Using Satellite All-Sky Infrared Brightness Temperatures for Model Verification

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With contributions from many people!

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Utility of Satellite Brightness Temperatures

- Sensors provide detailed information about atmosphere and land/ocean surfaces
 - Obtained from visible, shortwave, infrared, and microwave bands
- Our studies have focused on all-sky infrared brightness temperatures because they provide valuable information about clouds and water vapor, both of which are susceptible to large errors in NWP model forecasts
- Geostationary satellites provide routine coverage over large areas, whereas polarorbiting satellites provide global coverage, but with less frequent updates

Project Description – Model Verification

• Results from a 24-hr case study from July 2015 were used to assess merits of different verification methods

- Assess HRRR forecast accuracy using GOES infrared brightness temperatures
 - Primary focus is to assess accuracy of cloud and moisture fields using observations from the 10.7 μm window band that are sensitive to clouds and surface temperatures
 - Accuracy assessed using grid-point statistics (RMSE), neighborhood methods (Fractions Skill Score), and object-based tools (e.g., MODE)
- Use of traditional, neighborhood, and object-based verification methods should provide a more complete analysis of the forecast accuracy

Case Study Description

SPC Storm Reports from 12 UTC on July 23 to 12 UTC on July 24 2015



23-24 July 2015 chosen because different types of convection were observed across the U.S., including:
Surface Low and Frontal

passage, Isolated Convection, and Stationary Front and Sea Breeze

Case Study – Fractions Skill Score (FSS)

SPC Storm Reports from 12 UTC on July 23 to 12 UTC on July 24 2015



Sector

North Plains

Central Plains

Southeast

• Threshold: 10% GOES and HRRR BT threshold applied, respectively

• Box plot includes forecast lead times between 00h and 12h

• Box width where FSS exceeds 0.5 is scale where HRRR contains useful information

• Lowest skill associated with the single cell convection, indicating most difficult to accurately forecast

MODE Configuration – Seasonal Analysis

Object Pair Attribute	Weight (%)	Description
Centroid Distance	25.0	Distance between objects' "center of mass"
Boundary Distance	18.75	Minimum distance between the objects
Convex Hull Distance	6.25	Minimum distance between the polygons surrounding the objects
Angle Difference	6.25	Orientation angle difference
Area Ratio	25.0	Ratio of the forecast and observation objects' areas (lowest value)
Intersection Area Ratio	18.75	Ratio of observation (forecast) object to the objects' intersection area (highest value)

• Equal weight given to object distance and size (area ratio) attributes

 Boundary Distance has a lower weight than the Centroid Distance so that more weight was placed on the displacement between the object's center of mass

Intersection Area Ratio
 lower than Area Ratio
 because it can be artificially
 high when a small object is
 fully enclosed by a larger
 object

Case Study Time Periods



• Summer Example: Forecast hours 0-24 from HRRR initializations from August 1-31, 2015 • Winter Example: Forecast hours 0-24 from HRRR initializations from January 1-31, 2016

Number of MODE Objects – Function of Time of Day



Number of MODE Objects based on Time of Day for

- MODE identifies more cloud objects during August
 - Average cloud object is smaller (larger) during August (January)
- Diurnal cycle is much larger during August
 - Minimum (maximum) near 12 UTC (20 UTC)

Number of MODE Objects – Function of Forecast Hour



Number of MODE Objects based on HRRRx Forecast Hour for

- Too many forecast cloud objects during August for forecast hours 0-1
 - Indicates cloud objects are too small in HRRR initializations
- More observed cloud objects than HRRR forecast objects overall
- Steady drift toward fewer forecast cloud objects during January

Assessing the Impact of Land Surface Models on Convection over the Southeastern United States Using High-Resolution Model Simulations and GOES-16 Satellite Observations

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Convective Initiation Over Southeastern United States

- Long-standing challenge to forecast the location and timing of convective initiation that could lead to severe thunderstorms
 - Southeast U.S., typically form and develop long-lived storms without synoptic forcing
 - Isolated storms make up a significant portion of the spring and summer precipitation totals
 - Numerical models struggle to predict the location and timing of this convection



Geostationary Satellites Important to Study CI



-0.2

0.0

-0.4

0.2

Mixing Ratio [g/kg]

0.4

0.8

0.6

1.0

Rev, 149, 1153-1172.

Geostationary Satellites Important to Study CI



• We can compare cloud objects identified in satellite imagery to similar objects in models to compare cloud characteristics, such as growth rates and brightness temperature distributions



Geostationary Satellites Important to Study CI



SNOW

GRAUP CLOUD

RAIN

Thompson

1.0

0.8

0.6

-0.2

-0.4

0.2

Mixing Ratio [g/kg]

0.4

0.0

to the observations, we can then use the model simulations to evaluate processes that lead to differences in convective initiation when using different model parameterization schemes

WRF Model Configuration



WRF ARW V3.9.1.1 simulations: 12 UTC to 23 UTC allowing 6 hours of model spin-up

- Model data output every 5 minutes to match GOES-16
- Inner domain: dX = 500 meter, 53 vertical levels up to 40 hPa
- Initial conditions: NCEP Final (NCEP FNL) at 0.25-degree

Microphysics:	Thompson	
PBL Scheme:	MYNN	
Surface Scheme:	NOAH LSM and NOAH-MP LSM	

Simulated GOES-16 ABI infrared brightness temperatures computed using the Community Radiative Transfer Model

Object-based Identification; Henderson et al. (2021)

Cloud objects are identified using 10.3 µm brightness temperatures by adapting methods from TOOCAN; Fiolleau and Roca, 2013



Comparing Cloud Top Characteristics CI Cloud Objects



Time [UTC]

Evolution of 10.35 µm Cloud-top Brightness Temperatures – Cloud Growth

30 minutes before and after CI is detected



Noah-MP clouds continue to cool after CI; most notable from 15 min onward Clouds that reach CI in Noah-MP more likely to develop into deep convection Noah-MP leads to more accurate cloud growth rates

Sensible Heat Fluxes [W m⁻²]



- Sensible heat fluxes are larger in general for the Noah-MP simulation in forested regions, but they are smaller in cropland and grassland areas, such as the north-south band of lower heat fluxes through the middle of the domain
- There is greater spatial heterogeneity in the sensible heat fluxes in the Noah-MP simulation

Impacts of Surface Energy Balance on Local Circulation [Noah-MP – Noah]



Differences in surface radiative balance leads to differences in the local circulation

Impacts on hydrometeor mixing ratios in CI cloud objects [Noah-MP – Noah]



Stronger updrafts and mass flux in the Noah-MP simulation led to substantially more cloud content, especially after 1800 UTC Higher ice formation helps confirm the need for a latent heat boost to sustain cloud growth beyond CI detection

Summary of the Convective Initiation Study

- Differences in the land surface models impact the magnitude and spatial heterogeneity of the sensible and latent heating components of the surface radiation balance that in turn drive differences in local circulation patterns
- These differences then impact the characteristics of the updrafts, the spatial and temporal evolution of the hydrometeor mixing ratios, and the growth of the convective clouds
- Results also show that more research is necessary to improve the accuracy of land surface models
 - We are working on a NOAA project that is using surface sensible and latent heat flux measurements to examine the accuracy of UFS S2S forecasts