

NATIONAL WEATHER SERVICE

HAFS Ensemble in Real Time on AWS Cloud

Presented by Zhan Zhang, Jiayi Peng, Rajendra Panda, Weiguo Wang, Bin Liu, Yonghui Weng, Avichal Mehra, Vijay Tallapragada Matt Long, Unni Kirandumkara William Komaromi, Aaron Poyer, Xuejin Zhang, Sundararaman Gopalakrishnan

@EMC Hurricane Team; @parallelworks.com; @HFIP management Team; @NOAA Cloud Team

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Outline

- Hurricane Ensemble in Real-time on the Cloud (HERC)
 - Project goals
 - Configuration
- HERC in 2023 Hurricane Season
 - Products
 - Performance
 - Statistical Characteristics
- Future plan
- Porting HERC on AWS Cloud



Hurricane Ensemble in Real-time on the Cloud (HERC)

Project Goals

- Port HAFS on AWS cloud
- Improve deterministic and probabilistic model forecast products
- Deliver TC ensemble predictions to NHC forecasters in real-time
- Proof of concept for operational hurricane ensemble system on the cloud







HERC Configuration for 2023 Real-time

- Basic configuration, based on HAFSv1
 - Lower horizontal resolution: ~6 km (vs. ~6/2 km), Lower vertical resolution: L66 (vs. L81), Single domain with no nest
 - No vortex initialization and Data assimilation
 - Ocean: Coupled with HYCOM ocean model
 - One unperturbed control member and 20 perturbed ensemble members
 - Runs 4 times a day (00Z/06Z/12Z/18Z), storms in Atlantic basin only
- Computer resources: AWS Cloud, details see next slide
- □IC/BC Perturbations:
 - IC/BC: GEFS grib2 (0.5x0.5) vs. GFS
- **OMdel Physics Perturbations:**
 - Stochastically perturbed physics tendencies (SPPT)
 - Stochastic kinetic energy backscatter (SKEB)
 - Stochastically perturbed PBL humidity (SHUM)





HAFS Ensemble Products





HAFS Ensemble Products

https://www.emc.ncep.noaa.gov/HAFS/HAFSEPS/index.php



RI happened before Idalia's landfall.



Wind Speed Probability (>34kts; 50kts; 64kts)



HAFS ensemble surface wind fields could be applied to drive the WW3, P-surge model, etc., for hazards forecasts.



Precipitation Probability (> 1;4;8 inches)



https://www.emc.ncep.noaa.gov/HAFS/HAFSEPS/index.php



Intensity Error and Bias (HAFSens VS. GEFS/EENS)



HAFSens: reduced intensity errors in first 4 days; reduced bias by 10kts.



Track Error and Cross-track Bias (HAFSens VS. GEFS/EENS)



HAFSens: track error increased, most contributed by "Philippe (AL17-2023)".







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HAFS ensemble VS. GEFS ensemble Fujiwhara Effect: *Philippe (AL17) and Rina (AL18)*



Intensity and Track Error/Spread Relationship





GEFS error (solid blue)) GEFS spread (dashed blue) HAFSens error (solid red) HAFSens spread (dashed red)

Under dispersive for HAFSens track/intensity forecasts



Intensity Error/Bias, Track Error and Cross-track Bias (HAFSens VS. HFSA/B)



HAFS ens (HEMN):

Intensity error is close or better than HFSA after 24 hours. Initial large intensity error /bias due to "no VI/DA". HAFS-ens has larger cross-track bias







Intensity Error/Bias and Track Error (HAFSens VS. CTCXens)





CTCXens control member (XC00) CTCXens 20-member mean (XEMN) HAFS control member (HC00) HAFS 20-member mean (HEMN)

CTCXens did better job for intensity forecast. HAFSens has bigger negative intensity bias in the first 72-hours.



