



The HWRF Basin-Scale Data Assimilation and Ensemble Forecasting Project

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Introduction

Basin-scale HWRF DA and forecasting system:

- Pure ensemble-based analysis and prediction system for HWRF using DTC EnKF
- Built into HWRF rocoto framework—takes advantage of same community DA software as operational model
- No vortex relocation
- No dependence on GFS for initial conditions and satellite biases (after initial spin-up period)

Motivation for HWRF regional-scale DA

- 1 Tie together model and DA development at AOML/HRD
 - Perform vigorous evaluation of HWRF model through continuous cycling of model states over fixed domain
 - Explore regional biases using observation-space diagnostics
- 2 Turn basin-scale HWRF into a regional ensemble forecasting platform
 - Basin-scale HWRF ensemble has higher resolution than GFS ensemble
 - Adopting own DA system allows control over how ensemble is created and calibrated

Motivation for HWRF regional-scale DA

- 3 Establish a framework for satellite DA research
 - Initial tests examine impact of online bias correction for clear-sky radiance
 - Future research will focus on all-sky radiance DA

- 4 Build a testbed for developing new DA strategies with community software
 - Localized particle filter (Poterjoy 2016) will be tested using pre-existing DA software from DTC
 - All developments are directly applicable to operation models (GFS, FV3, etc.)

HFIP Real-Time Demo Project

Perform continuous 3-month ensemble analysis of August, September, and October, following spin-up from July.

Establish a benchmark for future DA and ensemble forecasting research and model development.

Configuration:

- 60-member 6-h cycling DA for reduced basin-scale parent domain (18-km grid spacing).
- Lateral and surface boundaries provided from GFS ensemble.
- All available conventional and clear-air satellite measurements assimilated (no TC vitals).

HFIP Real-Time Demo Project

Perform continuous 3-month ensemble analysis of August, September, and October, following spin-up from July.

Establish a benchmark for future DA and ensemble forecasting research and model development.

Configuration:

- Conventional obs thinned to model grid spacing.
- Satellite obs thinned using default HWRF options.
- Between 1.5M – 2.8M observations assimilated every 6 h.
- Radiance biases estimated online through continuous cycling.

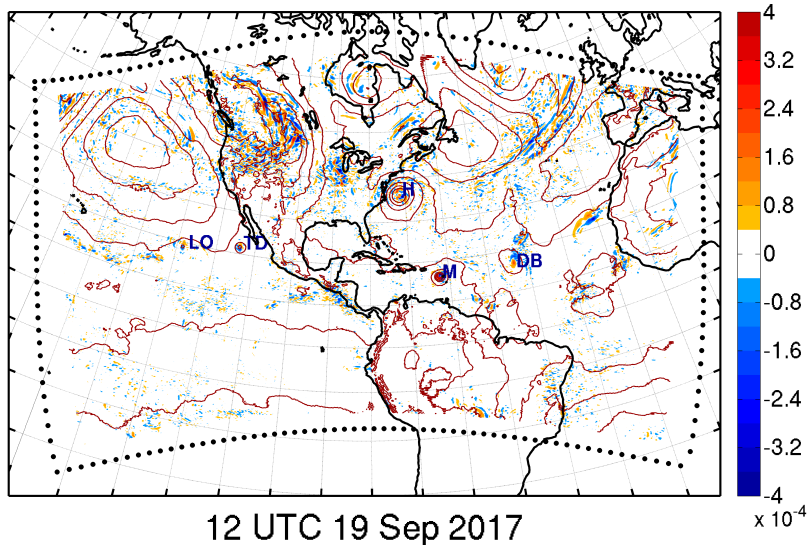
HFIP Real-Time Demo Project

System tuned based on July testing:

- 1200 KM localization ROI in NH and SH
- 1000 KM localization ROI in tropics
- 1 scale height vertical localization ROI for conventional observations
- 2 scale height vertical localization ROI for radiance
- 95% relaxation to prior spread (posterior variance inflation)
- Typical EnKF update takes between 30 – 45 min on 123 nodes of Jet.

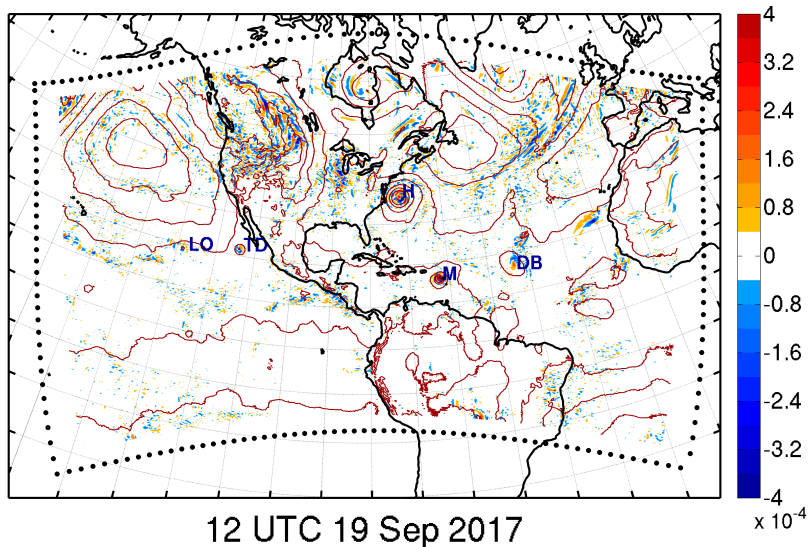
Posterior member 1

MSLP and vorticity increments at level 20 (~ 875 mb)



