



EnKF Assimilation of Cloud Properties Retrieved from GOES

Thomas A. Jones (Thomas.Jones@noaa.gov)¹ and David Stensrud^{1,2} with Patrick Minnis³ and Palikonda Rabindra³

1. Cooperative Institute for Mesoscale Meteorological Studies 2. NOAA/OAR/National Severe Storms Laboratory 3. NASA Langley Research Center

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Objectives:

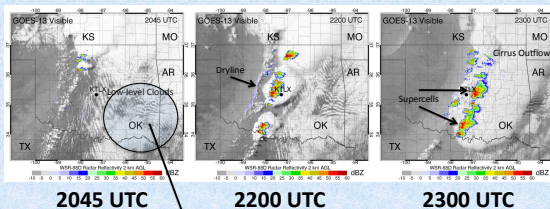
- Assimilate satellite derived cloud properties in hi-resolution NWP models
- Validate that the assimilation improves analysis fields relative to not using satellite data
- Investigate the potential for improved 0 – 3 hour convection forecasts
- Use results as basis for developing procedures for assimilating the array of satellite products to be generated by the GOES-R staring in 2015-16.

Satellite Data:

- GOES-13 (East) 1 km resolution cloud property retrievals from algorithms developed at NASA Langley Research Center.
- Retrieved properties include:
 - Cloud Top Pressure (CTP)
 - Cloud Base Pressure (CBP)
 - Cloud Phase
 - Cloud liquid water (or ice) path (CLWP, CIWP)
- Uncertainty in CTP / CBP is approximately ± 1 km
- Use retrievals to determine locations of cloud vs. cloud-free areas to create cloud and humidity variables suitable for data assimilation

Case Study Characteristics:

- Evaluate satellite data assimilation on a severe weather outbreak from 10 May 2010
- This event produced multiple reports of severe wind, hail, and tornadoes during the late afternoon in Oklahoma (OK) and Kansas (KS)
- Supercells developed ahead of an eastward progressing dryline around 2000 UTC
- Convection moved eastward at speeds of approximately 25 ms^{-1}
- Convection widespread by 2300 UTC generating large cirrus shield in eastern OK



East-west wave pattern in low-level cloud field

Acknowledgements:

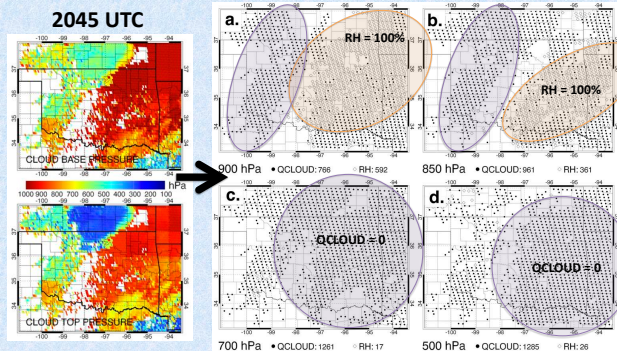
GOES Imager cloud property retrievals were kindly provided by the Langley Research Center in Hampton, Virginia and processed specifically for this case by Rabindra Palikonda. Raw visible GOES imagery and WSR-88D level 2 radar reflectivity were retrieved from National Climate Data Center archives. This research was supported by the NOAA National Environmental Satellite, Data, and Information Service. Partial funding for this research was also provided by NOAA/Office of Oceanic and Atmospheric Research under NOAA–University of Oklahoma Cooperative Agreement NA17RJ1227, under the U.S. Department of Commerce.

Model Characteristics:

- Advanced Weather Research Forecast (WRF-ARW) forecast model version 3.2.1
- Thompson microphysics, MYJ boundary layer physics, Noah LSM
- Use Ensemble Kalman Filter (EnKF) data assimilation approach: 36 members
- Assimilate traditional observations (surface, marine, aircraft, and radiosondes) into 15 km resolution mesoscale run at hourly intervals between 1200 – 2100 UTC 10 May 2010
- Assimilate cloud and humidity variables at 15 minute intervals starting at 1800 UTC in nested 3 km domain using mesoscale analyses as boundary conditions
- Generate forecast output at 15 minute intervals starting at 2100 UTC
- Compare two experiments: one that assimilates the satellite derived variables (CLD) and an otherwise identical run that does not (NOCLD)

