9A.6   **The Developmental Testbed Center Objective Evaluation Performed During the 2010 NOAA Hazardous Weather Testbed Spring Experiment**

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1. Introduction

The collaboration between HWT and Developmental Testbed Center (DTC) for the HWT Experimental Forecast Program (EFP) Spring Experiment started in 2008 and has strengthened and grown over the past three years. In addition to using deterministic and ensemble-based convection-allowing model forecasts as guidance for experimental probabilistic severe convective weather forecasts, the 2010 Spring Experiment included additional convective hazards for QPF/extreme precipitation events and aviation-related thunderstorm impacts.

The DTC objective evaluation during the HWT 2010 Spring Experiment (SE2010) complemented the subjective evaluation that has traditionally taken place. With the addition of probabilistic verification capabilities in the DTC's Model Evaluation Tool (MET), both probabilistic products and deterministic forecasts were evaluated this year. DTC evaluated output from the CAPS Storm Scale Ensemble Forecast (SSEF), the NOAA/ESRL/GSD High Resolution Rapid Refresh (HRRR), and the North American Mesoscale (NAM) model and the Short Range Ensemble Forecast System (SREF), both produced by NOAA/NCEP/EMC. The evaluation focus was on products derived from the simulated reflectivity and quantitative precipitation forecast (QPF) fields.

It is anticipated that both the subjective and objective evaluations performed in near-real time at the SE2010 will eventually lead to greater use of latest convection-allowing model forecasts by the NOAA/NWS Storm Prediction Center, NOAA/NWS Hydrometeorological Predication Center (HPC) and NOAA/NWS Aviation Weather Center (AWC). This talk will describe the DTC objective evaluation performed during the 2010 Spring Experiment, highlight key results, and describe anticipated future work.

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2. Methodology

The NOAA HWT Spring Experiment is a yearly experiment that, in recent years, has investigated the use of convection-allowing model forecasts as guidance for the prediction of severe convective weather. A variety of model output has been examined and evaluated daily during the experiment and experimental severe weather forecasts have been created and verified. The variety of models available to the Spring Experiment has allowed the HWT to explore different types of guidance, including products derived from both ensembles and deterministic forecasts. This year, the models available to the Spring Experiment include: NOAA National Severe Storms Laboratory (NSSL) 4km WRF-ARW, NOAA National Centers for Environmental Prediction (NCEP) Environmental Modeling Center (EMC) 4km WRF-NMM, National Center for Atmospheric Research (NCAR) 3km WRF-ARW, University of Oklahoma Center for Analysis and Prediction of Storms (CAPS) 4km Storm Scale Ensemble Forecast (SSEF) 26 member multi-model ensemble, and the NOAA Earth System Research Laboratory (ESRL) High Resolution Rapid Refresh (HRRR) 3km WRF-ARW.

The overarching goals of the 2010 HWT-DTC Spring Experiment evaluation are to 1) provide objective evaluations of the experimental forecasts; 2) supplement and compare to subjective assessments of performance; and 3) familiarize forecasters and researchers with both new and traditional approaches for evaluating forecasts. The objectives for the 2010 Spring Experiment include: 1) augmenting the samples available to evaluate the impact of radar assimilation on short-term forecasts; 2) providing evaluation of probabilistic quantitative precipitation forecasts (QPF) of extreme precipitation events; and 3) providing evaluation of predicted radar echo top heights, defined by the height of the 18 dBZ reflectivity.

This year DTC is evaluating (in near real-time) all 26 members of the CAPS SSEF ensemble as deterministic models, some of the CAPS SSEF ensemble products, and the HRRR deterministic model. DTC is also bringing in two operational models, the EMC North American Mesoscale (NAM) 12 km WRF-NMM

deterministic model, and the EMC Short Range Ensemble Forecast (SREF) 32-35 km 21 member multi-model ensemble. Table 1 indicates the forecast variables, observation fields, and evaluation metrics applied to these models. A full description of the each model contributed to HWT may be found on their website at:

http://hwt.nssl.noaa.gov/Spring_2010/

Objective evaluation was performed using the DTC's MET tool. Traditional statistics for categorical (thresholded) forecasts were computed using the Grid-Stat tool in MET. Specific statistics provided include:

Gilbert Skill Score (GSS), Critical Success Index (CSI), Probability of Detection Yes (PODY), False Alarm Ratio (FAR), and Frequency Bias (FBIAS). Evaluation using a spatial method was performed using the Method for Object-Based Diagnostic Evaluation (MODE) tool in MET. Metrics provided include: Median of Maximum Intensity (MMI), Intersection Area, Area Ratio, Centroid Distance, Angle Difference, % Objects and Area Matched, and 50th and 90th percentile within the object. Davis *et al.* (2006) provides an overview of the application of MODE, including many of these attributes to the QPF verification problem. A complete description of the MET tools may be found at: http://www.dtcenter.org/met/users/docs/overview.php.