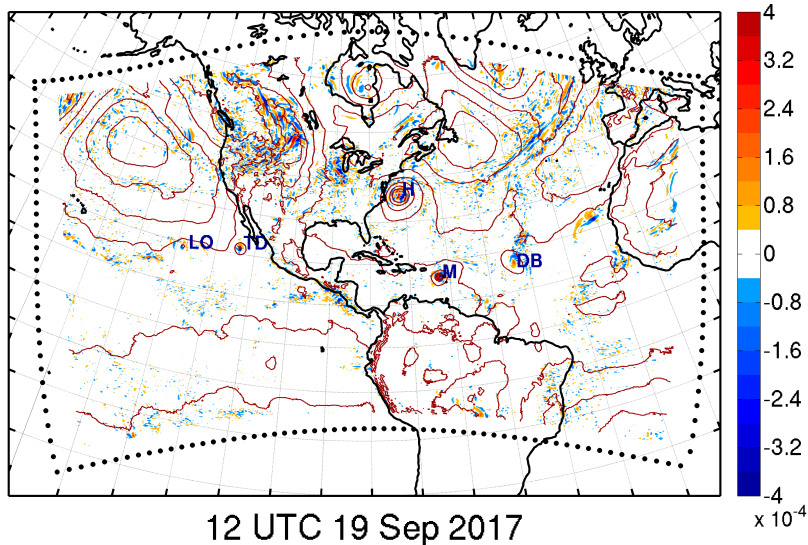
Posterior member 2

MSLP and vorticity increments at level 20 (~ 875 mb)



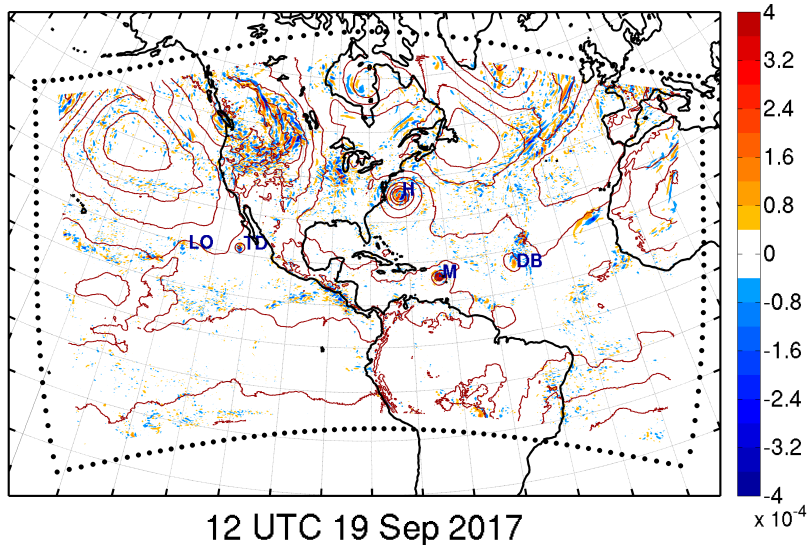
Posterior member 3

MSLP and vorticity increments at level 20 (~ 875 mb)



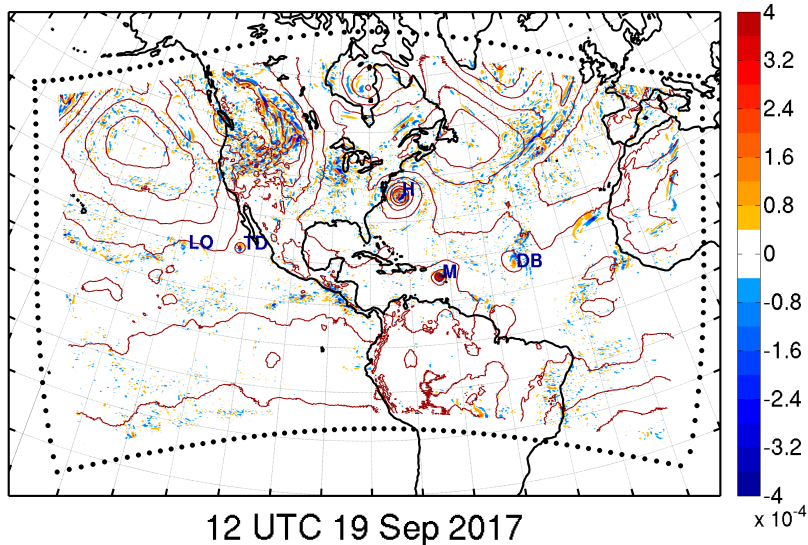
Posterior member 4

MSLP and vorticity increments at level 20 (~ 875 mb)



