008)

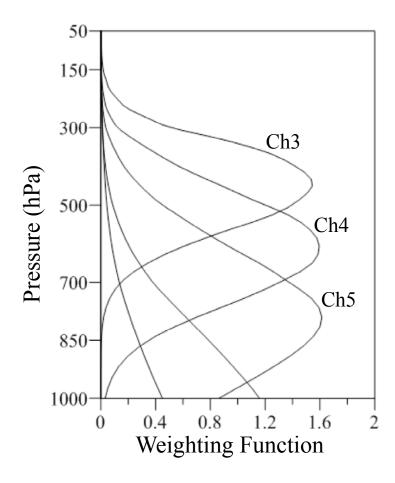


MHS MHS collocated with GOES

Part II

Improved QPFs by MHS Radiance Data Assimilation with a Newly Added Cloud Detection Algorithm

MHS Data Quality Control (QC)



- ✓ MHS QC in GSI and GSI QC results
- ✓ A new MHS QC for cloud detection
- ✓ Impact of the modification of MHS QC to QPFs

LWP Index Used for MHS QC in GSI

Over Ocean:

$$LWP_{index}^{ocean} = \begin{cases} 0.13 \times \left\{ \left(T_{b,1}^{o} - T_{b,1}^{m}\right) - 33.58 \times \frac{\left(T_{b,2}^{o} - T_{b,2}^{m}\right)}{300 - T_{b,2}^{o}} \right\}, & \text{if } T_{b,2}^{o} \le 300 \\ 9, & \text{otherwise} \end{cases}$$

Over Land:

$$LWP_{index}^{land} = 0.85 \times \left(T_{b,1}^{o} - T_{b,1}^{m}\right) - \left(T_{b,2}^{o} - T_{b,2}^{m}\right)$$

$$\left. \begin{array}{c} T_{b,1}^{o} - T_{b,1}^{m} \\ T_{b,2}^{o} - T_{b,2}^{m} \end{array} \right\} \text{ O-B differences of MHS channels 1-2}$$

*TPW*_{index} Three Steps for MHS Data Rejection in GSI

Step I:

$$TPW_{index} \equiv \left\{ \left[\left(T_{b,1}^{o} - T_{b,1}^{m} \right) - 7.5 \times LWP_{index} \right] / 10.0 \right\}^{2} + LWP_{index}^{2} > 1$$

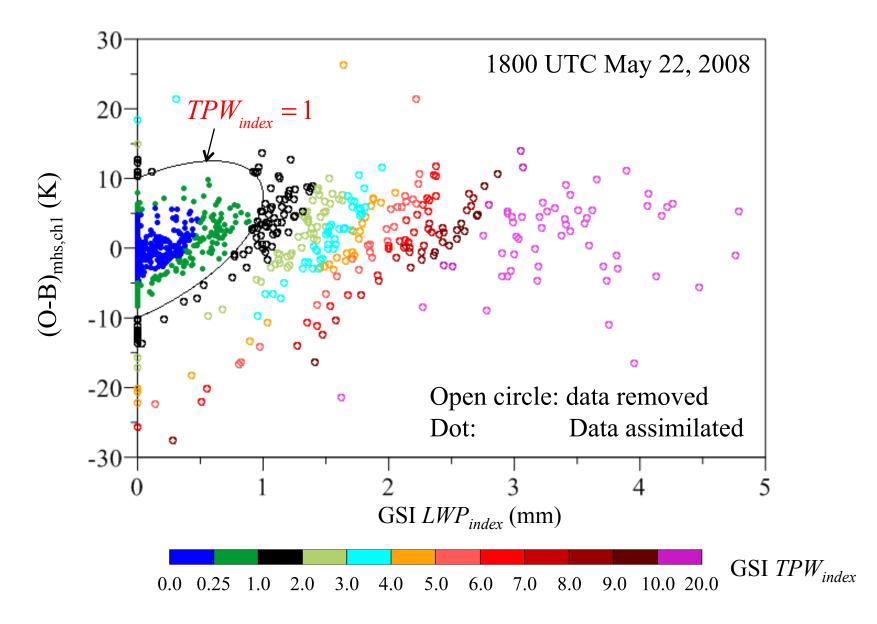
Step II:

$$|O - B| > 3\left(e_i \times \left(1 - TPW_{index}^2\right) \times f_H \times \tau_i^{top}\right)$$
or: $|O - B| > 6K$
 $e_i \text{ is accuracy of obs.}$
 $f_H = 2000/H, H \text{ is terrain height} > 2km$
 $\tau_i^{top} \text{ is ransmittance at model top}$

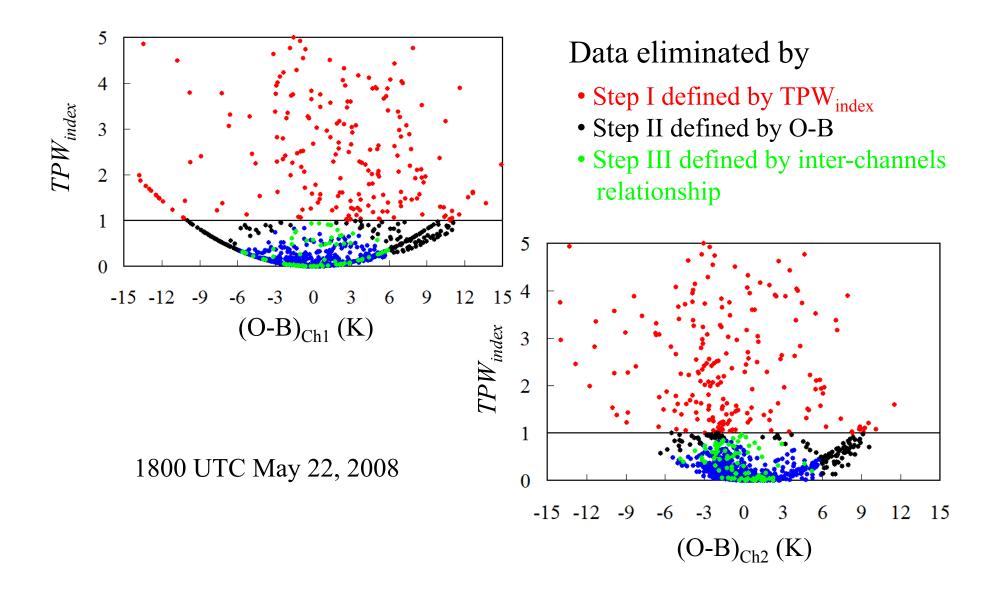
Step III:

All five channels if data of any other channel was removed by the first two QC steps

