

Evaluation of Unified Forecast System Tropical Cyclone Quantitative Precipitation Forecasts

Kathryn Newman^{1,2}, B. Nelson^{1,2}, L. Pan^{2,3,4}, M. Biswas^{1,2}, E. Grell^{2,4,5}, W. Li^{1,2}

¹ National Center for Atmospheric Research
 ² Developmental Testbed Center
 ³ NOAA/Global Systems Laboratory
 ⁴ NOAA/Physical Sciences Laboratory
 ⁵ CU Cooperative Institute for Research in Environmental Sciences



Overview

Introduction

- Quantitative precipitation forecast (QPF) verification provides insight on both storm structure and total precipitation, which is useful for understanding model processes
 - Microphysics, PBL, and other parameterizations & interactions between parameterizations
 - Establishing tools for large sample evaluations of QPF enables regular assessments
- Provide assessment of recently operational UFS-based Hurricane Analysis and Forecast System (HAFS) for TC precipitation forecasts
 - Storm focused evaluation using various methods
 - Over land and water
- HAFSv1 went operational 27 June 2023
 - Two configurations replacing operational Hurricane Weather Research and Forecast (HWRF) and Multi-scale Ocean-coupled Non-hydrostatic Model (HMON)

Fig1: HAFSv1 domain for NATL/EPAC basins, Mehra et al. 2023

NATL/EPAC Basins

-140 -120 -100

Overview

Model configurations

- HAFSv1 (HFSA, HFSB) evaluated for all 2021-2022 N. Atlantic basin storms
 - Evaluation using parent domain (6 km), masking for storm region
 - Operational baseline: HWRF

	HFSA	HFSB	HWRF
Land surface	Noah	Noah	Noah
Surface layer	GFS, HWRF TC-specific sea surface roughness	GFS, HWRF TC-specific sea surface roughness	GFDL surface layer (updated)
Boundary layer	SA-TKE-EDMF, TC-related calibration, mixing length tuning	SA-TKE-EDMF, TC-related calibration, tc_pbl=1, mixing length tuning	GFS-EDMF
Microphysics	GFDL single-moment	Thompson double-moment	Ferrier-Aligo
Radiation	RRTMG	RRTMG	modified RRTMG
Convection	Scale-aware SAS calibrated entrainment	Scale-aware SAS	scale-aware SAS

3

Overview

Methodology

- Enhanced Model Evaluation Tools (METplus)
 - Tools: gen_vx_mask, regrid_data_plane, PCP-combine, Grid-stat, TC-RMW, MODE
- 6 hour precipitation accumulations, track shifting, land/water and storm based masking
 - 600-km mask around best track for each valid time
- Observations:
 - Integrated Multi-satellitE Retrievals for GPM (IMERG) verification over water
 - 1/10 deg, satellite precipitation product combining active, passive microwave, and geostationary satellite data
 - Climatology-Calibrated Precipitation Analysis (CCPA) verification over land
 - 5-km gauge corrected radar observation product (combines gauge analysis + stage IV)
- All model/observations re-gridded to common grid





Newman et al. 2023



ETS: 0 = no-skill, 1 = perfect forecast

Grid-based QPF

Equitable Threat Score



- Impact from shifting less when there are many grid cells with precipitation (low thresholds)
- Shifting helps stabilize skill scores at longer lead times
- Low skill scores: issues with ETS calculation of random chance adjustment with many rainy grid cells over a small domain (Wang et al. 2014)



Frequency Bias



• Over forecast precipitation for lower thresholds, under forecast for larger thresholds

• Shifting does not impact results - exception of HWRF at lowest thresholds (potentially large track errors)

FBIAS: "good forecast" = 1,
> 1 is forecasted too frequent,
< 1 is not forecasted frequently enough</pre>

UIFCW

A UFS Collaboration Powered by EPIC

6

Equitable threat score by threshold (over land)

2021-2022 Season: 6 hr APCP



HFSA 0.1 ···· HFSA 0.5 ··· HFSA 1.0 ··· HFSA 1.5 ··· HFSA 2.5
 HFSA 3.5 ··· HFSA 5.0



2021-2022 Season: 6 hr APCP

Shifted tracks verified against CCPA





- Large thresholds & lowest (>= 0.1") have lowest skill
- Intermediate (>= 0.5-1.5") perform better for HAFS configurations
- HWRF skill more stable by lead time



Equitable threat score by threshold (over water)