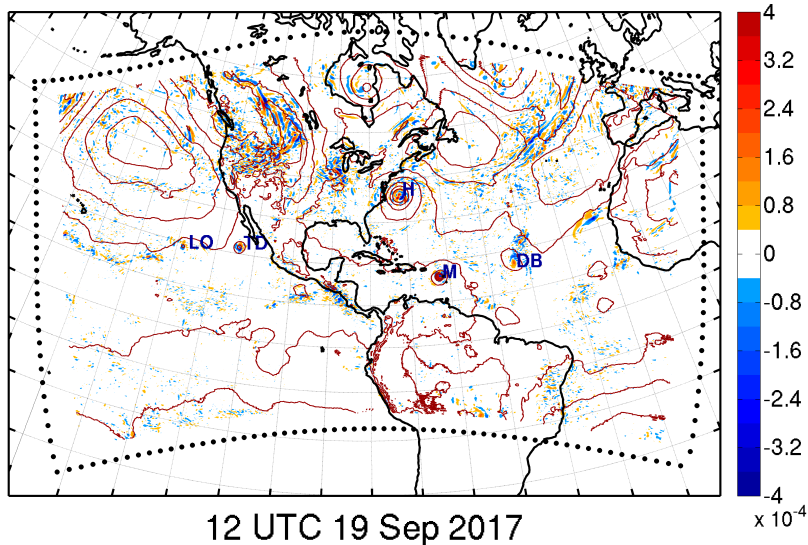
Posterior member 5

MSLP and vorticity increments at level 20 (~ 875 mb)



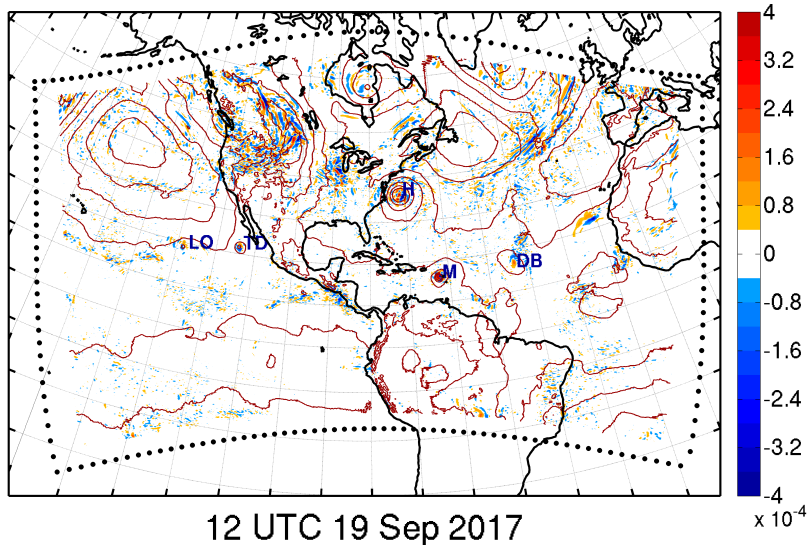
Posterior member 6

MSLP and vorticity increments at level 20 (~ 875 mb)



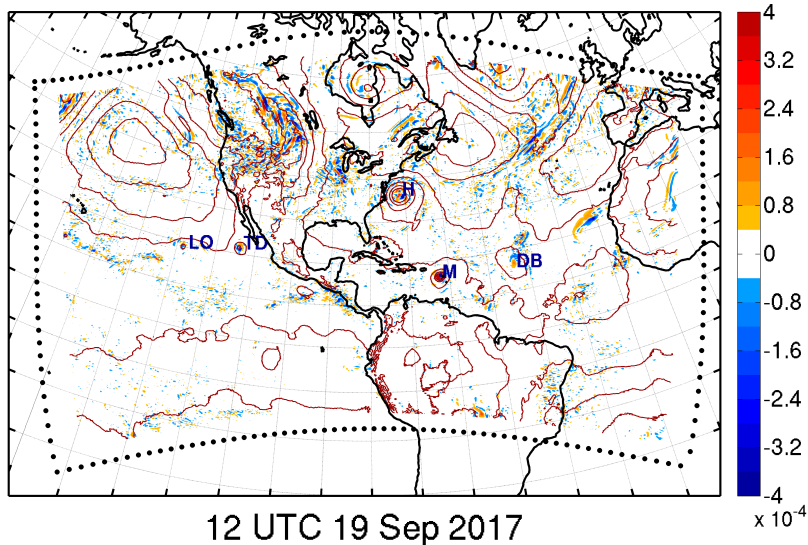
Posterior member 7

MSLP and vorticity increments at level 20 (~ 875 mb)



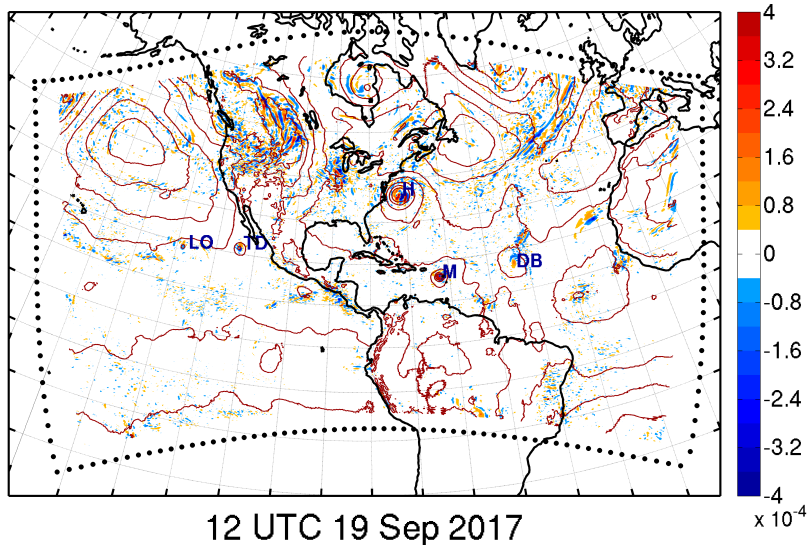
Posterior member 8

MSLP and vorticity increments at level 20 (~ 875 mb)



