Introduction

Ensemble forecast sensitivity seeks to describe the relationship between a scalar forecast—referred to as the response function—and initial or early forecast conditions. This relationship is able to highlight features of the flow that are dynamically relevant to the chosen forecast.

Forecast
$$\rightarrow J$$

Initial conditions $\rightarrow x$
sensitivity $= \frac{\partial J}{\partial x} = \frac{cov(J,x)}{var(x)}$

Goals of research

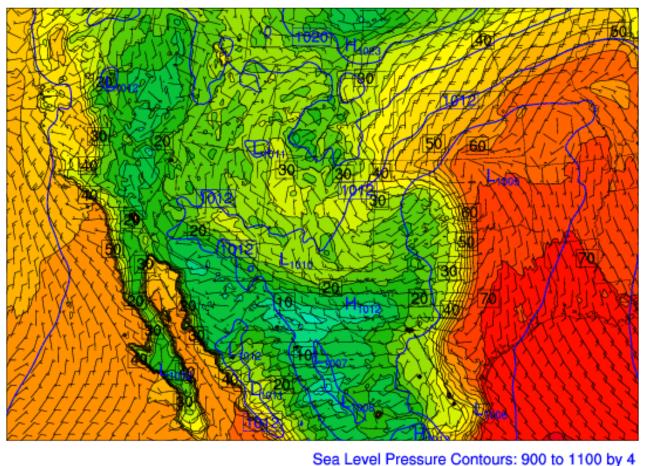
- Explore ensemble sensitivity applied to convection
- What obvious and non-obvious features are important to a convection forecast? Evaluate benefit of a cross-grid sensitivity
- Will relating a high resolution forecast back to a larger and coarser domain highlight synoptic structure and more coherent features?
- Design unique and appropriate response functions for convection What is the best way to represent convection?
- Integrate sensitivity into NWS forecasting process
- Can sensitivity be used to increase forecaster awareness?

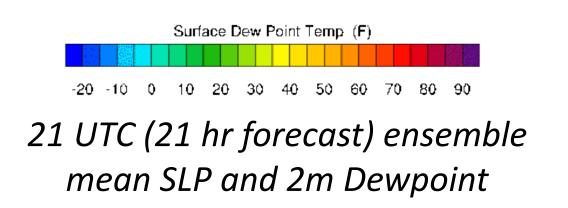
Model and Method

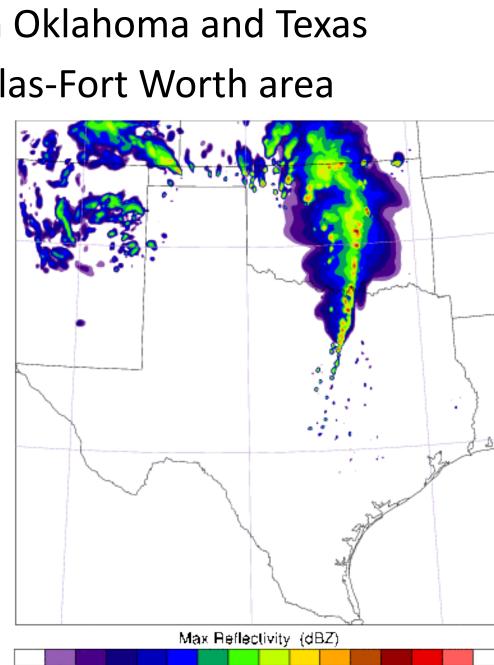
- WRF Version 3.3
- 3 domains: 36, 12, 4 km
 - No cumulus parameterization on 4km domain
- DART EAKF (Anderson et al. 2009; Anderson 2001)
- 50 members
- Configuration reflects TTU Real-time Ensemble (http://www.atmo.ttu.edu/bancell/real_time_ENS/ttuenshome.php)
- Ensemble cycled for 2 days to achieve flow dependent spread prior to forecast
- Observations assimilated every 6 hours • radiosonde, ACARS, satellite wind, METAR, maritime surface, mesonet
- Ensemble forecast initialized at 00UTC 3 April 2012 and run for 36 hours on all domains

3 April 2012 Case Overview

• Convection focused along frontal boundary in Oklahoma and Texas • Many severe storms were reported in the Dallas-Fort Worth area