Diagnosis of MHS QC Results



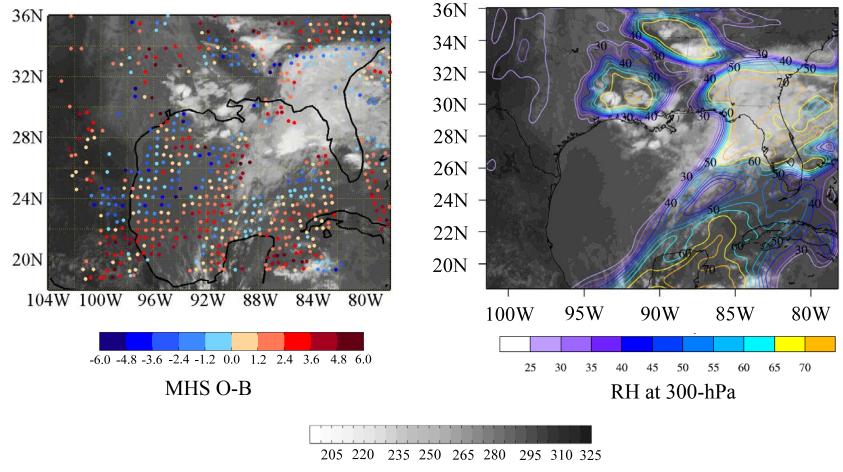
Diagnosis of MHS QC Results (cont.)



Diagnosis of MHS GSI QC

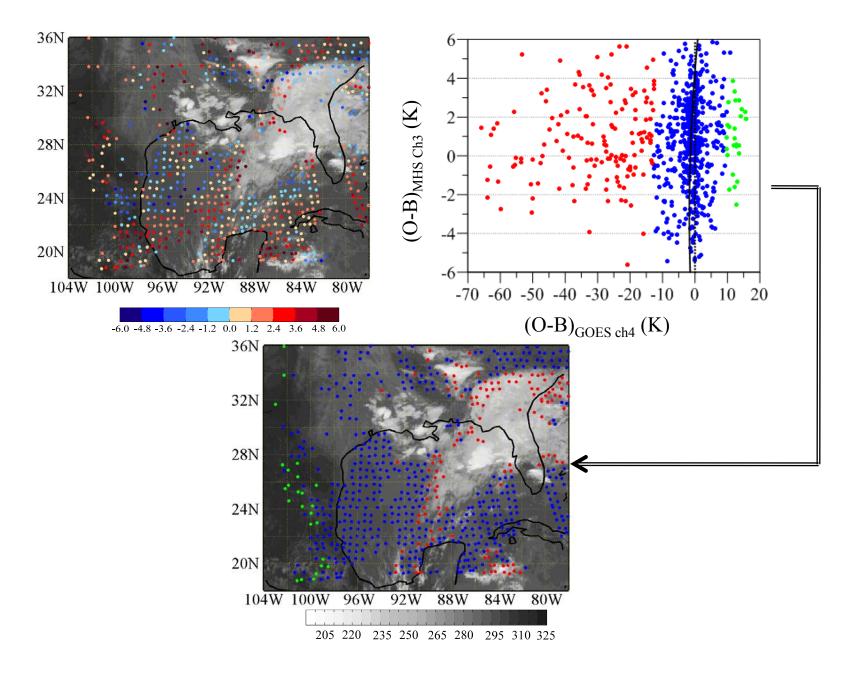
Modeled RH at 300 hPa

Data that pass GSI QC



GOES 10.7µm

Infrared O-B More Sensitive to Cloud Than Microwave



"(O-B)_{GOES}" Regressed by MHS Channels 1, 2 and 5

Over Ocean:

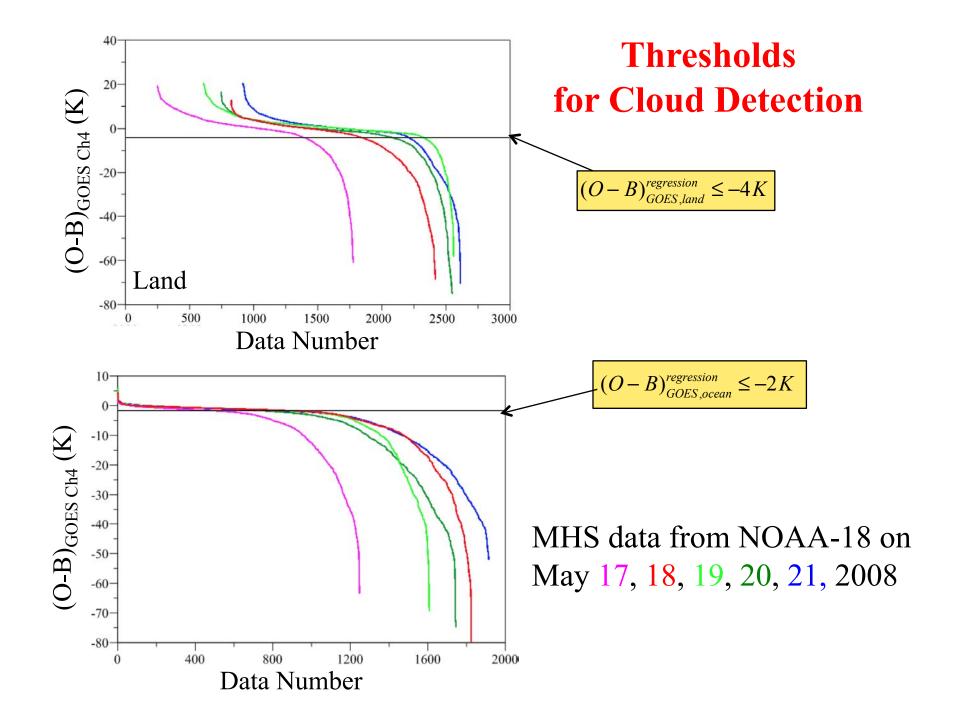
$$(O-B)_{GOES,ch4}^{regression} = -0.536 \times T_{b,MHS_{ch1}}^{obs} + 1.132 \times T_{b,MHS_{ch2}}^{obs} + 0.537 \times T_{b,MHS_{ch5}}^{obs} - 321.318$$

Over Land:

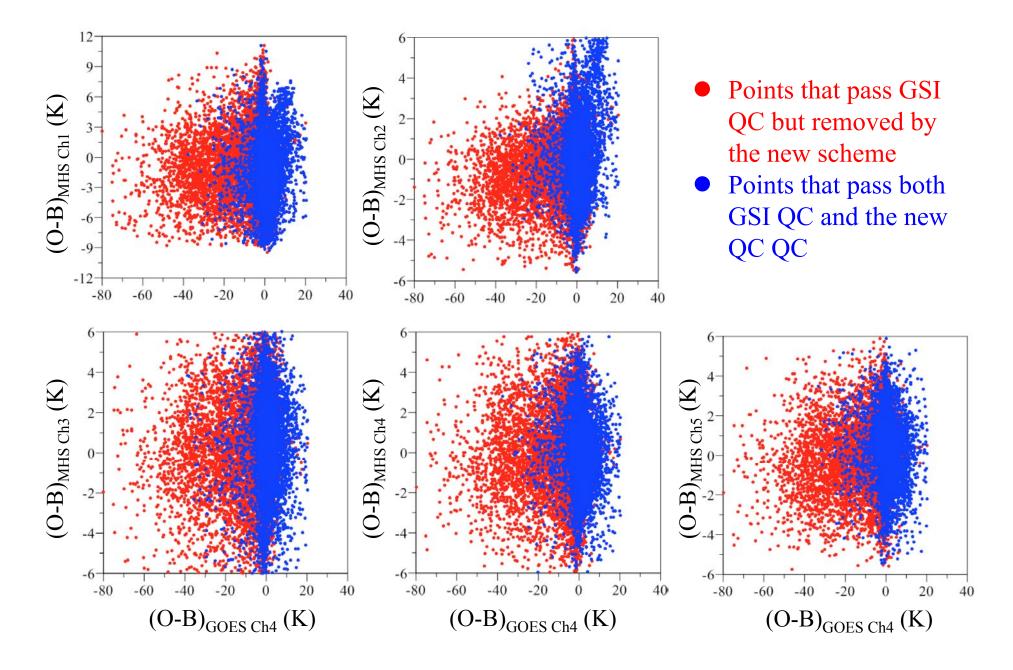
$$(O-B)_{GOES,ch4}^{regression} = 0.009 \times T_{b,MHS_{ch1}}^{obs} + 0.085 \times T_{b,MHS_{ch2}}^{obs} + 0.877 \times T_{b,MHS_{ch5}}^{obs} - 274.255$$

Observations of MHS channels 1-2, 5 are used in the regression.

- Channels 1-2 are affected by the radiation from both the Earth's surface and emission and scattering from ice phase clouds
- Channel 5 is most sensitive to scattering from thin clouds



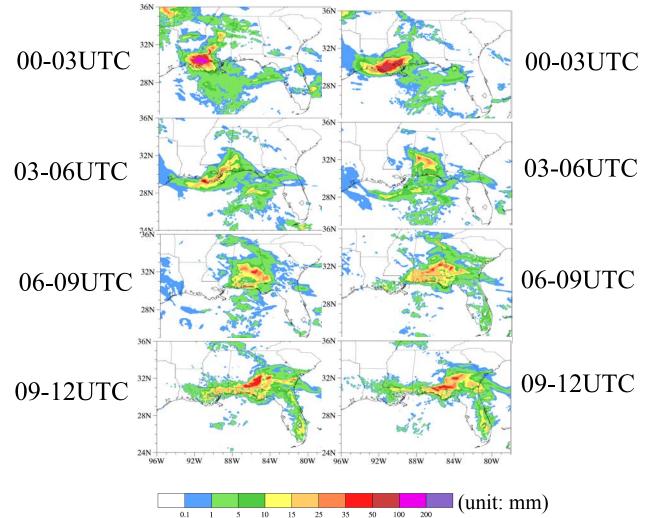
O-B Scatter Plots for MHS Channels 1-5 versus GOES-12 Ch4



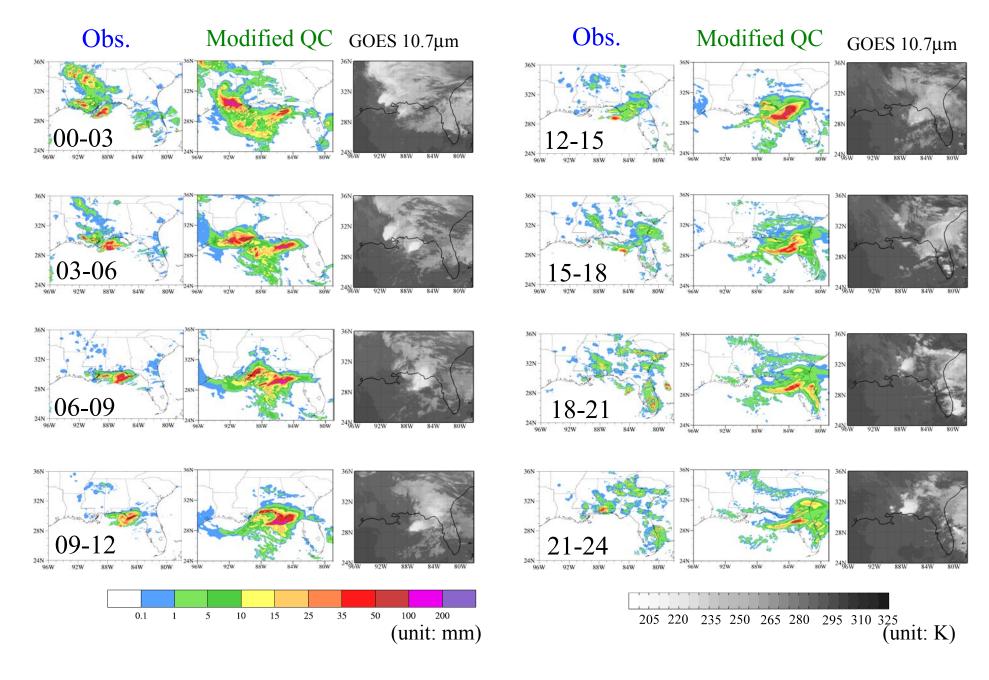
24-h QPFs of 3-h Accumulative Rainfall

EXP1: The original GSI QC for MHS data is used.