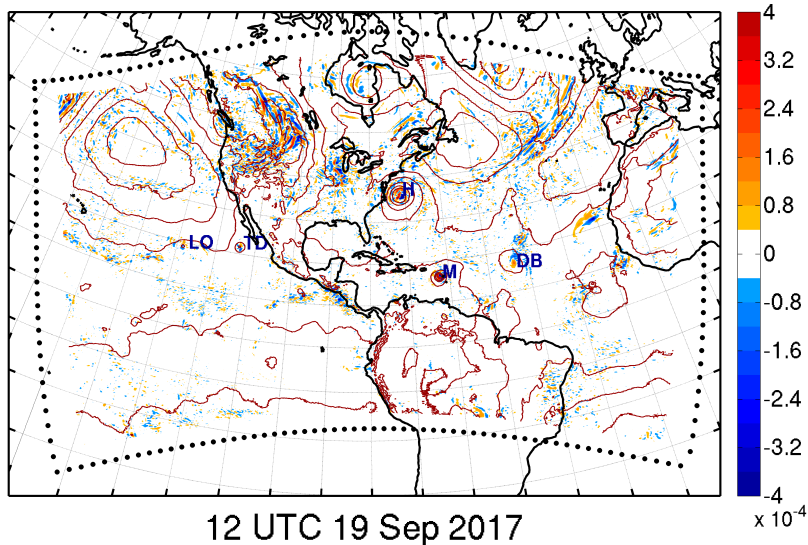
Posterior member 9

MSLP and vorticity increments at level 20 (~ 875 mb)

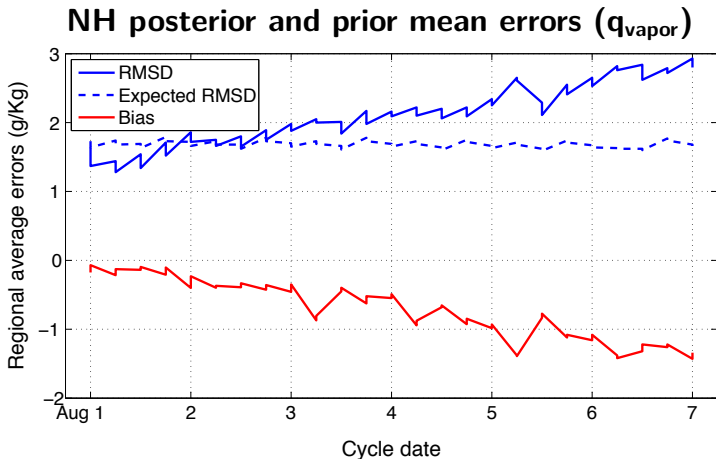


Posterior member 10

MSLP and vorticity increments at level 20 (~ 875 mb)



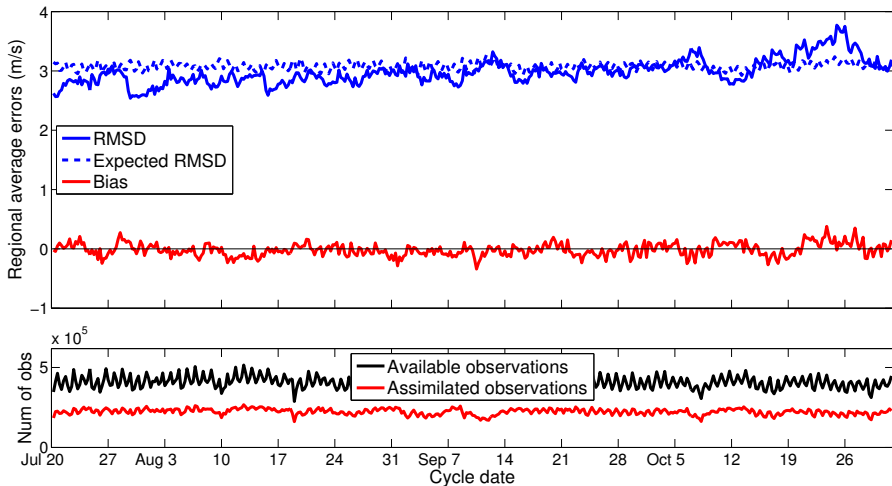
Identifying model and DA problems



Obs-space verification uncovered error caused by incorrect specification of lat/lon in single-domain HWRF configuration.