40 45 50 55 60 65 70 7 Simulated reflectivity from one of the ensemble members

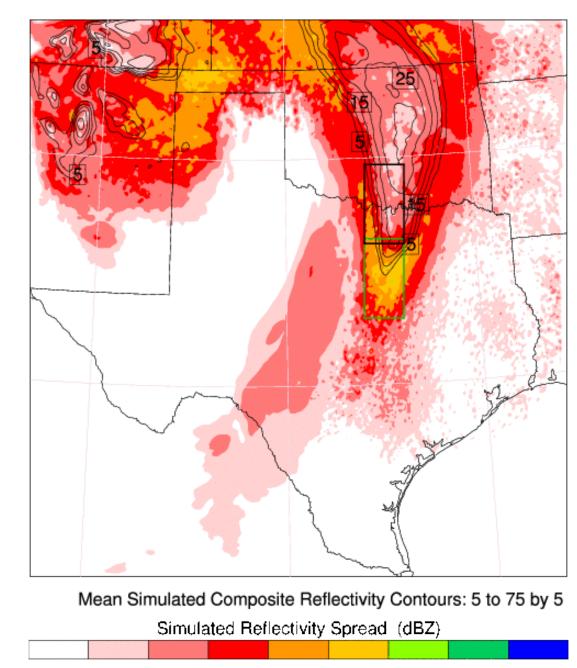


Model domains

Forecast Sensitivity of an April 2012 Southern Plains Convective Event

Christopher Bednarczyk and Brian Ancell Atmospheric Science Group, Department of Geosciences, Texas Tech University, Lubbock, TX Christopher.Bednarczyk@ttu.edu;Brian.Ancell@ttu.edu

Results



Two response function areas were chosen. The first (black) is farther north and located in an area of less spread. The second (green) is positioned in an area of greater ensemble spread. The response function is the area-averaged simulated reflectivity in the box.

5 10 15 20 25 30 35 40 Mean and spread of simulated reflectivity at time of response function (21 UTC)

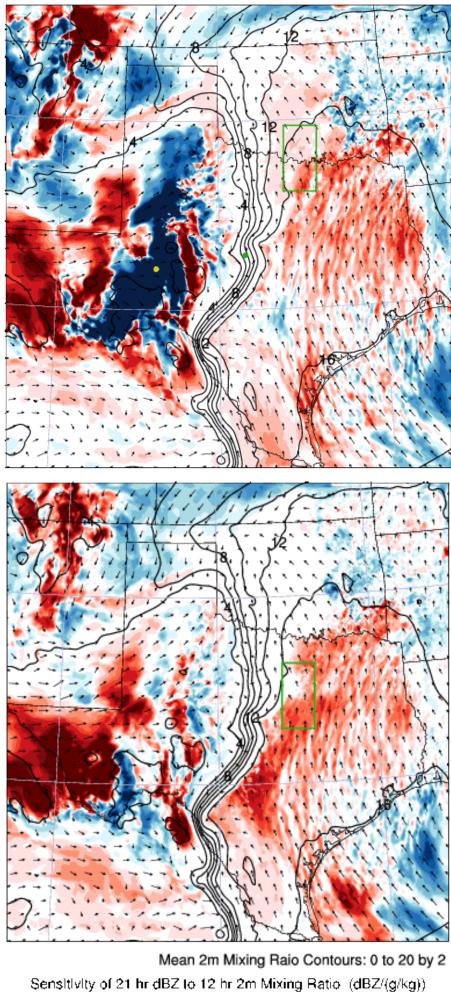
Sensitivity

Both cross-grid and same-grid sensitivities of simulated reflectivity were calculated relative to an early forecast variable (forecast hour 12). Features that are important to the forecast are highlighted by regions of large sensitivity. Positive values indicate that an increase in the early forecast conditions are related to greater values of reflectivity.



Same-grid

4 km domain



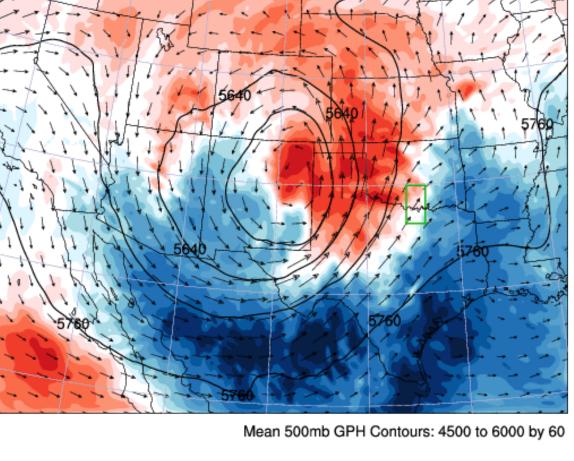
Sensitivity of dBZ to 2m Mixing Ratio

The same and cross-grid sensitivities seem to be highlighting similar features. Two clearly sensitive regions are the warm, moist Gulf air ai the dry air behind the boundary.



Cross-grid

Sensitivity of dBZ to 500mb gph



-1 -.8 -.6 -.4 -.2 0 .2 .4 .6 .8 1

Convection is sensitive to upper-level low position.