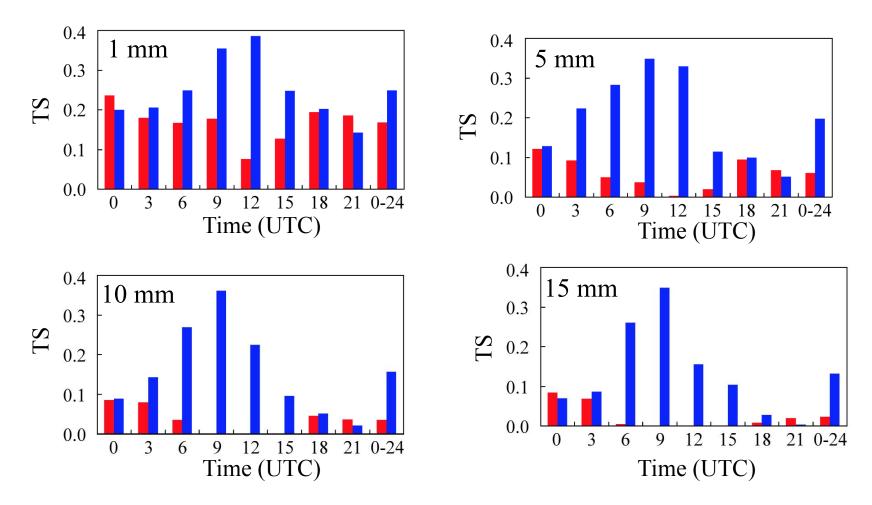
May 23, 2008



24-h Forecasts of 3-h Accumulative Rainfall



Threat scores (TS) of 3-hour Accumulative Rainfall

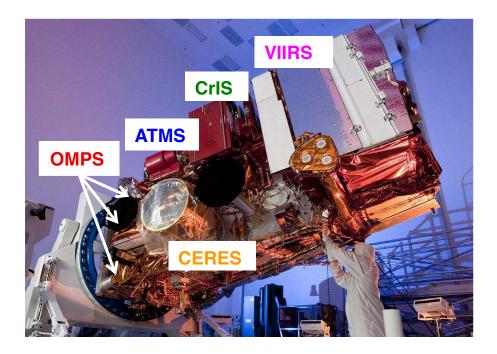


EXP1:CONV+AMSU-A+MHS EXP2:Same as EXP1 except for modified MHS QC

Part III

Impacts of ATMS Data Assimilation on Hurricane Track and Intensity Forecasts

Suomi NPP Satellite Instruments



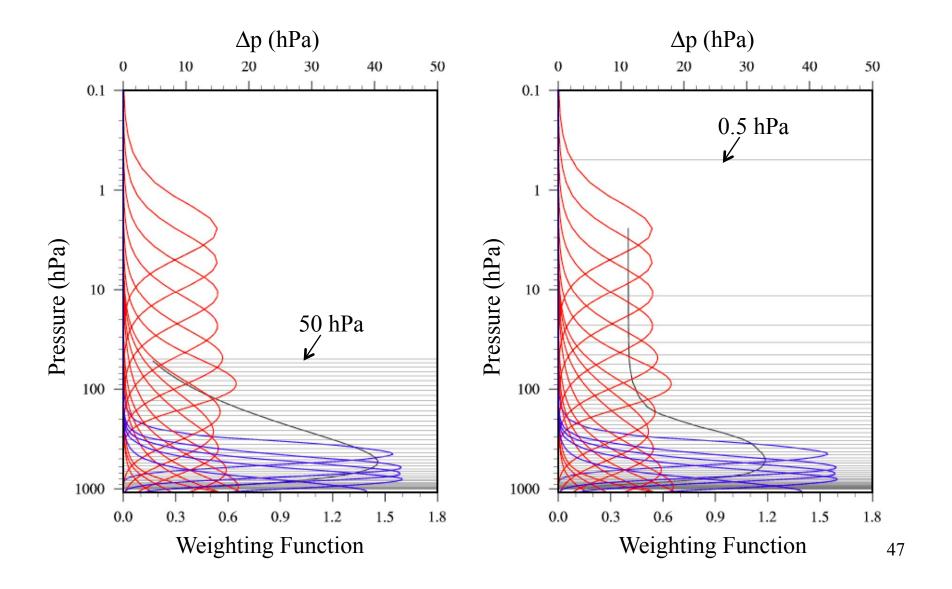
- ATMS ---- Advanced Technology Microwave Sounder
- CrIS --- Cross-track Infrared Sounder
- VIIRS --- Visible/Infrared Imager/Radiometer Suite
- **OMPS** --- Ozone Mapping and Profiler Suite

CERES --- Cloud and Earth Radiant Energy System

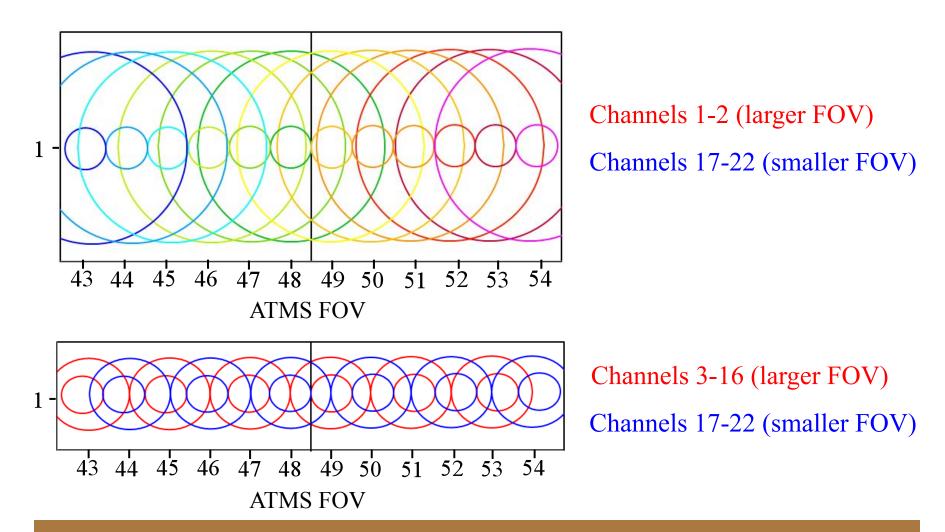
Channel Characteristics of ATMS and AMSU

Channel		Frequency (GHz)		ΝΕΔΤ (Κ)		Beam width (°)		Peak WF (hPa)	
ATMS	AMSU	ATMS	AMSU	ATMS	AMSU	ATMS	AMSU	ATMS	AMSU
1		23.8		0.50	0.30	5.2	3.3	Surface	
2		31.4	31.399	0.60	0.30	5.2	3.3	Surface	
3		50.3	50.299	0.70	0.40	2.2	3.3	Surface	
4		51.76		0.50		2.2		Surface	
5	4			0.50	0.25	2.2	3.3	1000	
6		5 53.596±0.115		0.50	0.25	2.2	3.3	700	
7	6	54.4		0.50	0.25	2.2	3.3	400	
8	7	54.94		0.50	0.25	2.2	3.3	270	
9	8	55.5		0.50	0.25	2.2	3.3	180	
10	9	57.29		0.75	0.25	2.2	3.3	90	
11	10	57.29± 0.217		1.00	0.40	2.2	3.3	50	
12	11	$57.29 \pm 0.322 \pm 0.048$		1.00	0.40	2.2	3.3	25	
13	12	$57.29 \pm 0.322 \pm 0.022$		1.25	0.60	2.2	3.3	12	
14	13	$57.29 \pm 0.322 \pm 0.010$		2.20	0.80	2.2	3.3	5	
15	14	57.29± 0.322± 0.0045		3.60	1.20	2.2	3.3	2	
16	15	88.2	89.0	0.30	0.50	2.2	3.3	Surface	
17	16	165.5	89.0	0.60	0.84	1.1	1.1	1000	Surface
18	17	183.31±7.0	157.0	0.80	0.84	1.1	1.1	800	Surface
19	18	183.31±4.5	183.31±1.0	0.80	0.60	1.1	1.1	700	400
20	19	183.31±3.0		0.80	0.70	1.1	1.1	600	
21	20	183.31±1.8	183.31±7.0	0.80	1.06	1.1	1.1	500	800
22		183.31±1.0		0.90		1.1		400	