2021-2022 Season: 6 hr APCP



2021-2022 Season: 6 hr APCP

← HFSA 0.1 → HFSA 0.5 → HFSA 1.0 → HFSA 1.5 → HFSA 2.5 ← HFSA 3.5 → HFSA 5.0



Shifted tracks verified against IMERG



2021-2022 Season: 6 hr APCP

→ HWRF 0.1 → HWRF 0.5 → HWRF 1.0 → HWRF 1.5 → HWRF 2.5 → HWRF 3.5 → HWRF 5.0

- >= 0.1" thresholds: lowest skill (ETS calculation)
- Track shifting results in fairly constant skill throughout forecast



Frequency Bias by threshold (over land)

2021-2022 Season: 6 hr APCP



HFSA 0.1 ---- HFSA 0.5 ---- HFSA 1.0 ---- HFSA 1.5 ---- HFSA 2.5
 HFSA 3.5 ---- HFSA 5.0





---- HFSB 0.1 ---- HFSB 0.5 ---- HFSB 1.0 ---- HFSB 1.5 ---- HFSB 2.5 ---- HFSB 3.5 ---- HFSB 5.0 Shifted tracks verified against CCPA





- Largest thresholds perform well, near 1.0
- Smaller thresholds tend to over forecast precipitation for all models/configurations

2021-2022 Season: 6 hr APCP

Frequency Bias by threshold (over water)



2021-2022 Season: 6 hr APCP

HFSA 0.1 → HFSA 0.5 → HFSA 1.0 → HFSA 1.5 → HFSA 2.5
 HFSA 3.5 → HFSA 5.0

Forecast Lead Time (hours)

0.2

6 18 multi thresh, AL, IMERG 30

42 60 72 84 96







2021-2022 Season: 6 hr APCP



A UFS Collaboration Powered by EPIC

UIFCW7

Largest thresholds tend to under forecast precipitation for all models/configurations

108 120

• Smaller thresholds tend to over forecast precipitation for all models/configurations



Hurricane Ian

12-hr forecasts: Histograms by RMW

bins of 0.4 RMW

HFSB HFSA HWRF **6 hr Precipitation Accumulation 6 hr Precipitation Accumulation 6 hr Precipitation Accumulation** 160 12 hr Forecast 12 hr Forecast 160 12 hr Forecas 160 IMERG IMERG IMERG 140 140 140 Accumulation (mm) 09 00 09 00 09 00 Accumulation (mm) 09 00 09 00 09 00 40 20 20 20 1.4 1.8 2.2 2.6 3 3.4 3.8 4.2 4.6 5 5.4 5.8 0.2 0.6 1 0.2 0.6 1 1.4 1.8 2.2 2.6 3 3.4 3.8 4.2 4.6 5 5.4 5.8 0.2 0.6 1 1.4 1.8 2.2 2.6 3 3.4 3.8 4.2 4.6 5 5.4 5.8 Distance (RMW) Distance (RMW) Distance (RMW)

- HAFS gradient moving from center better match IMERG
- HWRF has higher intensities closer to the RMW with a steep drop after about 2-3 RMW

UIFCW 7

A UFS Collaboration Powered by EPIC

Hurricane Ian

Method for Object Based Evaluation (MODE)

MODE object identification algorithm mimics subjective matching of observed and forecasted objects using a multistep process and fuzzy logic engine

MODE: APCP at A6 vs APCP at A6



MODE: APCP at A6 vs APCP at A6 Observation



Observation Objects with Forecast Outlines



- Example output from MODE algorithm: forecast and observed 6-hr acc precipitation
- Objects identified by the MODE algorithm: red observations, blue output model



Hurricane Ian

MODE: 6-hr precipitation accumulation PDF (log frequency)

Includes all grid points with precipitation within object

UIFCW 2023

A UFS Collaboration Powered by EPIC



- HAFS (HFSA, HFSB): more light precipitation, lower through most of the distribution
- HWRF: less light precipitation, more heavy precipitation (likely due to over forecast near the core)

Conclusions

- The more complex microphysics in the HAFS configurations better represent the tropical cyclone (TC) precipitation and the features of the TC
- HAFSv1 configurations tend to over forecast precipitation for smaller thresholds and under forecast precipitation for larger thresholds
- Considerations are needed for assessing skill for lowest thresholds for smaller verification domain with high number of precipitating grid cells

HFSA and HFSB retrospective runs were conducted by NOAA EMC and HRD hurricane teams

Kathryn Newman National Center for Atmospheric Research knewman@ucar.edu

A UFS Collaboration Powered by EPIC

DTC Visitor Program

https://dtcenter.org/visitor-program

Propose a project to work on with us!

Two types of visitor projects:

- PI Up to 2 months salary, travel and per diem can be split into multiple visits
- Graduate Student Up to 1 year of temporary living per diem and travel expenses for graduate student, plus support for advisor visits
- See Announcement of Opportunity on DTC website for more information on how to apply and guidance on topics of interest