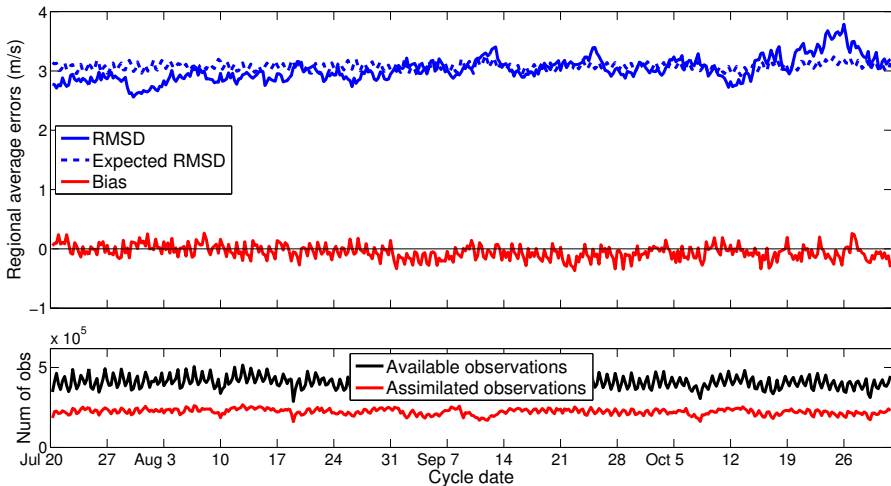
Obs space verification

Domain average prior errors (U-wind)



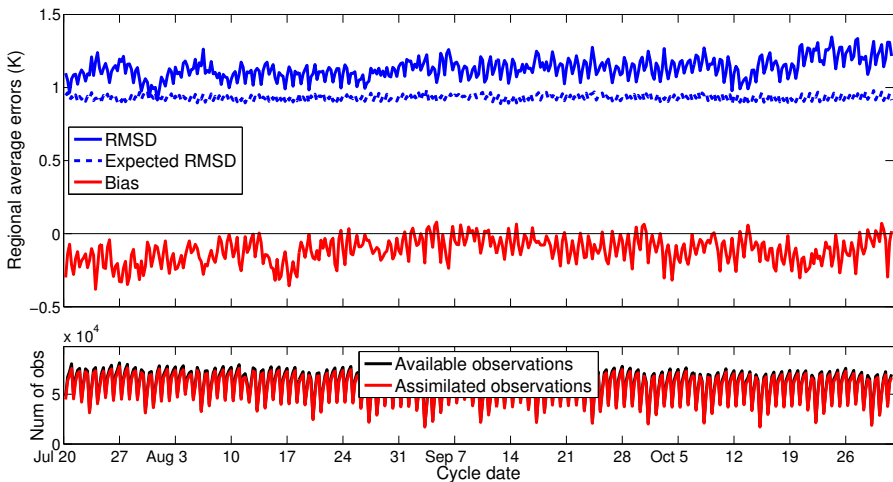
Obs space verification

Domain average prior errors (V-wind)



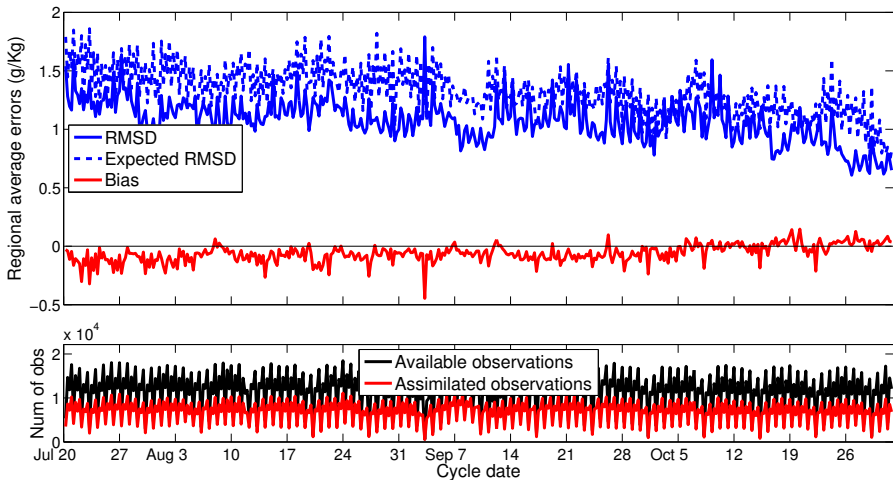
Obs space verification

Domain average prior errors (T)

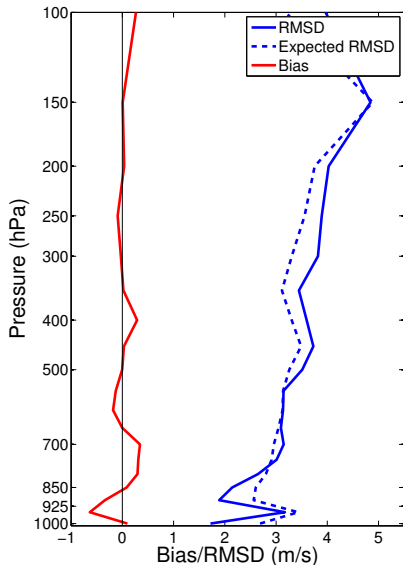


Obs space verification

Domain average prior errors (q_{vapor})



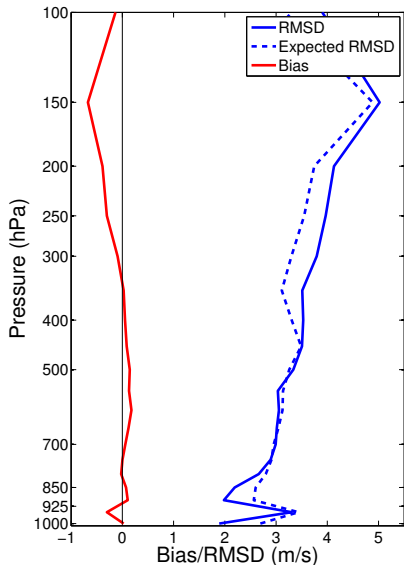
Obs space verification



U-wind

- Vertical profiles of time-space averaged prior errors.
- Mean RMSD (solid blue), mean “expected RMSD” (dashed blue), and mean bias (red).