ATMS Weighting Functions and HWRF Model Levels

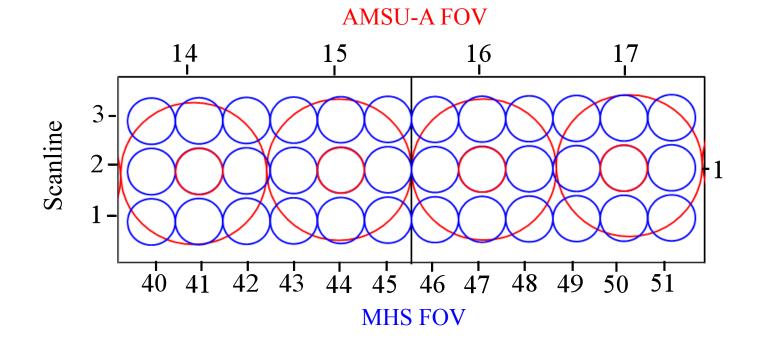


Same FOVs for All ATMS Channels



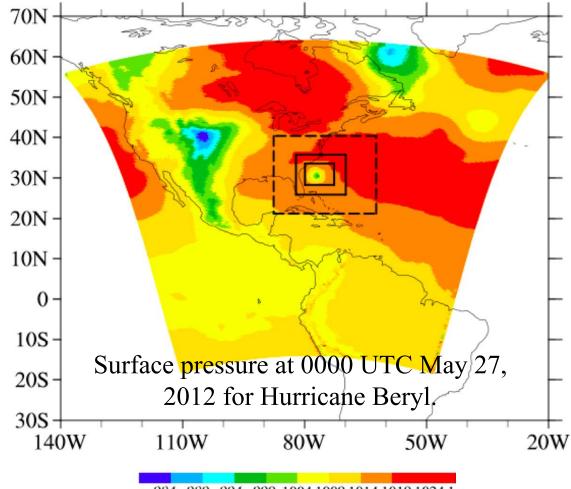
A consistent FOV distribution between temperature and humidity channels on ATMS makes the cloud detection easy to implement.

Inconsistent FOVs between AMSU-A and MHS



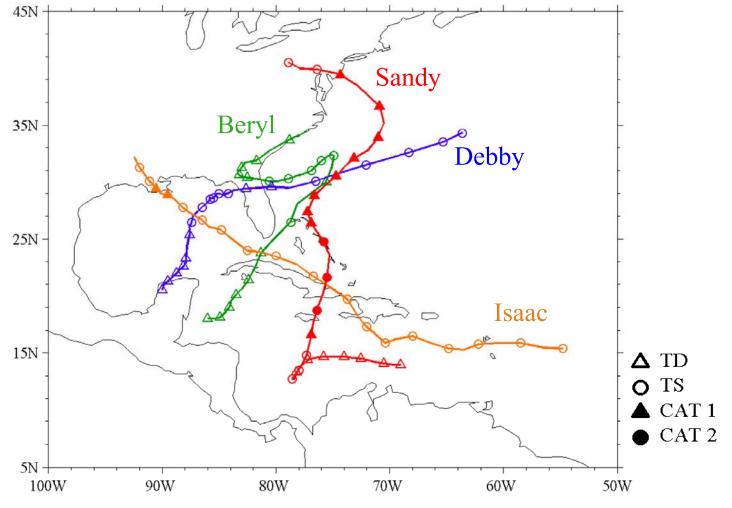
An inconsistent FOV distribution between AMSU-A and MHS channels makes the cloud detection for MHS data difficult.