Table 1. List of variables (and thresholds) to be evaluated during SE 2010. ROC (Receiver Operator Characteristics Curve), GSS (Gilbert Skill Score), CSI (Critical Success Index), FAR (False Alarm Ratio), PODY (Probability Of Detection Yes), FBIAS (Frequency Bias), MMI (Median of Maximum Interest)

FCST Field	Observation	Grid-Stat	MODE	Models
Prob of Exceed (0.5", 1", 2" over 3 and 6 hrs)	0.5", 1", 2" QPE over 3 and 6 hrs	Brier Score, Decomp of Brier score, Area under ROC	None	Ensemble products from CAPS and SREF
50% Prob of Exceed (0.5", 1", 2" over 3 and 6 hrs)	0.5", 1", 2" QPE over 3 and 6 hrs	None	MMI, Intersection Area, Area Ratio, Centroid Distance, Angle Difference, % Objects and Area Matched, 50th and 90th percentile	Ensemble products from CAPS and SREF
QPF (0.25", 0.5", 1.0", 2" over 3 and 6 hrs)	0.25", 0.5", 1.0", 2" QPE over 3 and 6 hrs	GSS, CSI, FAR, PODY, FBIAS	Same as above	CAPS members, CAPS ens products, SREF ens mean, HRRR, NAM
Sim. Comp. Refl (20,30,40,50 dBZ)	Q2 Composite refl (20,30,40,50 dBZ)	GSS, CSI, FAR, PODY, FBIAS	Same as above	CAPS members, CAPS ensemble products, HRRR, NAM
18 dBZ Echo Top (18, 25, 30, 35, 40, 45 kft)	Q2 18dBZ Echo Top (18, 25, 30, 35, 40, 45 kft)	GSS, CSI, FAR, PODY, FBIAS	Same as above	CAPS members, CAPS ens mean, HRRR
Prob of 40dBZ echos	Q2 Composite reflectivity (40dBZ)	GSS, CSI, FAR, PODY, FBIAS	None	Ensemble products from CAPS and SREF
50% Prob of 40dBZ echos	Q2 Composite reflectivity (40dBZ)	None	Same as above	Ensemble products from CAPS

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Included in the evaluation were the 00 UTC initialization of each model. All models were re-gridded to the 4 km Stage IV grid configuration. The 00 UTC models were evaluated over three regions: the entire domain, the static VORTEX-2 domain provided by CAPS at the 12 UTC initialization time, and a regional, movable area-of-interest domain selected by HWT Spring Experiment participants each day. Figure 1 depicts examples of these domains.

