Exploration of the NSSL Maximum Expected Size of Hail (MESH) Product for Verifying Experimental Hail Forecasts in the **2014 Spring Forecasting Experiment**



Introduction and Motivation

- Objective forecast verification of experimental severe weather forecasts was conducted in near-real time in the 2014 Hazardous Weather Testbed (HWT) Spring Forecasting Experiment (SFE), similar to what had been accomplished the prior two years (Melick et al. 2013).
- For the first time, however, individual hazard (tornado, wind, hail) probabilistic forecasts were produced by the Severe Desk led by the Storm Prediction Center (SPC) instead of one probabilistic forecast for total severe. Preliminary local storm reports (LSR) served as the primary verification dataset when computing forecast verification metrics.
- The goal of this work is to further explore radar-derived maximum expected size of hail (MESH) from the National Severe Storms Laboratory (NSSL) as a valuable surrogate to document the occurrence of hail and to verify the experimental hail forecasts. Thus, gridded MESH fields retained from the 2014 HWT SFE time period were pursued as an alternative by using a procedure similar to that for severe storm reports. From this, a comparison of results between LSRs and MESH for objective forecast verification are presented.

2014 HWT SFE: Experimental Hail Forecast

•Individual Hazard Product: SFE participants from the Severe Desk produced severe hail (≥ 1 " in diameter) forecasts using the same probability contours in SPC Day 1 operational outlooks (5, 15, 30, 45, and 60%) with the option of adding contour lines (every 5%) for localized maxima.

> *Time Period:* 16-12Z outlook period was issued in the morning by 15Z. > Philosophy: The probabilistic forecasts were defined using radius of influence (ROI) within 25 miles [40-km] of a point. *Grid:* GEMPAK graph-to-grid routine generated grid values (40-km grid; NCEP 212) of probabilities from drawn contours. > Sample Size: 17 days between May 5th – June 6th where forecast was issued and all verification data available.

2014 HWT SFE: Verification Datasets

•LSRs: Hail reports (\geq 1" in diameter) received from National Weather Service forecast offices.

Raw Grid: Counts over 20-hr valid period (through just after 12 Z) within 25 miles [40-km] of grid point on NCEP 212 grid *Binary grid of LSRs:* One or more severe hail reports at each grid point constitutes an observed object > PP LSR Grid: Following Hitchens et al. (2013), "practically perfect" [PP] hindcast created by applying 2-D Gaussian smoother (sigma=120-km) to binary grid of LSRs.

•MESH: High resolution grid (0.01 deg) of hourly maximum MESH obtained from NSSL

Raw grid: Multi-hour (20-hr) maximum field is created to coincide with the 16-12Z forecast period

Filtered grid: Remove isolated pixels by using 2-D Gaussian smoother (sigma=0.01 deg).

> Neighborhood Max Grid: Applied 40-km radius of influence [ROI]) to raw and filtered MESH 0.01 deg grid boxes. > Interpolation: 20-hr, neighborhood maximum MESH fields (raw and filtered) interpolated to NCEP 212.

Final QC check: NLDN flash count data used; MESH kept if 1 or more flashes within 40-km ROI of grid point

 \geq Binary grid of MESH: MESH \geq 29mm (Cintineo et al. 2012) at each grid point constitutes an observed object. > **PP MESH Grid:** Similar to LSR process, PP hindcast produced from binary grids of raw and filtered MESH fields.

2014 HWT SFE: Forecast Verification Metrics

> Domain: All of the daily evaluations were restricted to a mesoscale "area of interest" for possible severe convection. This small domain was movable to locations in the eastern and central US.