The outer domain, ghost domain, middle nest and inner nest of HWRF

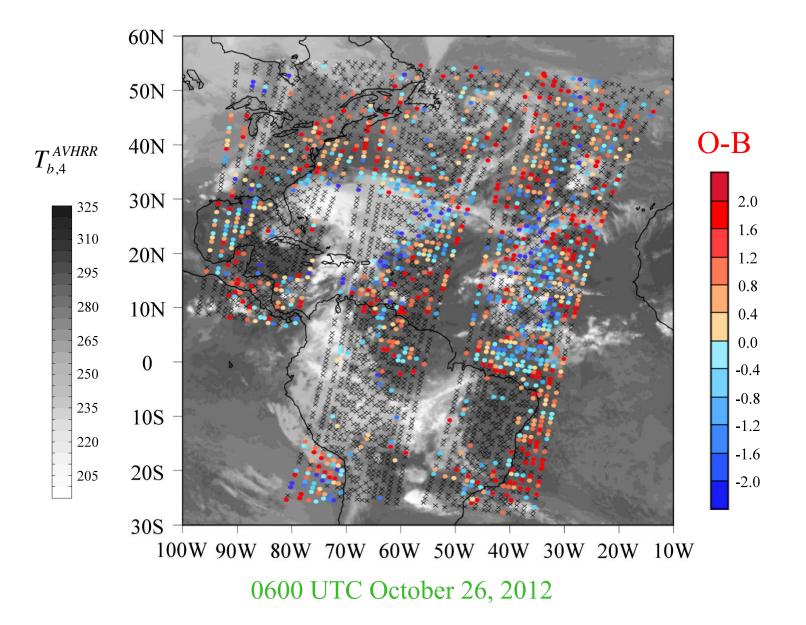


984 989 994 999 1004 1009 1014 1019 1024 1

Four 2012 Atlantic Hurricanes which Made Landfall

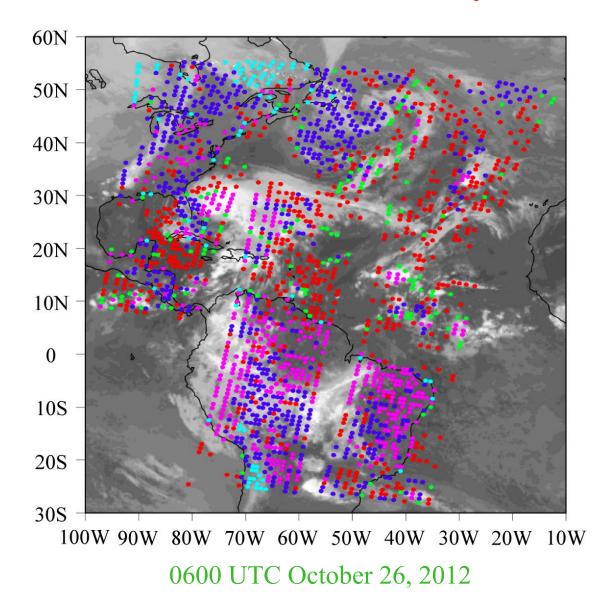


O-B Values for Those Data Points that Pass QC



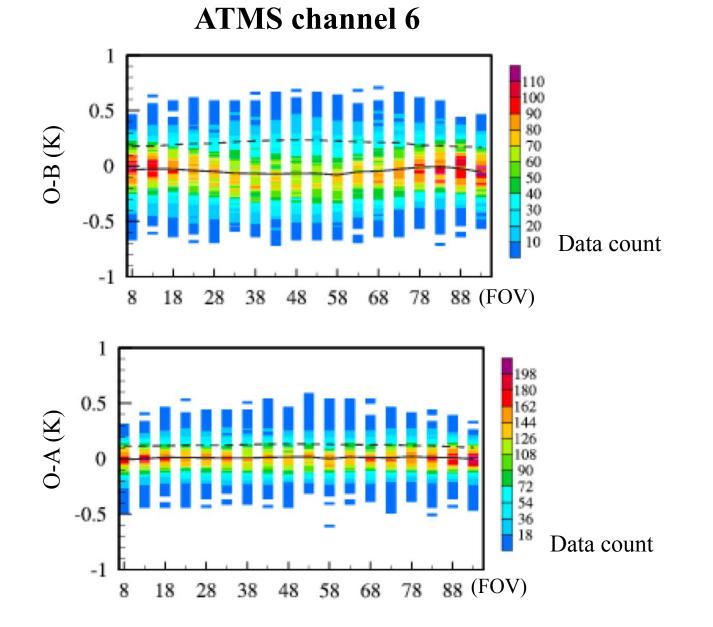
52

Data Points Removed by QC



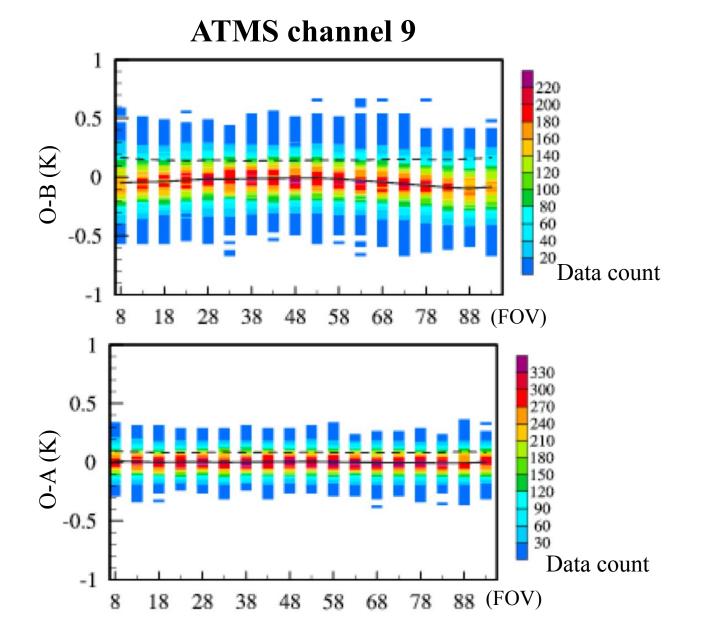
the 5th QC the 6th QC the 7th QC the 8th QC the 9th QC criteria

Convergence of ATMS Data Assimilation (Isaac)