Ensemble Product Description

For this paper, the Ensemble Simple Mean represents the arithmetic mean at each gridpoint. Ensemble Simple Frequency is the number of ensemble members meeting or exceeding the threshold at a given grid-point. Probability Neighborhood method (PN) applies a neighborhood method using a Gaussian filter with radius of interpolation of 40 km and prior to computing the ensemble relative frequency. The Probability Matching method (PM) orders the PDF of the maximum intensity of all members at each grid point (Ebert, 2001). It assigns the maximum intensity value to the location of the maximum in the simple ensemble mean, the second highest intensity value to the location of the second maximum value in the simple mean and so on. This allows the shape of the ensemble mean to be maintained but the intensities to be more dynamic.
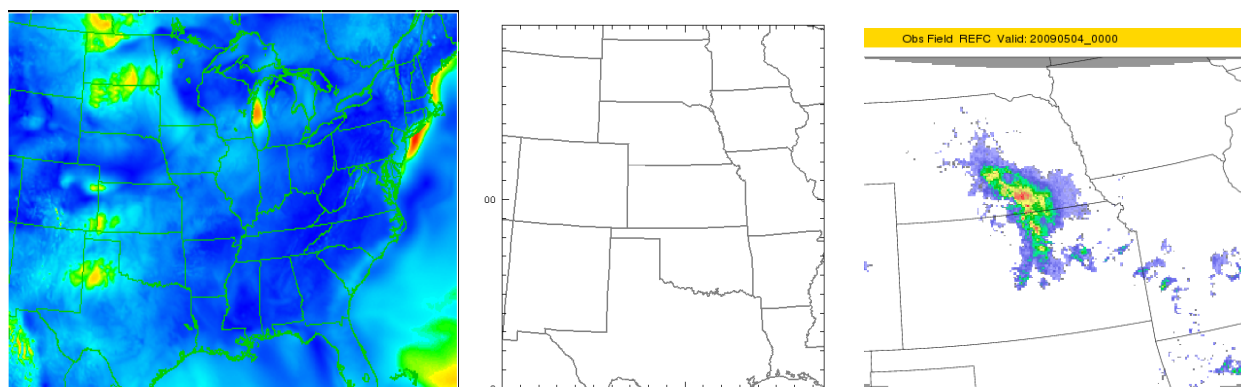
Figure 1. Examples of the three evaluation domains used in the DTC objective evaluation of HWT SE2010 models. Left: Full domain represents two-thirds CONUS; Upper Right: VORTEX-2 domain; and Lower Right: regional area-of-interest domain that moves daily; referred to as daily domain.

3. 2010 Spring Experiment Results

The goal of the objective evaluation is to help the forecasters and researchers understand the skill of these storm scale models. A comparison of subjective evaluation with the objective results was performed daily during the Experiment. The following section discusses the objective evaluation. The results represent a homogeneous sample of 20-25 (depending on the variable) which is too small of a sample size to infer statistical significance.

3a. Simulated Radar Reflectivity

Simulated radar reflectivity was used by Spring Experiment participants taking part in the Severe Weather and Aviation Weather forecast exercises. Kain et. al, 2010 showed that SE2008 and SE2009 participants subjectively found that radar data assimilation methods provide somewhat more realistic forecasts of simulated radar reflectivity fields during 00-06hr lead times but small scale features of the simulated reflectivity appeared to lose coherence more rapidly. They also showed there was better spatial over-lap of the simulated field when radar assimilation methods were employed.

The SE2010 objective evaluation supports the subjective impression and agrees with the objective findings in Kain et. al. Figure 2 shows Gilbert Skill Score (GSS - also known as Equitable Threat Score) simulated reflectivity greater than 20, 30 and 40 dBZ aggregated over all available runs during the Spring Experiment. At all thresholds and during lead times 0 - 3 hours, there is a marked increase in GSS for the CAPS ensemble control member, the ensemble mean, and the 1-km deterministic s with radar assimilation (red, black, and purple lines respectively) as compared with control members with no radar assimilation (red dashed line) implying the benefits gained from radar assimilation are washed out fairly quickly. A nominal increase in skill is maintained out to 12 hours. We note there that positive impacts of radar data in hourly precipitation GSS were reported by Xue et al. (2008) and Kong et al (2009) to last between 6 to 12 hours for the 2008 and 2009 spring seasons of CAPS forecasts, depending on the threshold.

The CAPS Probability Matched mean (black), and CAPS 1 km simulation (purple) appear to outperform HRRR (blue) and NAM (green). It should be noted upper air data is not available for data assimilation into the HRRR 00 UTC and hence skill may be

reduced due to this deficiency. The HRRR 01 UTC run has the upper air data assimilation and may make a better benchmark for future comparison. GSS increases at later lead times which may be explained by an increase in convection during the afternoon and evening hours. This increase is reflected in Base Rate (depicted by gray bars). The period of lower GSS scores between 14 and 18 hours correspond to the morning minimum of convective activities. However, all models except the NAM, which is not convection allowing, seem to struggle with dissipation of convection through the early morning hours and then convective initiation at lead times of 19-23 hours. Harrold et al (2011) discusses an object based evaluation of this trend using MET.