1.	Probability thresholds	define binary	event to	construct 2x2	Contingenc

Cont. Table Observed YES Forecast YES Forecas

S	Observed NO
a"	"b"
Hit	False Alarm
c"	"d"
ss	Correct Negative

Probability Threshold Forecast 5%, 15%, 30%, 45% Experimental Severe Hail

2. Direct comparison of probabilities between experimental hail forecast and PP:

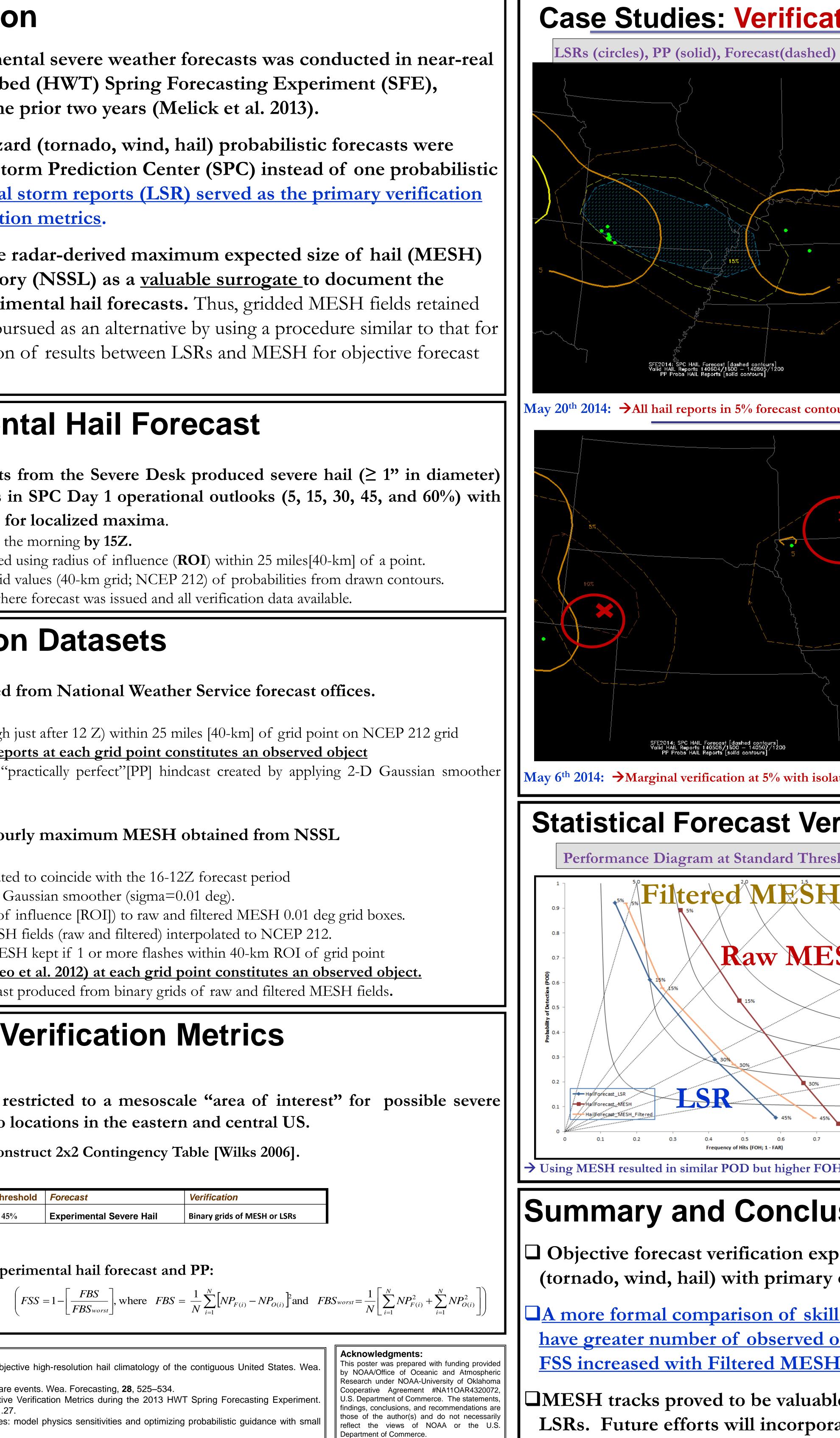
Fractions Skill Score (Schwartz et al. 2010)

References:				
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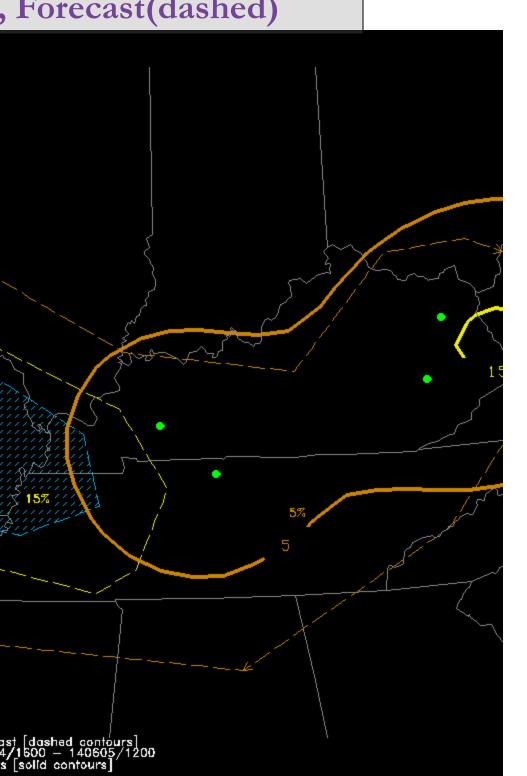