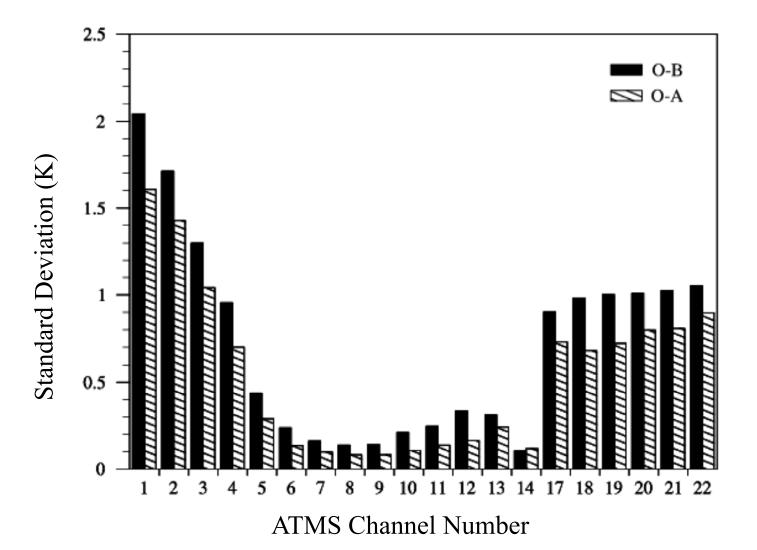
54

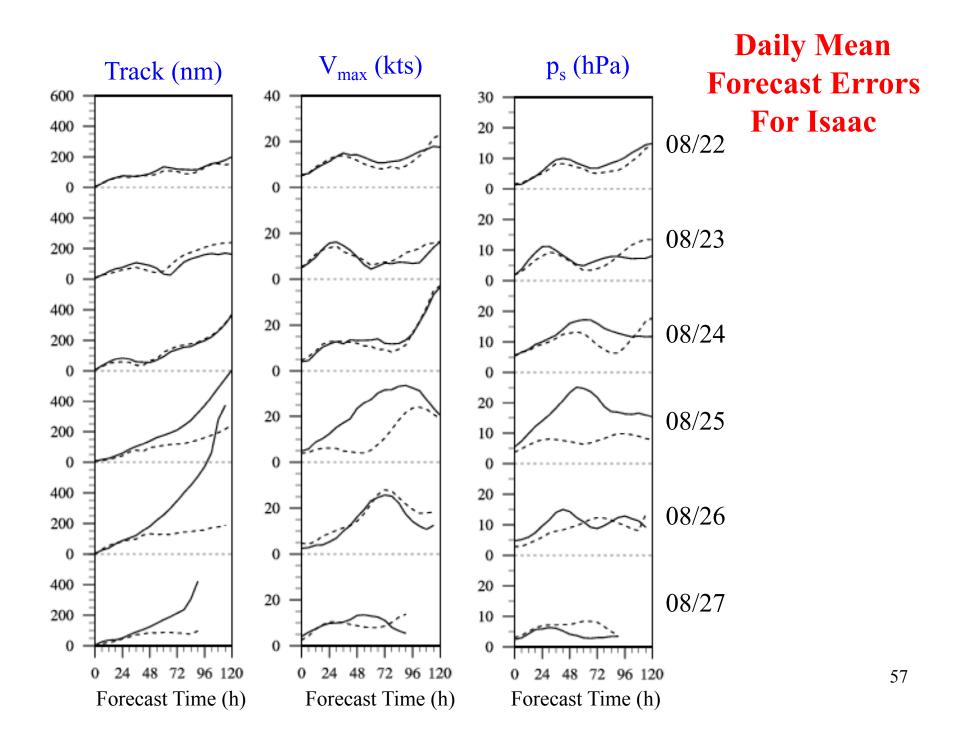
Convergence of ATMS Data Assimilation (Isaac)



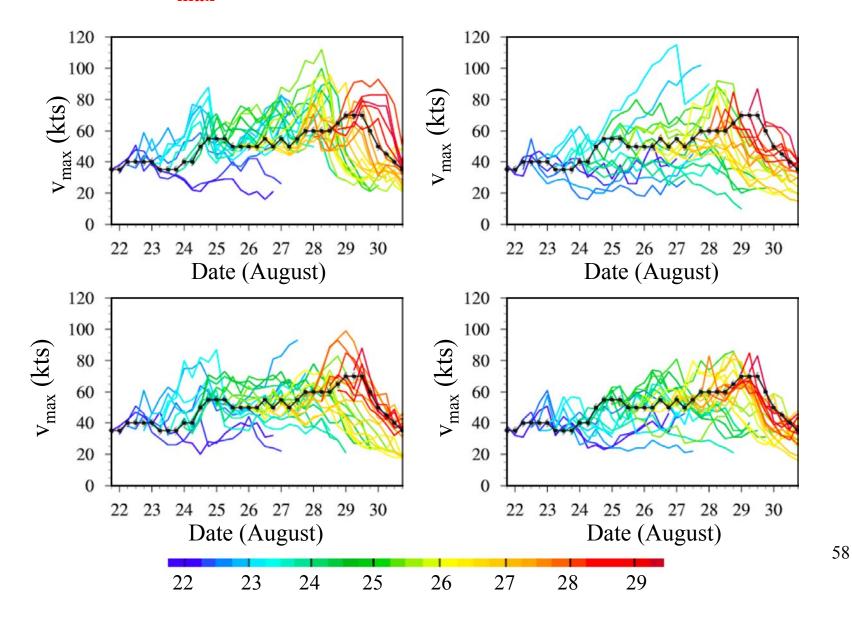
55

Standard Deviation before and after Data Assimilation For Hurricane Isaac





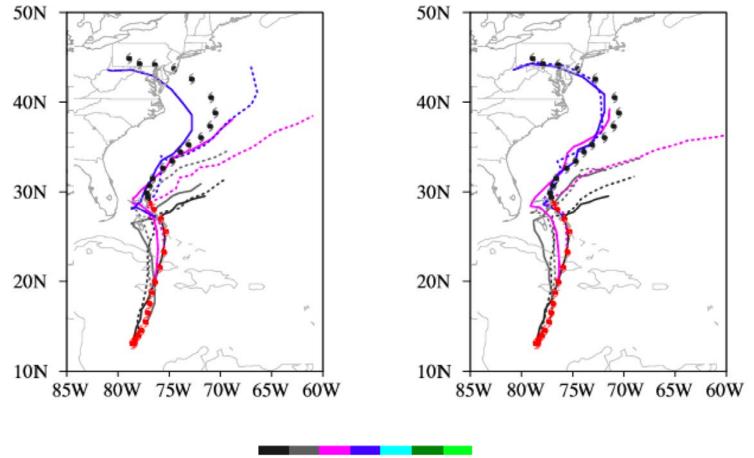
Impacts of Satellite Data Assimilation on v_{max} Forecast Errors for Hurricane Isaac