Satellite Retrievals to Model Variables:

- Retrieved products such as CTP, CBP, CLWP are not suitable for direct assimilation into model as they are strictly 2-D variables where 3-D information is needed
- Instead, use retrieved products to generate cloud and humidity variables corresponding to WRF model state variables:
 - Cloud liquid water content (QCLWD)
 - Cloud ice content (QICE)
 - Graupel (QGRAUP)
 - Cloud Rain (QRAIN)
 - Relative Humidity (RH)
- Create new 3-D variables using same horizontal resolution as satellite data with a vertical resolution of 50 hPa from 950 – 200 hPa
- If no clouds are present: $Q^* = 0.0$
- Where CBP > pressure > CTP: $RH = 100\%$
- Translating CLWP/CIWP into Q^* variables where clouds exist is left for future research



Assimilated Data at 2045 UTC:

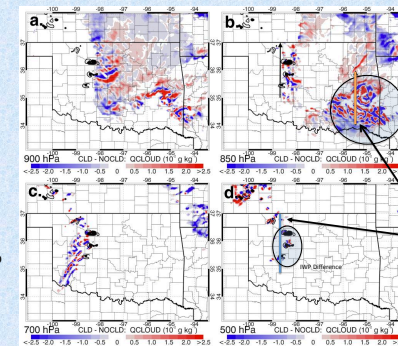
- Large number of null cloud microphysical variables assimilated (e.g. $QCLWD = 0$)
- Primarily in western OK, southern KS below 850 hPa, everywhere above
- No clouds detected at these locations and/or levels from GOES data
- Relative Humidity ($RH = 100\%$) assimilation confined to low-levels in eastern OK
- Corresponds to cloud field observed in GOES visible imagery
- Other assimilation times show similar characteristics

VARIABLE	N _{total}	N _{nonnull}	%nonnull
QCLWD	205492	187042	91.0
QICE	205492	190796	92.8
QRAIN	205492	190388	92.6
QGRAUP	205492	195329	95.1
RH	37188	22875	61.1
TOTAL	859117	789967	91.9

Total 18-2100 UTC Sample Size

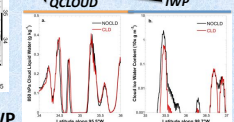
CLD minus NOCLD Cloud Liquid Water (QCLWD) at 2045 UTC

• RED = CLD greater BLUE = NOCLD greater



Results:

- $QCLWD$ Differences greatest at lower levels
- Wave-like difference pattern in southeast OK similar to the cloud pattern observed by GOES-13
- Further differences in west-central OK near developing convection
- Small differences in IWP analysis also present (black shading)

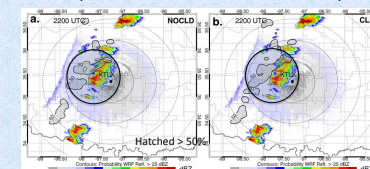


South-North cross sections of QCLWD and IWP

- Small differences in location (5 – 10 km) and magnitudes between CLD and NOCLD

Reflectivity Probability Forecast at 2200 UTC

- Probability of ensemble mean simulated reflectivity > 25 dBZ



Color contours = KTLX WSR-88D Reflectivity at 2 km AGL

Results:

- Forecast reflectivity west of observations for both models
- Both capture central OK storm
- CLD generates higher probabilities in south-western OK
- Neither model skillful over the other

Summary:

- Assimilating GOES derived cloud and humidity variables produced a measurable effect on model analysis and forecast fields
- Greatest differences occur with water clouds at or below 850 hPa
- Interesting match in $QCLWD$ difference patterns and GOES visible imagery
- One hour simulated reflectivity forecast changed, but no significant increase in skill produced

Ongoing Work:

- Derive and assimilate vertical profiles of CLW and CIW from satellite data
- Determine appropriate cloud microphysical schemes to implement
- Test different vertical and horizontal localization radii
- Determine impacts of assumptions about cloud property uncertainties and errors
 - Very non-Gaussian
- Use idealized storm and environment to tune model to take full advantage of satellite data