The skill demonstrated by GSS is sensitive to hits and therefore should be interpreted in light of FBIAS (i.e. the ratio of forecast area to observed area, can be used to provide a sense of over or under-prediction of a particular variable). The optimal value of frequency

bias is 1. Scores larger than this indicate an over-forecast of the given threshold. Less than 1 indicates an under-forecast. The median daily FBIAS values aggregated over the experiment is not shown here but during 0-6 hr lead times, some of the skill demonstrated by convective allowing models with respect to the NAM baseline may be due to the higher FBIAS.

Prediction of reflectivity exceeding 40 dBZ is notoriously difficult for mesoscale models, such as the NAM. Figure 3 shows the box and whisker plot of the frequency bias for thresholds exceeding 40 dBZ. As expected the NAM has no frequency bias, and hence no prediction of reflectivity greater than 45 dBZ. The convection allowing models seem to do reasonably well up to a reflectivity of 50 dBZ and under forecast 60 dBZ reflectivites. Additionally, it appears this is one area where the 1 km deterministic model excels.
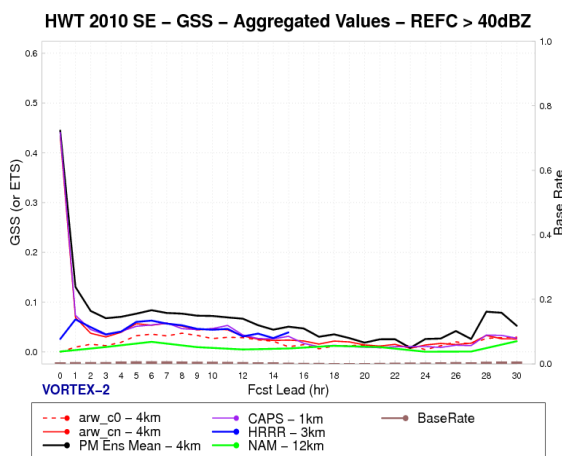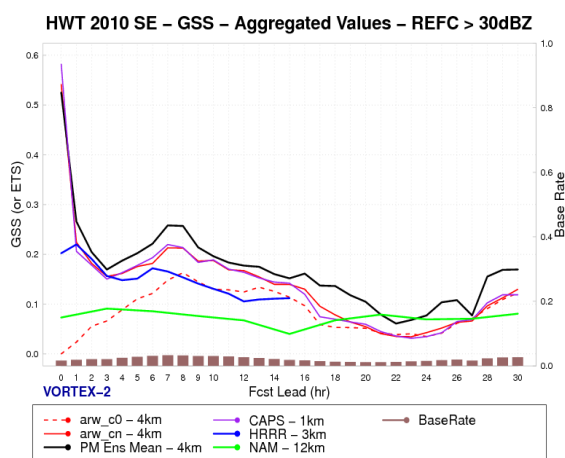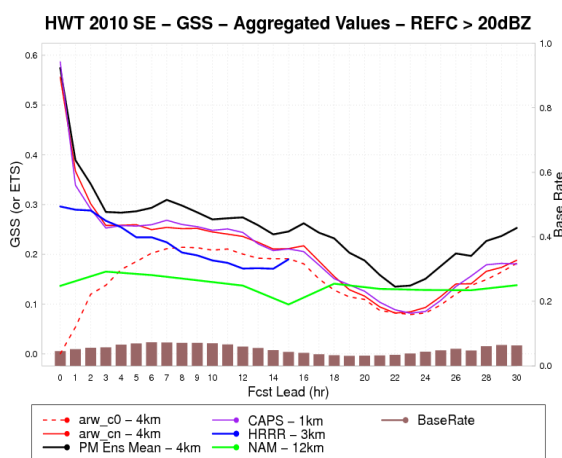
Figure 2. GSS for simulated radar reflectivity thresholded at 20 dBZ (upper left), 30 dBZ (upper right), and 40 dBZ (lower left) aggregated over five weeks of the HWT Spring Experiment. Red lines: control members of the CAPS SSEF representing models with (ssef_s4cn) and without (ssef_s4c0) radar assimilation. Black: CAPS SSEF simple ensemble mean. Purple: CAPS SSEF 1 km WRF-ARW simulation. Green: NCEP/EMC NAM 12 km WRF-NMM simulation. Blue: NOAA/ESRL HRRR 3 km WRF-ARW simulation. Brown bars: Base Rate (or observed event fraction) and scaled on right axis.