Impacts of ATMS Data Assimilation on the Track Forecast of Hurricane Sandy

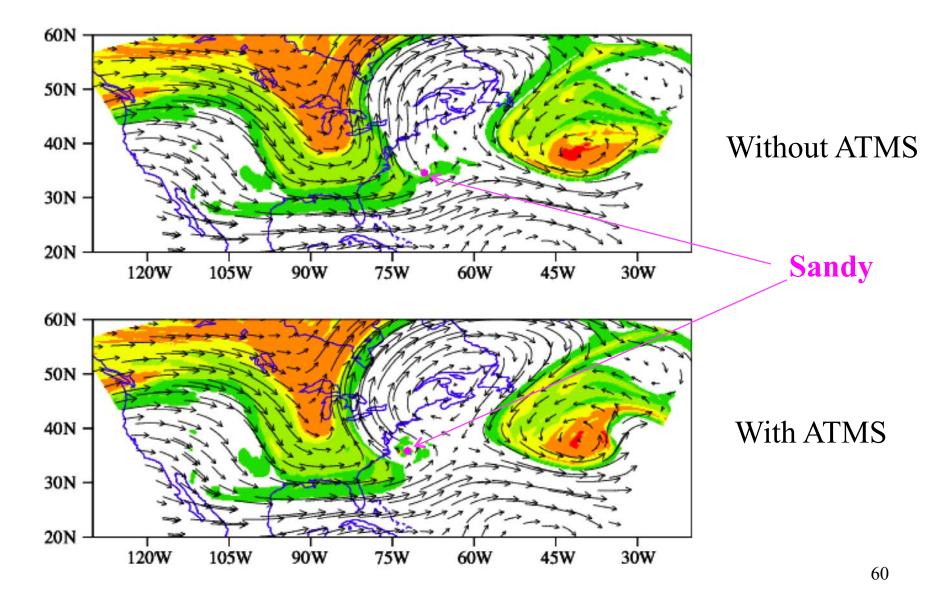
without ATMS

with ATMS

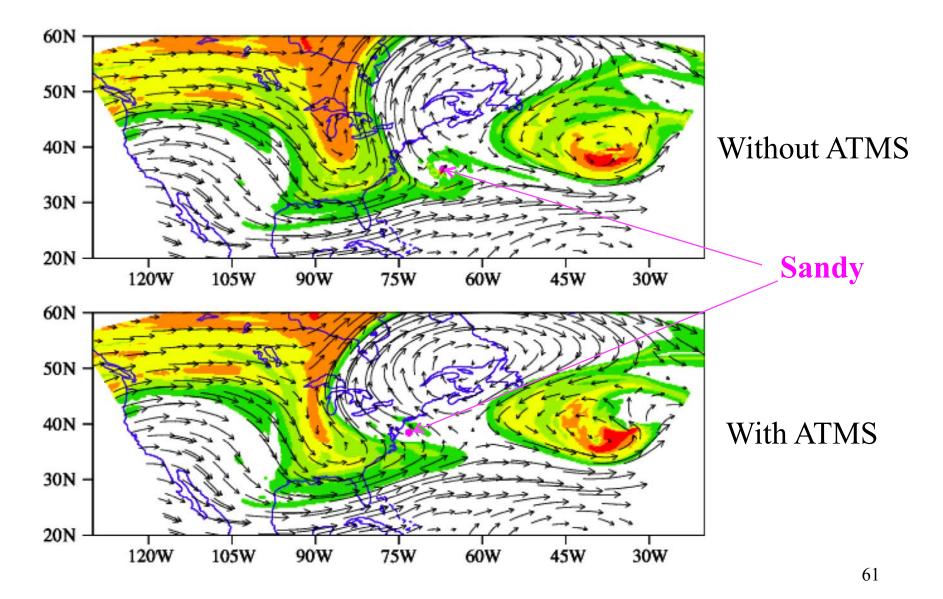


23 24 25 26 27 28 29 (October)

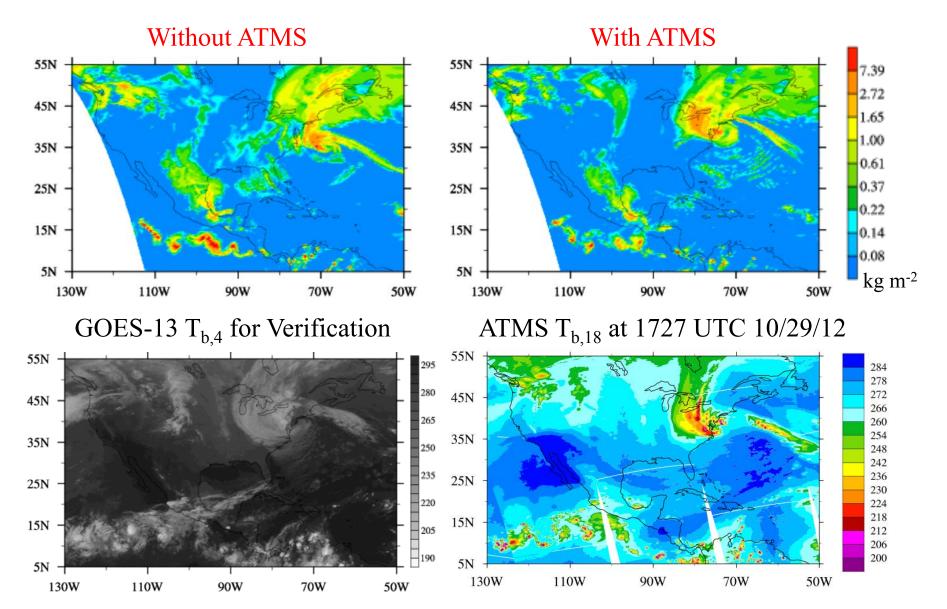
72-h Forecasts of PV and Wind Vector at 200 hPa



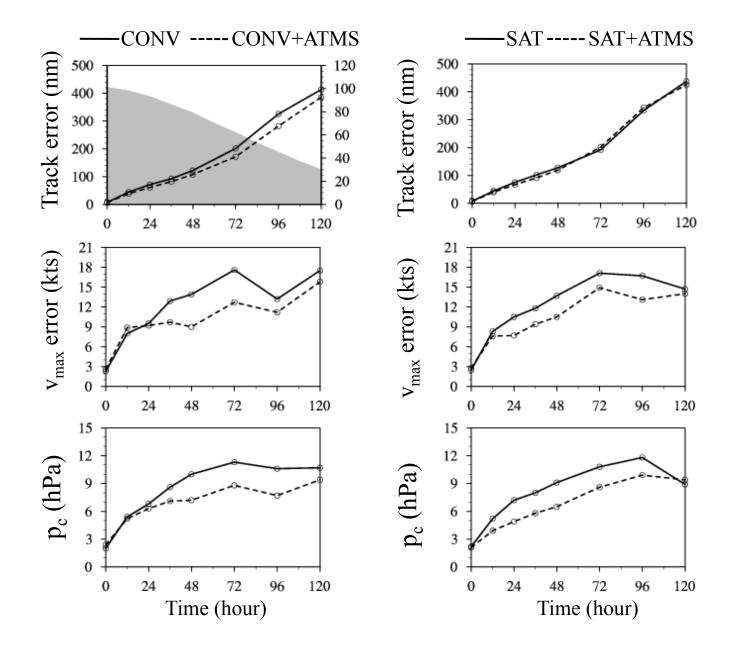
84-h Forecasts of PV and Wind Vector at 200 hPa



84-h Forecasts of Cloud Liquid Water Valid at 0000 UTC 30 October 2012



Mean Forecast Errors for 2012 Atlantic Landfalling



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Summary (Part I)

- AMSU-A and GOES imager data contribute most significantly to improved QPFs near Gulf of Mexico
- Assimilation of GOES imager radiances contributes positively to any single type of satellite data
- Assimilation of all types of satellite data in the GSI system did not produce a better forecast than any experiment assimilated a single type of satellite data
- An improved cloud detection for MHS observations results in a significant positive impact to coastal QPFs

Summary (Part II)

- Some cloudy radiances remain near cloud edges after the MHS QC in GSI
- The cloud detection algorithm effectively removes those cloudy radiances remaining near cloud edges after the MHS QC in GSI
- The MHS data assimilation with the revised QC is shown to significantly improve coastal QPFs

Summary (Part III)

- A consistent FOV distribution between temperature and humidity channels on ATMS makes the cloud detection easy to implement
- ATMS data assimilation in GSI/HWRF results in a consistent positive impact on the track and intensity forecasts of the four landfall hurricanes in 2012
- Hurricane Sandy's forecasts are significantly improved after ATMS data assimilation when verified with independent GOES and POES observations

More details can be found in

- Zou, X., Z. Qin, and F. Weng, 2012: Improved coastal precipitation forecasts with direct assimilation of GOES 11/12 imager radiances, Mon. Wea. Rev., **139**, 3711-3729.
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