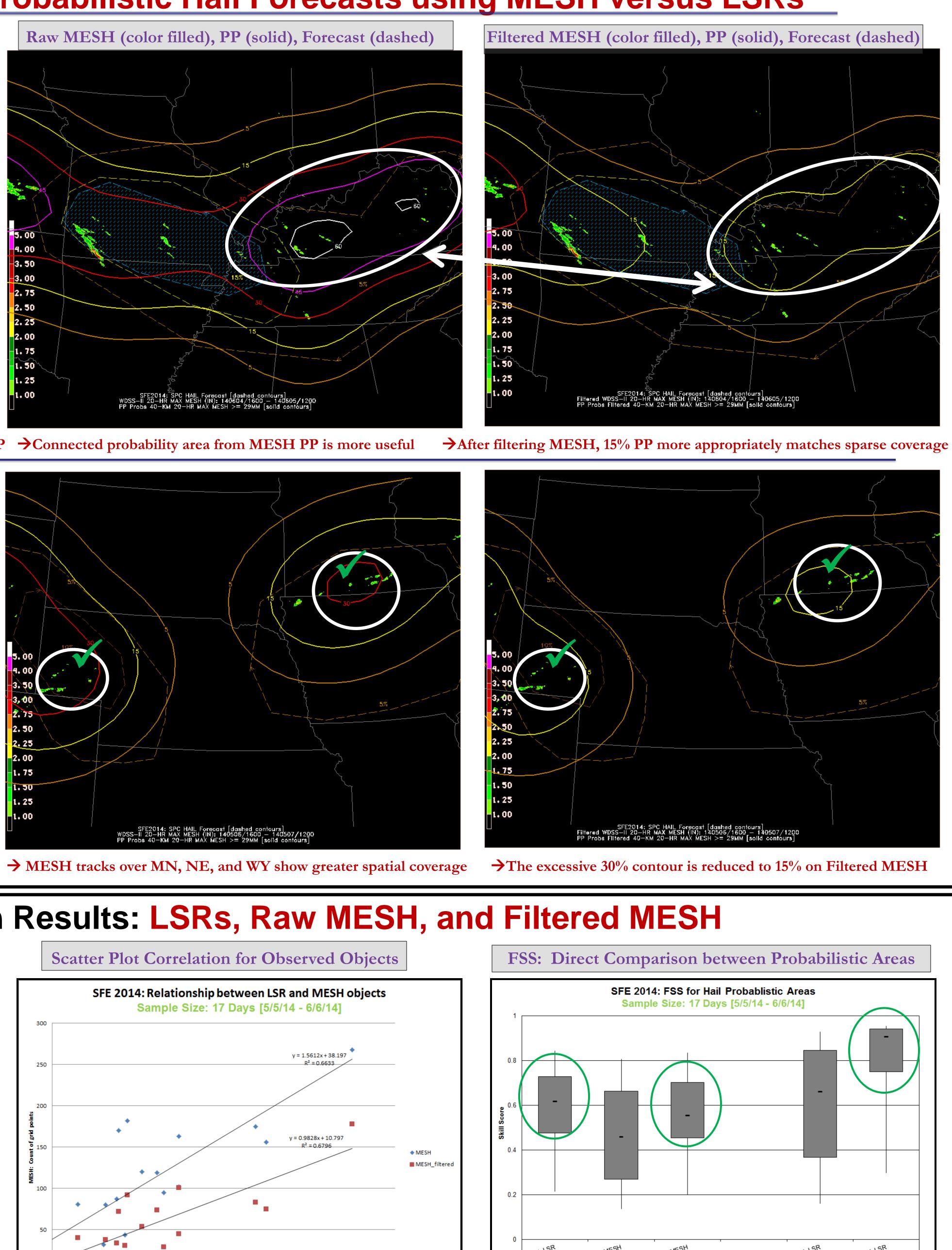
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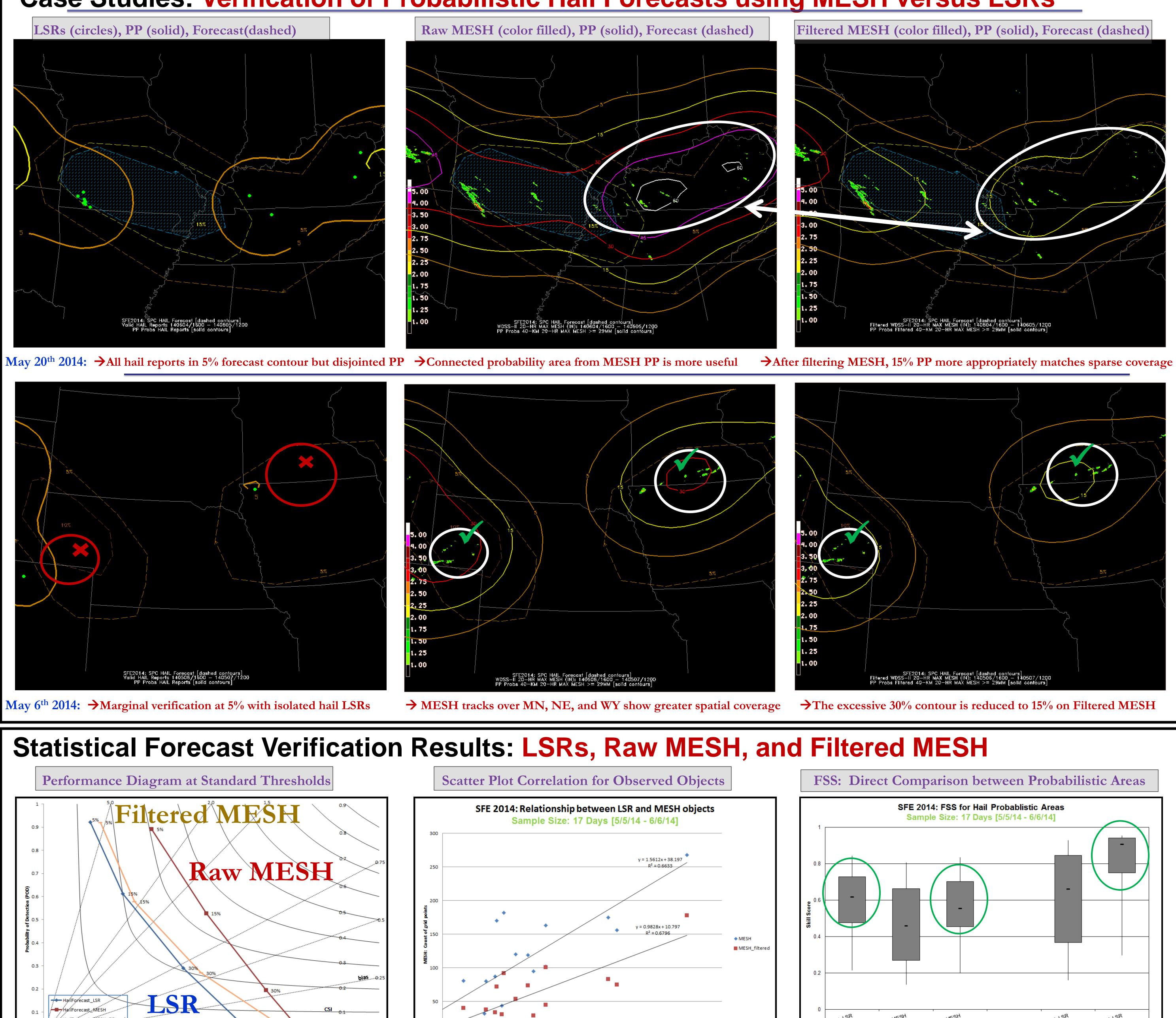
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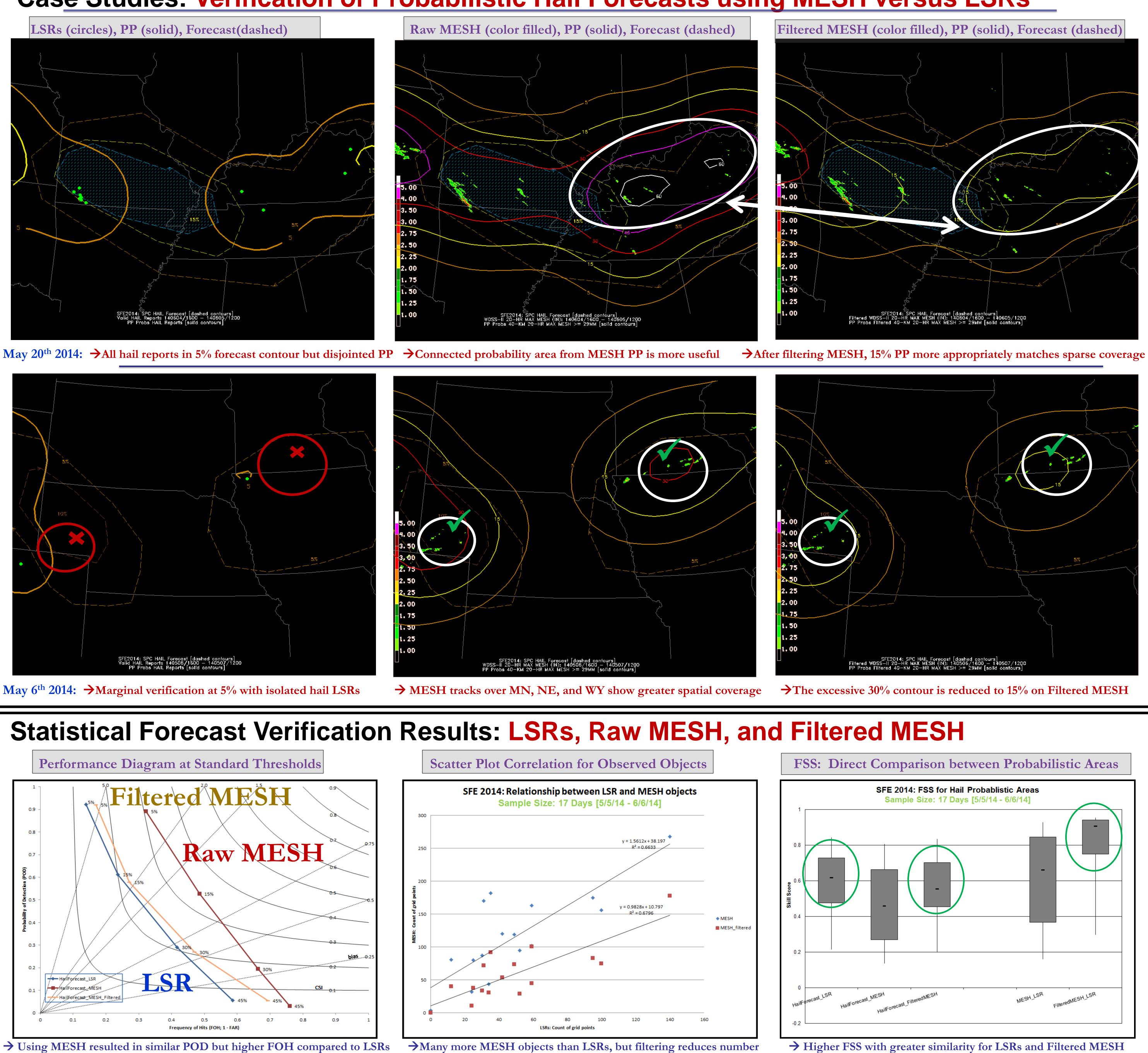


Case Studies: Verification of Probabilistic Hail Forecasts using MESH versus LSRs









Summary and Conclusions

• Objective forecast verification expanded in the 2014 HWT SFE to evaluate experimental probabilistic forecasts for individual severe hazards (tornado, wind, hail) with primary observations coming from LSRs. However, MESH plots were also created to test alternative verifying data.

A more formal comparison of skill using MESH occurred post SFE where a similar procedure to LSRs was followed. The results show MESH to have greater number of observed objects in contrast to severe reports resulting in lower false alarms and higher CSI at all probability thresholds. FSS increased with Filtered MESH as it better agreed with the forecast and LSRs.

UMESH tracks proved to be valuable in identifying events in low-density population areas but also as an independent dataset to supplement hail LSRs. Future efforts will incorporate forecast verification metrics from both LSRs and MESH in a side-by-side diagnosis in subsequent SFE years. In addition, methods to realistically combine both verification datasets will be investigated.