HWT 2010 SE – Aggregated Values – FBIAS
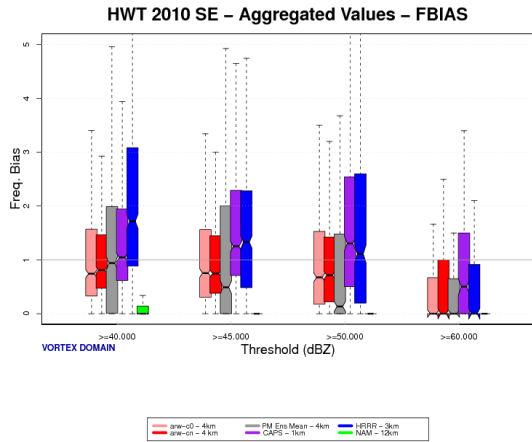
Figure 3. Box and whisker plot of Frequency Bias for reflectivity stratified by threshold. Notches represent the confidence interval around the median (dash in middle). Models represented include: CAPS arw-c0 (pink), arw-cn (red), ensemble mean (gray), 1 km (purple), HRRR (blue) and NAM (green)

3b. QPF and Probabilistic QPF

The QPF and probabilistic QPF fields were used primarily by the participants in the Hydrometeorological forecasting exercise.Subjectively, it appeared that the simple ensemble relative frequency probability product derived from the CAPS SSEF ensemble is a step-forward from using SREF and NAM for the QPF problem, but two other ensemble products appear to be equally promising. One method is the neighborhood method (PN QPF). The other is a probability matching method (PM QPF).

Figure 4 is a screenshot of one of the displays available on the DTC/HWT Objective Evaluation Website (http://verif.rap.ucar.edu/eval/hwt/2010/) and provides a sample of how the subjective impression was formed. The top row shows the observed field, CAPS simple ensemble relative frequency probability field, SREF simple ensemble relative frequency probability field, CAPS PN field, NAM deterministic QPF field, and CAPS PM QPF field. The bottom row shows the respective objects that were identified by the MET spatial verification package called MODE.

Traditional statistics and spatial attributes for these fields were calculated using MET. Based on the Brier Score for the 12 hr forecast (Figure 5) valid CAPS simple PQPF appears to have more skill than the other two products. However, the CAPS PN PQPF shows more resolution and has a higher Area under the Receiver Operating Characteristic Curve and could potentially be deemed more skillful for this model run.
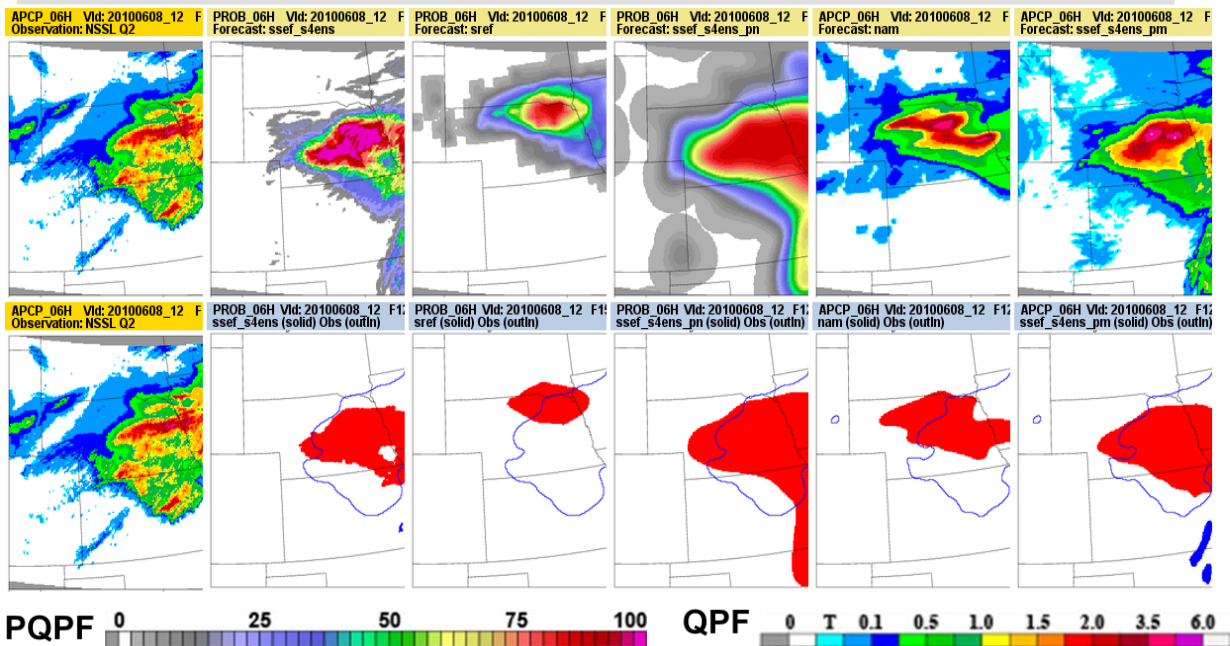

