Satellite Data Assimilation within a Prototype Warn-on-Forecast System

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Warn-on-Forecast Project

- Provide accurate, short range (0-3 h) probabilistic forecasts of severe convective storms is a key component of the Warn-on-Forecast project
- A regional, convection allowing ensemble model is essential to achieving this goal
 - Requires hi-resolution observations of convection and the near storm environment
 - Robust methods to assimilate high resolution remote sensing observations from multiple platforms and sensors
- Model-based probabilistic forecasts can aid in severe weather warning guidance and leading to significantly improved lead times
- Move from observation based warnings to a mix of observations and probabilistic forecasts

Satellite Data Assimilation

• GOES-Imager observations are suitable for storm-scale DA

- High temporal and spatial resolution, low data latency
- Supplements radar observations from WSR-88D Doppler radars
- WSR-88D not as sensitive to non-precipitating clouds
 - $\,\cdot\,$ Low-level stratus, cirrus outflow from storms
 - * Developing convection during CI
- *Clear-air reflectivity* is not the same as *cloud-free* radiances / retrievals
- Future GOES-R data will provide additional and higher resolution cloud products
 - Launches in November 2016
 - First useable data available Spring 2017
- Important consideration: Assimilating satellite observations must be able to show skill in high impact weather forecasting compared to only assimilating radar data

NSSL Experimental Warn-on-Forecast System for ensembles (NEWS-e)

- WRF-ARW: v3.6.1
 - Convection permitting horizontal resolution: 3 km, 51 vertical levels

• 36-member ensemble with physics diversity

- Cloud microphysics: Thompson
- PBL: YSU, MYJ, MYNN2
- Radiation (SW/LW): Dudhia/RRTM, RRTMG/RRTMG

• RAP Land Surface Model, 9 soil levels

• **IC/BCs** use members of an experimental HRRR ensemble generated by GSD run during the spring 2016.

Data assimilation (DA) procedure:

- DART parallel ensemble adjustment Kalman filter
- Prior adaptive inflation applied to state
- Gaspari and Cohn spatial vertical and horizontal localizations
- Localization radius is a function of observation type

http://www.nssl.noaa.gov/projects/wof/news-e/

Storm-scale Data Assimilation

- Assimilation of observations begins at 1800 UTC for each day
 - Continues until 0300 UTC the next day
- 15 minute assimilation cycle with observations partitioned into ±2.5 minute windows
- Assimilated observations:
 - WSR-88D Reflectivity from the MRMS product and Level 2 Doppler radial velocity from all radars in storm-scale domain
 - Liquid and Ice water path (LWP, IWP) retrievals from GOES Imager retrievals
 - All radar and satellite observations objectively analyzed to 6 km resolution (2 delta-X model grid)
 - Oklahoma mesonet observations (grid permitting)
- Additive noise applied to prior state (T, T_D, u, v) where reflectivity observations indicate strong precipitation
- Two sets of experiments are conducted
 - RADONLY: Assimilates only radar and mesonet observations
 - RADSAT: Assimilates radar, mesonet, and satellite observations

Example Cases: 8, 9 May 2016 Severe Weather Reports





- Multiple severe storms in western OK and KS.
- OK storms are mostly hail threats, both left and right movers persist

- Eastward shift of severe convection compared to 8 May
- Several tornados in central and southeast OK
 - One anti-cyclonic tornado

Example Observations



- Radar reflectivity, radial velocity, and satellite LWP and IWP assimilated at a single cycle at 2100 UTC 9 May
- Note the large area of clouds indicated by the satellite observations where no precipitation is detected from the radar
- Total number of observations > 50000

8 May Surface Analysis: 2200 UTC RADONLY RADSAT Assimilating CWP increases cloud cover in the model Cemp analysis Surface temperature 2 lowered Better matches observations Surface Temperature Surface Temperature 18 20 22 24 26 28 30 20 22 24 26 28 30 32 34 36 5 Downward SW Flux Downward SW Flux **GOES-13** Visible Wm 400 500 600 700 800 900 1000 400 500 600 700 800 900 1000 300

8 May Forecast Statistics: 2200 UTC

90-minute forecast verified against OK mesonet observations



Bias (Model - Ob) and RMSE calculated for each ensemble member and mean values plotted. Error bars for RMSE represent the standard deviation of RMSE over all ensemble members.

- Both experiments have a high (warm) bias in solar radiation and surface temperature.
- RADSAT substantially reduces this bias at the analysis time with the impact persisting throughout the forecast period

8 May Low-level Vorticity Forecasts



• RADSAT generates higher vorticity probability swathes in northern KS associated with the multiple tornado reports in this region



GOES-13 Visible

9 May Forecast Statistics: 2000 UTC

90-minute forecasts verified against OK mesonet observations



Bias (Model - Ob) and RMSE calculated for each ensemble member and mean values plotted. Error bars for RMSE represent the standard deviation of RMSE over all ensemble members.

- RADSAT lowers bias and errors with respect to RADONLY experiments, but the magnitude of the difference is not as great as in the 8 May case.
- Note that differences in Arkansas are not included in theses statistics due to lack of observations

9 May Low-level Vorticity Forecasts

Probability of 0-2 km vertical vorticity > 0.004 s⁻¹ 90 minute forecast starting at 2000 UTC



• RADSAT again generates somewhat higher vorticity probabilities than RADONLY associated with southern Oklahoma tornadoes

Summary

• The Good:

- Assimilating CWP improved surface thermodynamic conditions for several experiments
- This improvement generally corresponded to better forecasts of low-level vorticity

• The Bad:

- Using non-thinned radar and satellite observations, some evidence of storm cores weakening
- Assimilating satellite observations also introduces noise in the dynamical fields in some cases.

Fixing the Bad:

- Further analyze model configuration, specifically radar and satellite forward operators
- Determine the optimal combination of satellite and radar data to assimilate.
- Dumping everything into the model without considering the relationship between radar and satellite observations is certainty not optimal – work on adaptive data thinning techniques

Future Considerations

Challenges:

- Assimilating combined radar and satellite observation data remains challenging
- Further research in forward operators, data thinning techniques and new observation types (radiances, polarimetric radar) will be required going forward
- In addition to clouds, rapid updates of aerosol concentrations will also have to be considered

• Plans:

- + Transition ensemble data assimilation system to GSI-EnKF
- Take advantage of all the satellite QC options
- Integrate GOES-R water vapor radiances and atmospheric motion vector into data assimilation system.
- Perform experiments for other event types such as landfalling tropical cyclones and winter weather

Questions

- If you are interesting in participating in this and similar projects, please contact me: <u>Thomas.Jones@noaa.gov</u>
- CIMMS post-doc jobs available: <u>http://cimms.ou.edu/index.php/careers/</u>