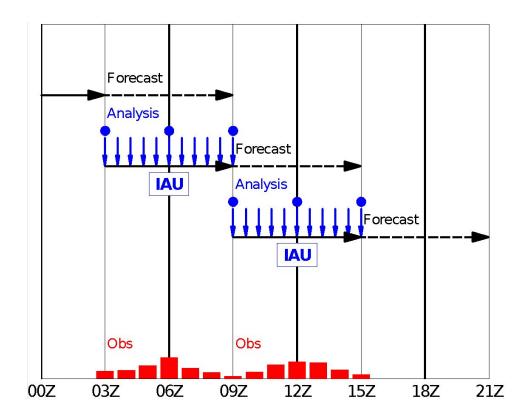
A "replay" reanalysis dataset for initializing coupled reforecast ensembles with the UFS

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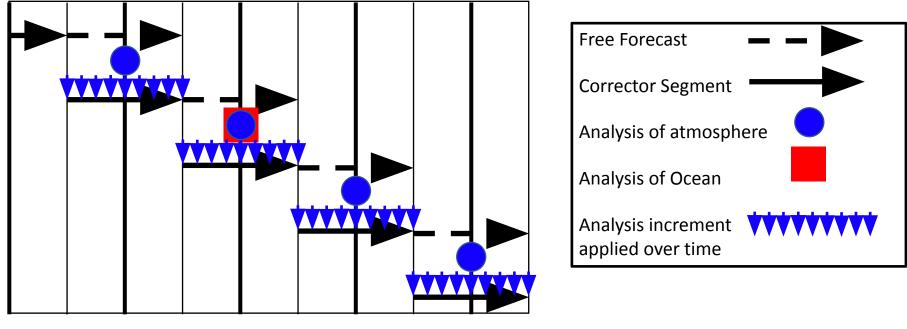
Motivation

- GEFSv13 will be coupled, need a coupled (atm/ocn/land/sea-ice) reanalysis to initial reforecasts.
- Don't have time or resources to run a coupled reanalysis (and then reforecasts) in time for implementation.
- As a stop-gap measure, we are using a technique called 'replay' to generate balanced coupled initial conditions for the UFS from existing reanalyses.
 - We do plan to run a coupled reanalysis to initialize SFSv1 reforecasts.

The data assimilation cycle

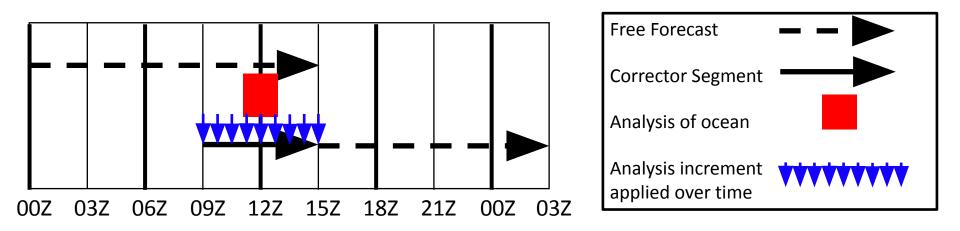


Coupled Replay Cycle



00Z 03Z 06Z 09Z 12Z 15Z 18Z 21Z 00Z 03Z

Ocean Replay Cycle



We could use a full 24-hour forecast and corrector segment, but this way is consistent with the coupled replay.

Advantages of replay

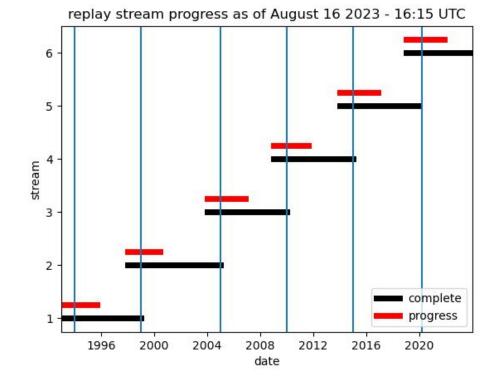
- Can easily re-run the reanalysis without needing the data assimilation if new output needed (different frequency, missing variables) or if model is upgraded but there is not yet a consistent reanalysis.
 - Envision reanalysis production on a 7 year cycle, model upgrades more like every 2 years.
- Can add additional components to modeling system and re-run (such as aerosols: <u>Orbe et al. 2017</u>).
- Use mean replay increments to assess model errors, train ML-based bias correction schemes: (<u>Chen et al 2022</u>).
- Constrain certain portions of the system, i.e. only replay tropics: Dias et al. 2021
- Avoids 'spin-up/spin-down' problems seen in 'cold start' forecasts by gradually adding in increments. Unconstrained components (e.g. land) can adjust consistently.