Intensity and Track Error/Spread Relationship



NORA Absolute Intensity Error/Spread (knots) BASIN (AL 2023) 20.0 (knots) CTCXerror - HENSerror 17.5 CTCXspread HENSspread σ Ba 15.0 l'SI 12.5 盀 10.0 Intensity 7.5 5.0 Absolute Track 2.5 0.0 0 12 24 36 48 60 72 84 96 108 120 # of Cases 186 162 124 111 101 84 73 174 149 138 93 Forecast Hour

CTCXens error (solid blue)) CTCXens spread (dashed blue) HAFSens error (solid red) HAFSens spread (dashed red)

Under dispersive for HAFSens track/intensity forecasts



Continuous Ranked Probability Score Track and Intensity



The smaller CRPS indicates better forecast



All 2023 NATL Storms



Statistical Characteristics

Track/Intensity Rank Histogram (2023 NATL hurricanes)



Rank histogram shows the frequency of best track value that falls into each ranked bin



Future Plan for FY24 and beyond

- Increase ensemble membership to improve uncertainty spectrums
- Apply re-center process to remove IC perturbation bias.
- Conduct TC VI and DA to improve the initial storm structures
- Improve initial and physics perturbations to have better ensemble spread
- Perform cluster analysis to better represent ensemble forecasts
- Explore use of nested grids for HERC
- Use ML/AI methods for surrogate ensemble members



Porting HERC on AWS Cloud



HAFS ensemble on AWS

- Objective develop a probabilistic hurricane forecast model using HAFS
 - O HAFS (Hurricane Analysis and Forecast System) is a state of the art, deterministic model
 - Run in real-time, a 21-member ensemble using HAFS during the 2023 Atlantic hurricane season (August to November)
- Requirements features
 - O Operational HAFS has many features
 - Ocean coupling, 2km resolution, moving nests, DA (data assimilation) etc.
 - For a first implementation of ensemble, reduced the number of features
 - Ocean coupling, 6km resolution
- Requirements compute resources
 - O Estimated number of Intel Skylake-based nodes is 400+
 - Estimated resources on on-premises systems not available for 4-month continuous use
 - O Decided on a Cloud (AWS) implementation of 21-member ensemble
 - Four month capacity reservation using Amazon Reserved Instances



AWS Implementation

- AWS resources and services
 - Compute 168x hpc6a.48xlarge instances (96 cores/instance)
 - Storage (short term) 48TB, persistent Lustre file system (AWS FSx for Linux)
 - Storage (long term) AWS S3 bucket to store results
- Cloud configuration
 - o A single parallel cluster is created with
 - 168 Reserved Instances (hpc6a.48xlarge)
 - 42TB Parallel Lustre file system (can store output from 2 cycles)
 - Job scheduler SLURM
 - Workflow engine Rocoto
 - Input data from NODD (NOAA Open Data Dissemination) S3 bucket
 - A small file (SYNDAT) is uploaded from a NOAA on-prem system to Cloud every 15 minutes
- "Spin-up" of ensemble model on the Cloud
 - Phase 1: Porting from on-prem system
 - Over a 4-week period using On-Demand instances (prior to reserved capacity turned on)
 - Tested only up to 3-member ensemble model to keep costs low
 - Phase 2: tested 21-member ensemble RI-based cloud configuration 2 weeks



Running HAFS ensemble on Cloud

- Computing
 - Using Rocoto, 21 workflows (one per member) are run continuously
 - Each workflow creates multiple jobs submitted to the same SLURM job queue
 - Number instances per job varies 7 instances if it is forecast with ocean coupling
 - Entire workflow for a cycle completes under 6 hours
- Data movement per cycle
 - Majority of input data accessed from S3 bucket intra-cloud transfer
 - Output data 21TB per cycle
 - Output is stored initially in Lustre filesystem
 - When a new cycle starts output from previous cycle is transferred to S3 bucket
 - Limited set of output files transferred to a NOAA on-prem system
 - Generate graphics for NOAA and other organizations



Lessons Learned

- Accurate estimate of cloud resources
 - ensure defined cloud configuration meets project budget and runtime goals
 - use existing performance (runtime) info from on-prem runs
 - create self-contained benchmarks for critical steps
- Port from an on-prem system
 - similar cloud compute env. as on-prem system easy to debug
 - verification of results before actual cloud simulation work starts
- Leverage SMEs
 - critical during debug and result verification steps
- Leverage Parallel Works & CSP
 - is the could configuration optimal? CSPs keep adding new types of instances, services, etc.
 - develop a plan of action for resolving issues as they arise (appl. Level -> PWorks-> CSP)
- Monitor and address issues



Thank you!







Department of Commerce // National Oceanic and Atmospheric Administration // 26

















Highlights of HERC Products

RI events, Hurricanes Franklin and Lee, 2023

Hurricane Franklin 202300826 06Z

Hurricane Lee 20230907 12Z



HFMN Parallel: TC Probabilistic RI forecast