Figure 4. Screenshot of a Probability QPF spatial verification display available on DTC/HWT objective evaluation website. Plots are the 12hr forecast valid at 8 June 2010. The top row: observed field; CAPS Simple PQPF field; SREF Simple PQPF field; CAPS PN PQPF field, NAM QPF field, and CAPS PM PQPF field. The bottom row shows forecast (solid) and observed (blue line) objects identified by MODE for probability >50% and accumulated precipitation > 0.5 inch. PQPF units – percent. QPF units – inches.

HWT 2010 Median Scores for Probability of Precipitation >= 0.5 inches over 6 hrs
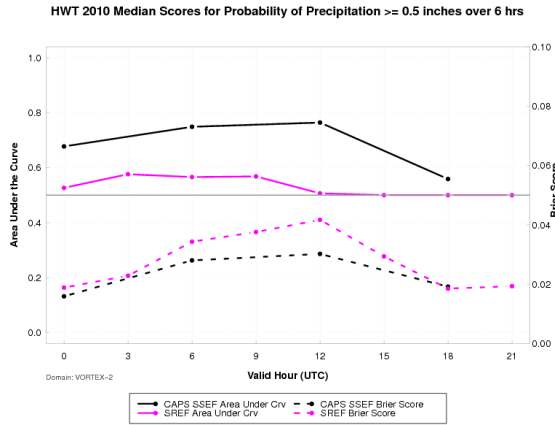
Figure 5. Area under the Receiver Operator Characteristic curve (solid) and Brier Score (dash) for CAPS ensemble mean (black) and SREF (pink).

One MODE attribute that provides an indication of a models ability to predict extreme events is the difference between Forecast and Observed near-peak (90th percentile) rainfall with matched objects. Figure 6 provides a box and whisker plot to compare the single value (aka deterministic) QPF. With a difference value near zero, the HRRR appears to capture the rainfall intensity the most reliably for the lead times it is available. Interestingly, the NAM appears to have a slightly less under-prediction versus the ensemble-mean.

**HWT 2010 SE – MEDIAN of DAILY MODE VALUES – APCP_06 > 0.5"**
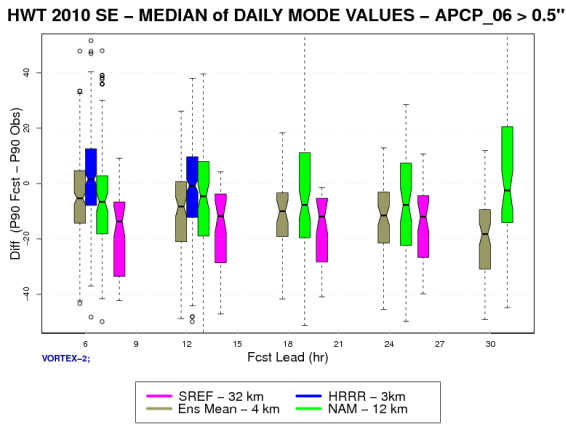
Figure 6. Box and whisker plot of near peak rainfall within MODE objects stratified by lead-time. Notches represent the confidence interval around the median (dash in middle). Models represented include: CAPS ensemble mean (gray), 1 km (purple), HRRR (blue), NAM (green), and SREF (pink).

### 3c. Radar Echo Top Product

The radar echo top product was used primarily by participants in the Aviation Weather forecast exercise. The Spring Experiment participants examined individual members of the CAPS SSEF system. One comparison was how different the simulated radar reflectivity field was for different cloud microphysics packages. One subjective impression that occurred on several days was that the Thompson scheme in WRF-ARW model tended to over develop the stratiform region. This over-prediction impacted the simulated reflectivity and hence 18 dBZ radar echo top height, and potentially accumulated precipitation products.