Implementation details

- 1994-2023, full coupled restarts once per day (at 03Z, informed by analyses up to including 00Z).
- Publicly available on AWS (no egress charges)
 - <u>https://noaa-ufs-gefsv13replay-pds.s3.amazonaws.com/index.html</u>
- Land unconstrained
 - Except for snow, which is updated using JEDI OI-based DA.
- ORAS5 sea-ice inserted.
- Ensemble perturbations for reforecasts from ERA5 analysis perturbations in atmosphere, and lagged differences between four reanalyses in ocean (no initial perturbations for land and sea-ice).
- Being run in 6 overlapping streams, with first year discarded from each stream.
 - Streams initialized from ORAS5/ERA5 and NOAH-MP forced by ERA5.

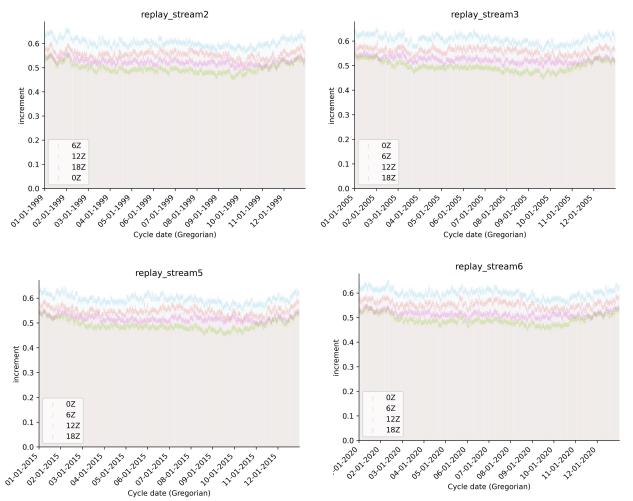
Replay progress

- All streams in at least 2nd year of production.
- Estimated completion Oct 1.



QC/Monitoring

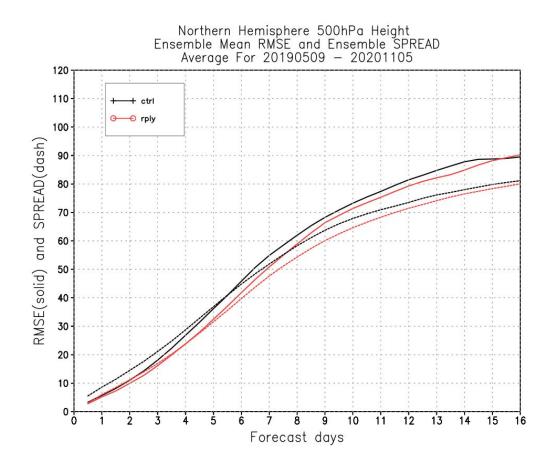
- Plots routinely generated to monitor quality, detect problems.
- Global mean atm T increment RMS show here.



Replay-initialized ensemble reforecast evaluation (courtesy Bing Fu/EMC)

- Test period that spans May 2019 through November 2020
- Reference forecasts are our best proxy for operational initial conditions for GEFSv13 which consists of experimental JEDI/SOCA ocean and ice analyses, GFSv16 atmospheric analysis and spun-up Noah-MP land initial conditions. Labeled ctrl
- Replay ensemble forecasts are initialized with restarts at the end of the IAU window (FHR03) and perturbations are applied to the atmosphere and ocean initial conditions. Labeled rply
- 40 cases of 6-member ensemble forecasts out 16-days

500 hPa spread/error



Ensemble forecasts with replay initial conditions are substantially better than the control experiments with about 12-hours of increased skill at day-5.

Ensemble spread is also lower in replay and matches the RMS error quite well for the first 6 days.

Summary

- A 30 year daily dataset (1994-2023) of coupled UFS restarts (for ATM, OCN, land, sea-ice and waves) at ¼ degree will be publicly available on AWS this fall.
 - Informed by ECMWF reanalysis of atmosphere, ocean and sea-ice, plus snow observations.
 - Includes perturbations for atmosphere and ocean for generating ensembles.
- Will be used to initialize reforecasts for GEFSv13.
- Also useful for research, model development.
- Increments should be useful for identifying regime-dependent biases
 - Ongoing work at PSL to use these increments to train an in-line ML-based bias correction.