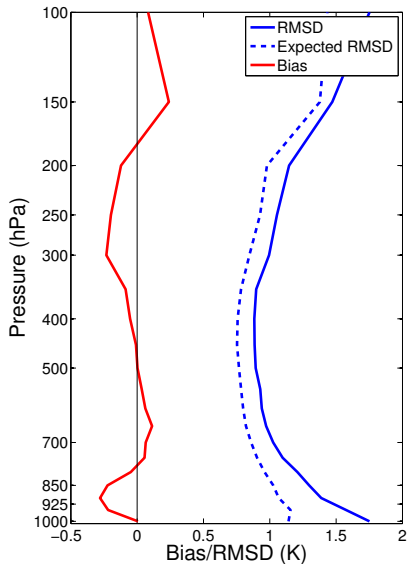
Obs space verification



V-wind

- Vertical profiles of time-space averaged prior errors.
- Mean RMSD (solid blue), mean “expected RMSD” (dashed blue), and mean bias (red).

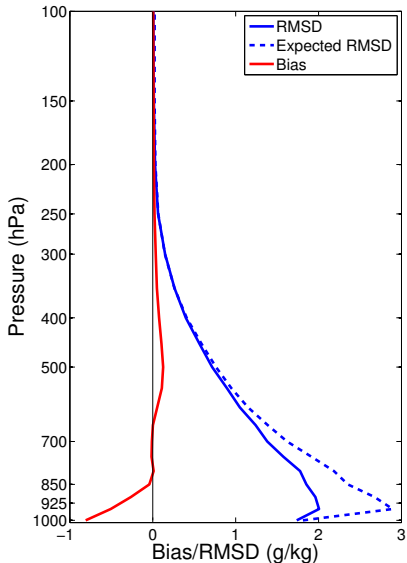
Obs space verification



T

- Vertical profiles of time-space averaged prior errors.
- Mean RMSD (solid blue), mean “expected RMSD” (dashed blue), and mean bias (red).

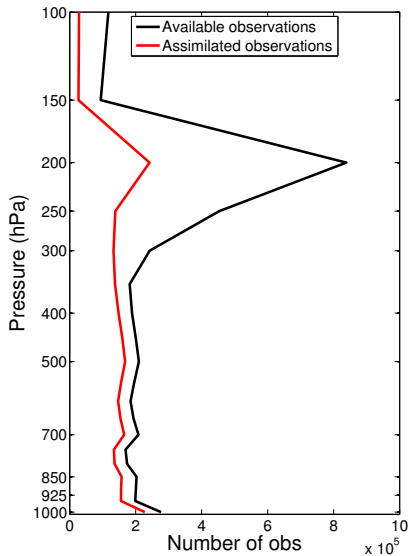
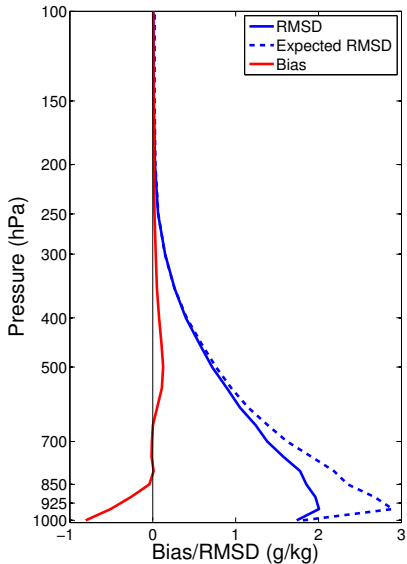
Obs space verification



q_{vapor}

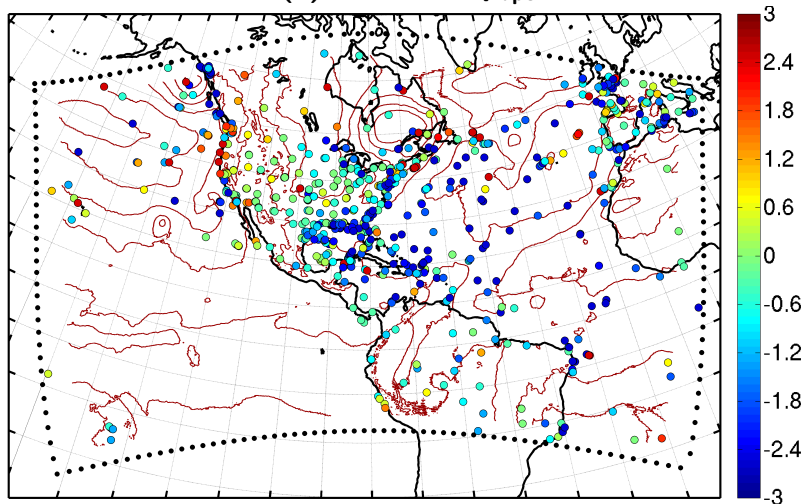
- Vertical profiles of time-space averaged prior errors.
- Mean RMSD (solid blue), mean “expected RMSD” (dashed blue), and mean bias (red).