Figure 7 is a screenshot of another display available on the DTC/HWT Objective Evaluation Website (http://verif.rap.ucar.edu/eval/hwt/2010/) and provides a sample of how the subjective impression was formed. The top row shows the observed field, CAPS PM mean, CAPS cn_arw member with Thompson microphysics scheme, CAPS m15_arw member with WRF double moment microphysics scheme, CAPS m16_arw member with WRF single moment microphysics scheme, and CAPS m17_arw member with Morrison microphysics scheme. The bottom row shows the respective objects that were identified by the MET spatial verification package called MODE. There is a significantly larger stratiform cloud shield in the WRF-ARW Thompson simulation then the other microphysics perturbations. This large shield appears to dominate the simple ensemble mean, likely reflecting the presence of 9 SSEF members that are configured with the Thompson scheme.

The CAPS PM mean appears to have a sizable over-prediction of radar echo top > 25,000 ft. This may be due to the over-prediction of stratiform region of the convection in several members, including the Thompson scheme members. Figure 8 provides the symmetric difference (non-intersecting areas) and centriod difference of MODE objects calculated from the daily values during the last 4 weeks of the experiment. In Figure 8 the HRRR (blue) has the lowest symmetric difference suggesting the forecast objects matched the observed objects more closely than the 1 km (purple). The over-prediction of the cloud shield is clearly evident in the large symmetric differences for the ensemble mean. Additionally, the large centroid differences evident for the ensemble mean during the 7-12 hour lead times suggests there may be a displacement error that may be causing the large symmetric differences.
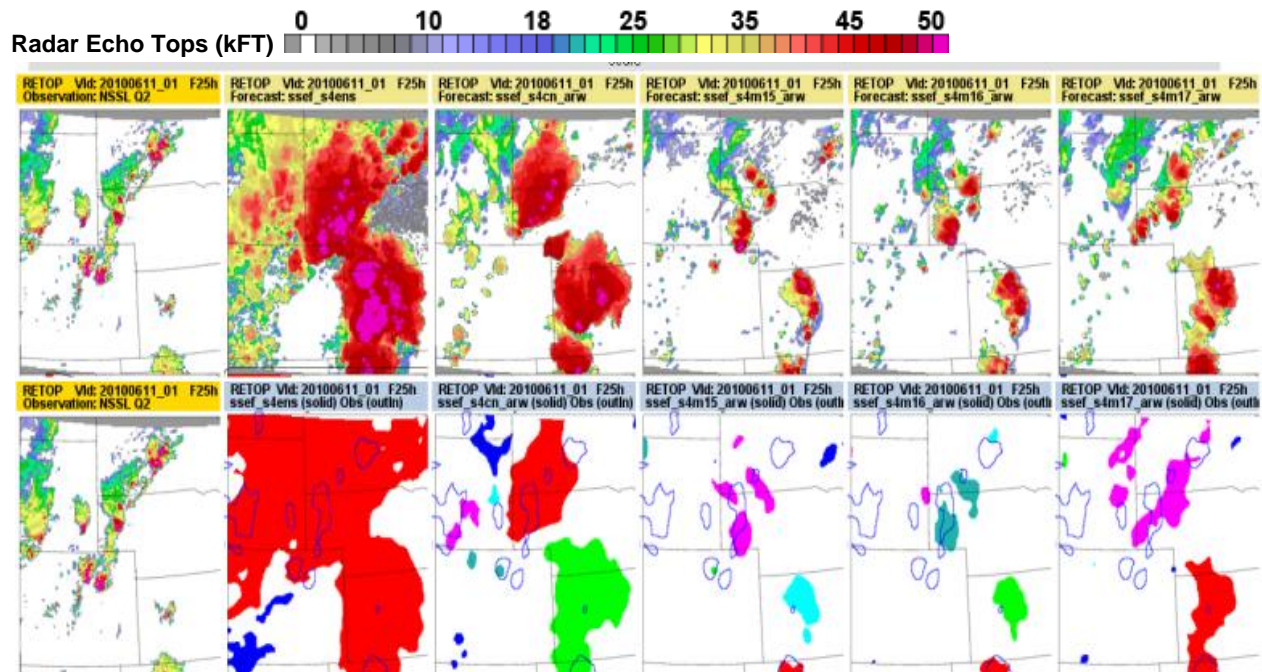
Radar Echo Tops (kFT)