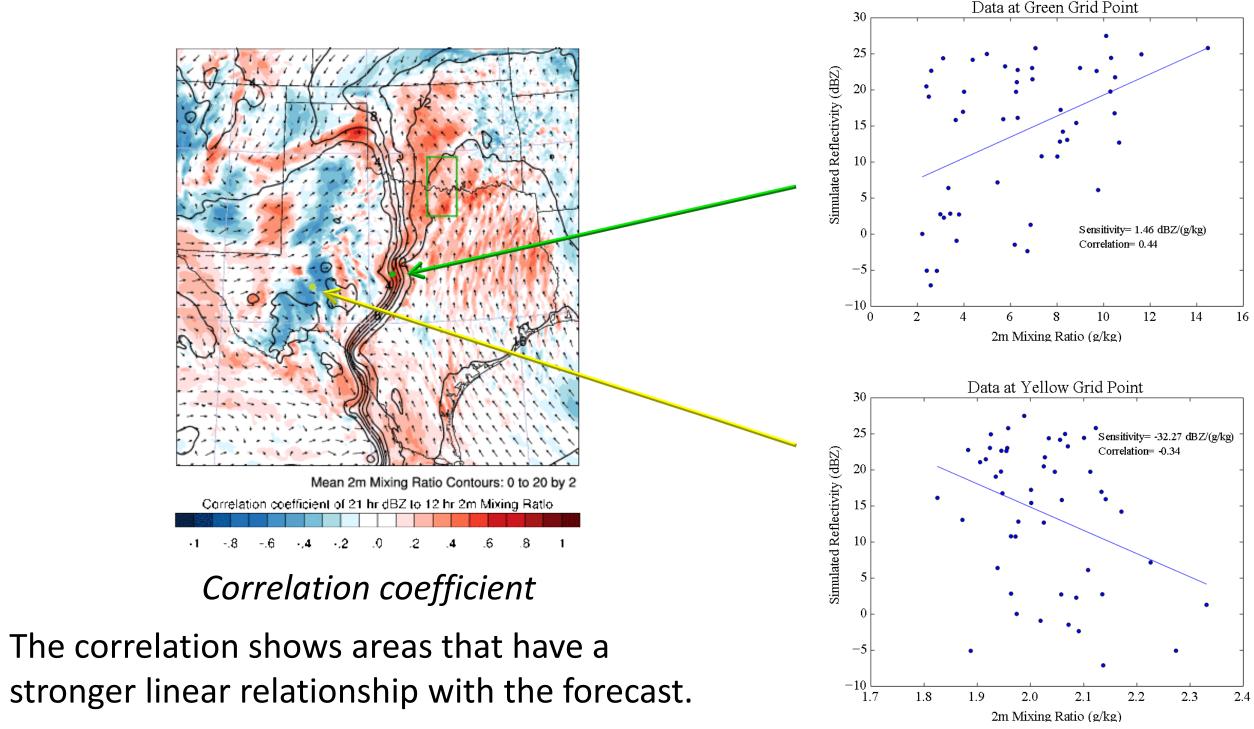
12 km domain

Mean 2m Mixing Raio Contours: 0 to 20 by 2

Sensitivity of 21 hr dBZ to 12 hr 2m Mixing Ratio (dBZ/(g/kg)) 8 12 16 20

Correlation

To see the strength of the linear relationship, correlation coefficient is plotted for the top-left sensitivity plot. Two points were chosen to create scatter plots of the data: one in an area of high sensitivity and moderate correlation, and one in an area of low sensitivity but higher correlation.



The correlation shows areas that have a

Conclusions

- importance to the forecast.
 - Mexico if sensitivity is animated in time.
 - than 200 km.

Future Work

- Continue exploring different response functions
- Lagrangian perspective of response function

Acknowledgments/References

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Targeting. Mon. Wea. Rev., 135, 4117–4134. Anderson, Jeffrey L., 2001: An Ensemble Adjustment Kalman Filter for Data Assimilation. *Mon. Wea. Rev.*, **129**, 2884–2903. Anderson, Jeffrey, Tim Hoar, Kevin Raeder, Hui Liu, Nancy Collins, Ryan Torn, Avelino Avellano, 2009: The Data Assimilation Research Testbed: A Community Facility. Bull. Amer. Meteor. Soc., 90, 1283–1296.

Torn, Ryan D., Gregory J. Hakim, 2008: Ensemble-Based Sensitivity Analysis. Mon. Wea. Rev., 136, 663–677. Torn, Ryan D., 2010: Ensemble-Based Sensitivity Analysis Applied to African Easterly Waves. *Wea. Forecasting*, **25**, 61–78. Torn, Ryan D., Gregory J. Hakim, 2009: Initial Condition Sensitivity of Western Pacific Extratropical Transitions Determined Using Ensemble-Based Sensitivity Analysis. *Mon. Wea. Rev.*, **137**, 3388–3406.



Scatterplots of ensemble members for two model grid points

Sensitivity to moisture showed strong values in the warm, moist air, which was expected. However, the largest sensitivity values were located behind the boundary in the dry air, suggesting that this area has a greater

• The area of large negative sensitivity behind the boundary for the northern response function could be highlighting the importance of downward mixing of air. The positive-negative dipole can be traced from western New

There are significant differences in the dry air sensitivity region between the two response function locations even though their centers differ by less

These could be described as second-order contributors; they do not have a role in the actual convection but may have an indirect influence • Low sensitivity yet high correlation values were found along the boundary

• Is its position 9 hours prior not as important as other features?

Perform tests on sensitivity for statistical significance

Move location relative to specific features such as a front

Ancell, Brian, Gregory J. Hakim, 2007: Comparing Adjoint- and Ensemble-Sensitivity Analysis with Applications to Observation