Obs space verification



Regional water vapor errors

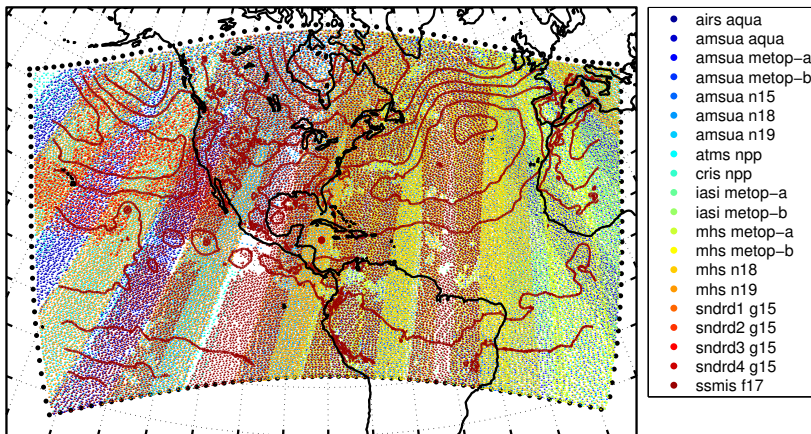
$H(x)$ - surface q_{vapor}



12 UTC 23 Aug 2017

Satellite observation coverage

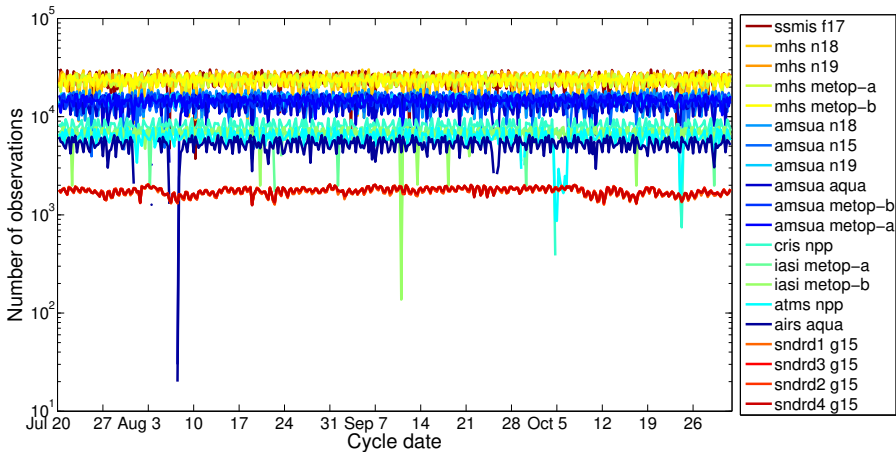
12 UTC radiance observation locations



12 UTC 20 Jul 2017

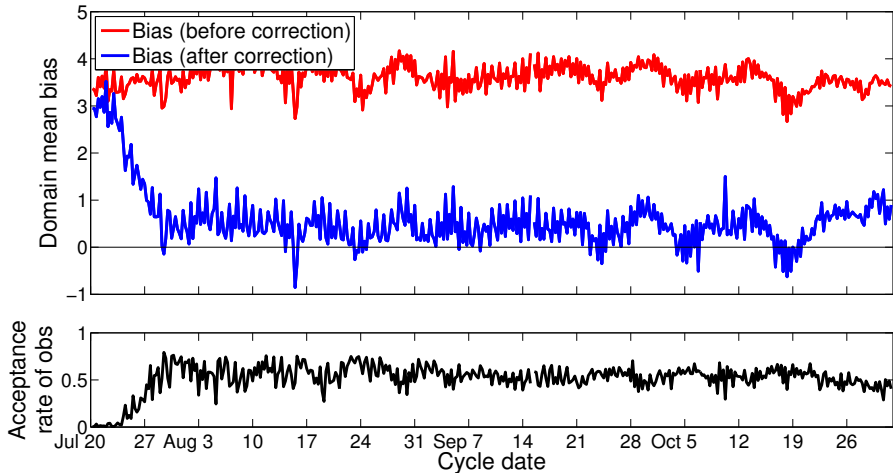
Satellite observation coverage

Count of radiance locations for each instrument



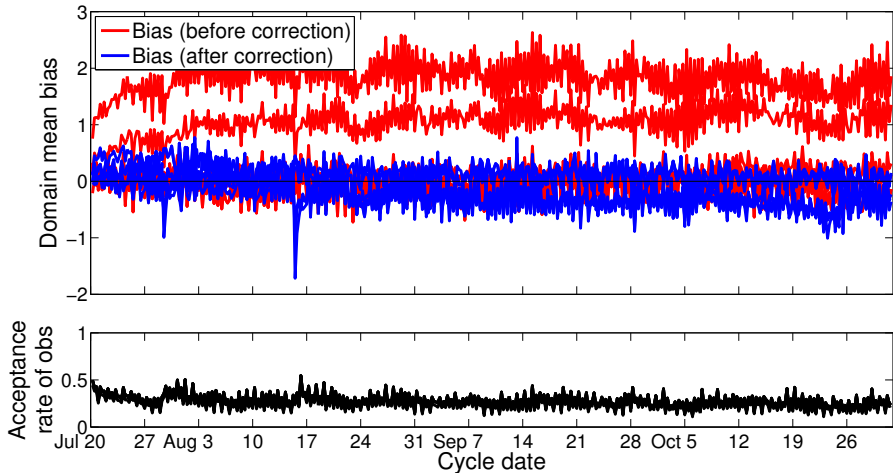
Radiance bias estimation

NOAA 18 AMSUA T_b bias (channel 10)