Figure 7. Screenshot of a 18dBZ radar echo top height and spatial verification display. Plots are the 12hr forecast valid at 8 June 2010 12UTC. The top row: Q2 observed field; CAPS Simple Probability field; SREF Simple Probability field; CAPS Probability Neighborhood field, NAM deterministic QPF field, and CAPS Probability Matched QPF field. The bottom row shows forecast (solid) and observed (blue line) objects identified by MODE for 18dBZ echo top height > 25000 ft. Different colors indicate forecast cluster of objects matched with underlying observation objects.
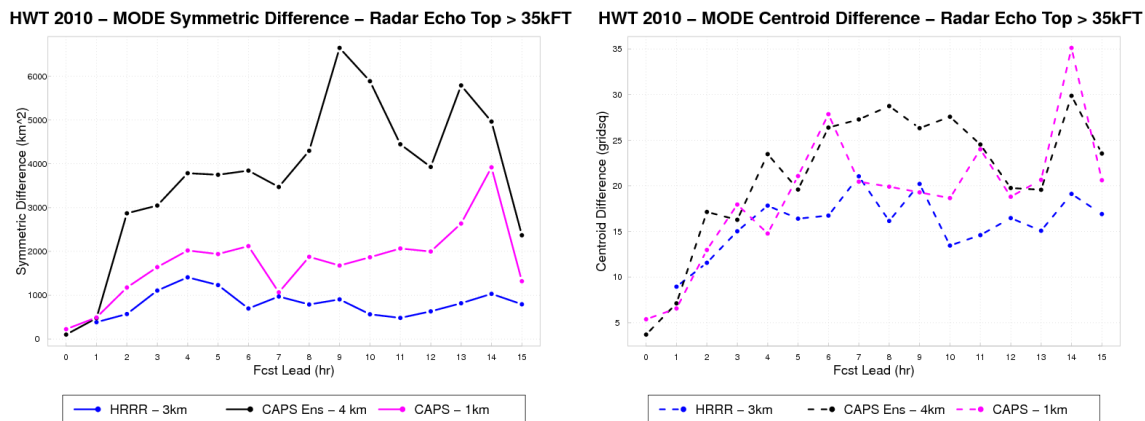
Figure 8. Median symmetric difference (left – solid lines) and centroid difference (right – dashed lines) for MODE objects identified in the Radar Echo Top field using a convolution radius of 5 and convolution threshold of >= 35,000 ft. Models evaluated include the HRRR (blue), CAPS 1km (purple) and CAPS ensemble mean (black).

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4. Summary

In recent years, the NOAA Hazardous Weather Testbed Spring Experiment has presented the research community with a unique opportunity to have their storm-scale simulations used and evaluated by the forecasting community. DTC has been collaborating with HWT by providing objective evaluation during the Spring Experiment for the past three years. Each year this collaboration has grown.

This year, 29 deterministic models and four probabilistic products from two ensemble systems were evaluated in near real-time. Evaluation of individual runs, as well as aggregations of the experiment, were available on-line for use each day in the Spring Experiment at http://verif.rap.ucar.edu/eval/hwt/2010/. DTC also had a staff member in attendance throughout the Spring Experiment to lead a discussion on objective verification.

This paper shows examples of results from the experiment along with some sample interpretation. Initial results support the findings on impact of radar assimilation on the 0-12 hr forecast presented in Kain et. al (2010). Additionally, evaluation of CAPS ensemble products indicate the neighborhood probability QPF (PQPF) and probability matched QPF products show potential for utility in a forecast setting. Finally, this paper provides one example of how the DTC/HWT objective evaluation may be able to provide feedback to model developers for model physics improvement. Convection allowing models appear to be more skillful in simulating higher radar reflectivity and pinpointing maximum rainfall within a convective element. The probabilistic forecasts of the convection allowing ensemble also seems to be more skillful with less conditional bias (reliability) and discrimination of event and non-event (reliability). The probability matching method does not appear to be effective for the radar echo top forecasts as it tends to have a significant high bias. A more thorough analysis of the results will be conducted retrospectively and detailed in a HWT-DTC SE2010 final report.

5. References

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