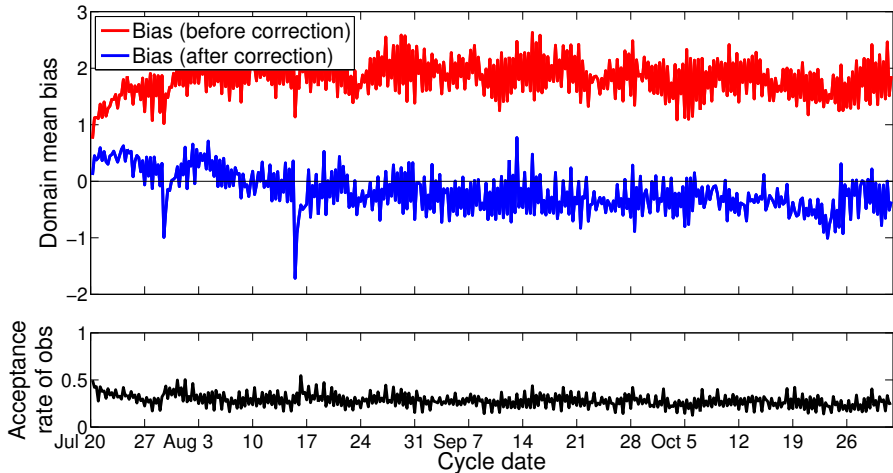
Radiance bias estimation

NOAA 18 MHS T_b bias (channels 55–59)



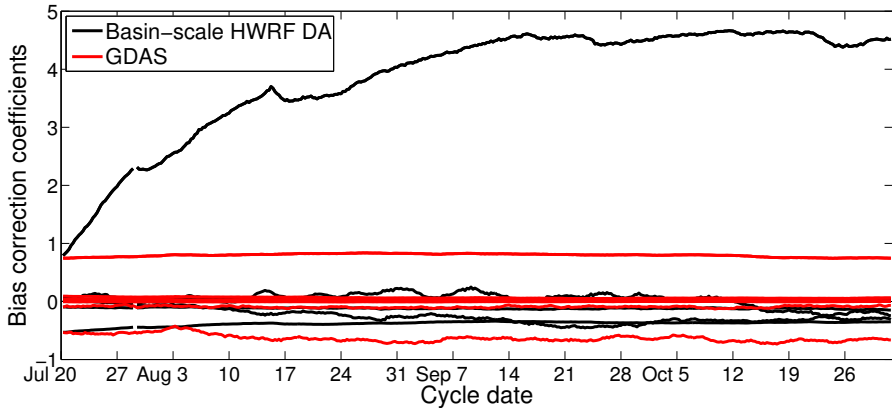
Radiance bias estimation

NOAA 18 MHS T_b bias (channel 57)



Radiance bias estimation

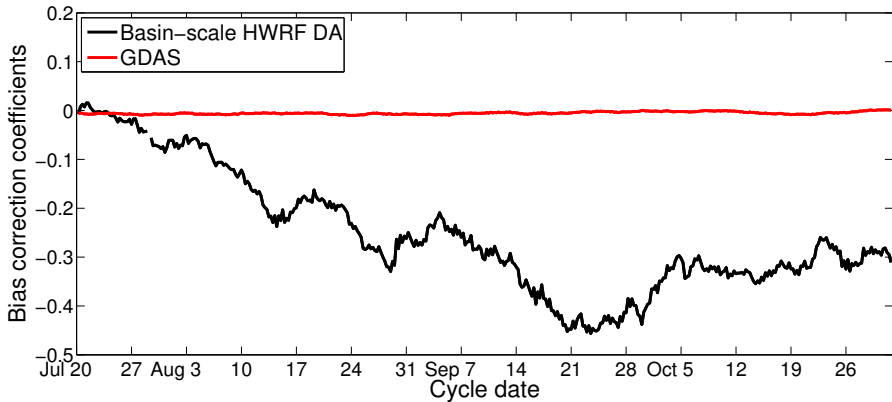
Bias correction coefficients



$$\tilde{h}(\mathbf{x}, \beta) = h(\mathbf{x}) + \sum_{i=1}^K \beta_i p_i(\mathbf{x})$$

Radiance bias estimation

Bias correction coefficients (cloud liquid water)



$$\tilde{h}(\mathbf{x}, \boldsymbol{\beta}) = h(\mathbf{x}) + \sum_{i=1}^K \beta_i p_i(\mathbf{x})$$

Ongoing and future work

Perform a more thorough analysis of observation-space errors.

- Apply regional and temporal stratification of prior errors to examine sources of bias.

Explore online bias correction in more detail and provide additional comparisons with GFS values.

- Use predictor weights to explore sources of error in the model or DA methodology.
- Currently performing additional data assimilation experiments using GFS bias correction values (no online estimate).

Ongoing and future work

Generate ensemble forecasts from EnKF analyses (10 – 20 members subsampled).

- Examine benefits and deficiencies in probabilistic skill over global models.
- Identify areas where improvement is needed.

Run localized particle filter for basin-scale HWRF

- Code efficiency still needs improving.
- Initial experiments will target select periods from 3-month seasonal analysis.