# Neuralio A.I.

# **How WRF-Data Assimilation benefits from HPC**

EuroCC@Greece HPC Training Series - Course 13

Paraskevi Vourlioti







Our company at a glance Our HPC Journey WRF-DA and HPC >WHY WE NEED DATA ASSMILATION AND WHY HPC IS VITAL >WHO IS USING DA? > WRF-WRFDA We have compiled- now what do we expect? Challenges and tips Future of WRF and HPC - AI challenge





# Our company at a glance

#### Year of **Establishment**

Established in 2022. Thessaloniki, Greece Estonian Branch in 2023

Personnel

8 persons, gender balance 50/50

#### **Expertise**

AI/ML, Atmospheric Modeling, Remote Sensing, HPC

#### **Projects**

Public Funded projects: 3 closed/ 2 ongoing Private Projects: 2 Closed/ 3 Ongoing

#### Services

Weather Intelligence Enviromental Foorprint Land Evaluation R&D as a Service











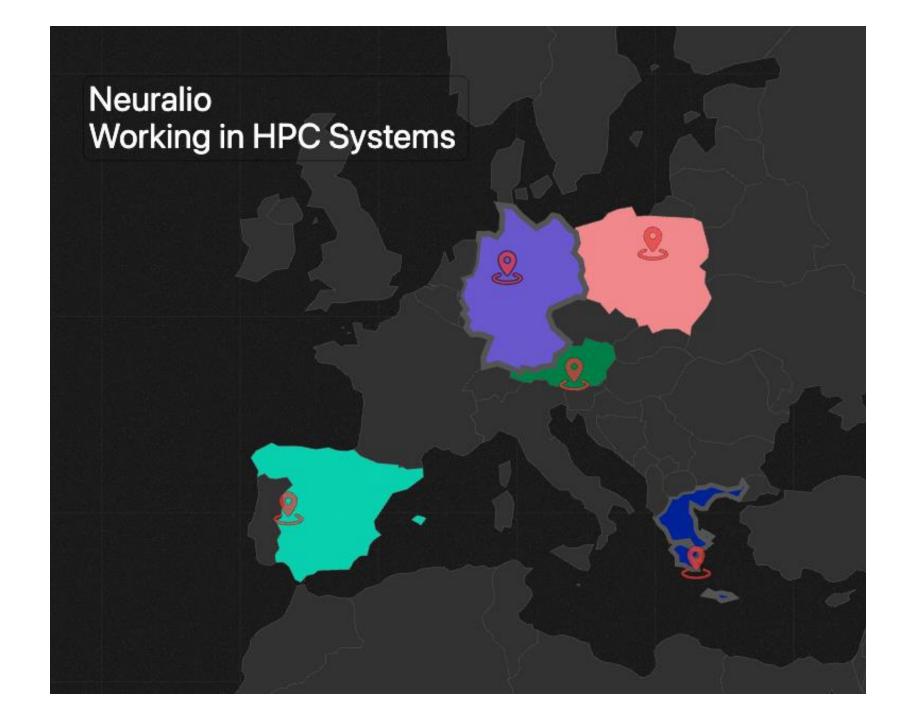


# **OUR HPC JOURNEY**



Many thanks to EuroCC network for providing resources also to SMEs.



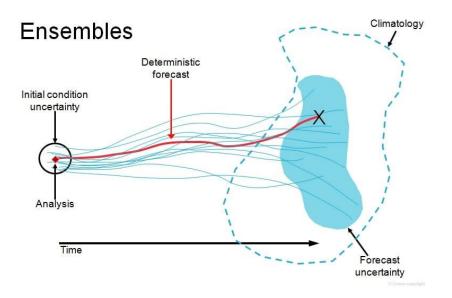


# **WRF-DA and HPC**



Why Data Assimilation: The atmosphere is a chaotic system, where the accuracy of short and medium-range forecasts depends critically on the precision of the initial state.

**What is Data Assimilation:** Our best effort to combine mathematically our model with observations to get the closest to the true atmospheric state



Source: https://opensnow.com/news/post/ensemble-vsdeterministic-weather-models Computational power to produce:

- ✓ High Spatial Resolution forecasts
- ✓ Data assimilation to improve our starting point.
- ✓ Fast production

→ Example in CESGA HPC: 3.5 hours on 336 cores for a 2km spatial resolution forecast and 24 hours forecast horizon

-> Example in CESGA HPC: also 3.5 hours to assimilate (with 4DVAR)!!!

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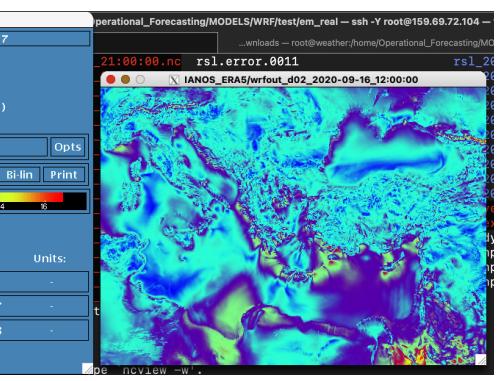
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#### In-house prediction with WRF model and intial conditions from GFS system

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# An example of how different intial conditions can lead to differen forecasts

In-house simulation with other intial conditions for the intialisation



# WHY WE NEED DATA ASSMILATION AND WHY HPC IS VITAL? ..... WHO IS USING DA?

Center	Operational DA?	Approx. daily obs assimilated (millions)	Main data sources assimilated	Main DA system type
ECMWF (European Centre)	Yes	~60 (2024)	~90% satellite measurements (radiance sounders, scatterometer winds, GPS-RO, etc.), plus all available in-situ data (surface stations, radiosondes, aircraft reports) <u>ecmwf.int</u> .	4D-Var (four-dimensional variational) with an ensemble of data assimilations for hybrid background error covariances <u>ecmwf.int</u> .
NOAA/NCEP (U.S. GFS)	Yes	~34 (2024)	Vast majority satellite data (e.g. radiances from polar orbiters, geostationary AMVs, GNSS-RO) and conventional observations (radiosondes, dropsondes, buoys, METAR surface reports, aircraft AMDAR/ACARS)repository .library.noaa.gov.	Hybrid 4D EnVar (four- dimensional ensemble- variational assimilation using 80-member EnKF for background errors) <u>nco.ncep.noaa.gov</u>
JMA (Japan Meteorological Agency)	Yes	~10–15 (est.)	Satellite radiances (NOAA/MetOp ATOVS, Aqua AMSU-A, Himawari AMVs, etc.), ground- based observations (SYNOP/ship/buoy surface reports, radiosondes/pilots), aircraft data (AIRED)	4D-Var (incremental variational) – JMA's global analysis employs 4D-Var (now with some hybrid ensemble contribution in recent upgrades)jma.go.jp.

- When you download WRF, before you compile it, make a copy of the source code and rename it WRFDA-vX.X.X. ---- EXECUTABLE: da\_wrfvar.exe
- Use the same libraries, compiler you used for WRF.
- Depending on the DA methodology you will need different compilation strategy.
- Simpler DA that can run without HPC: 3DVAR -- straightforward compilation
- For WRFDA 4DVAR, it is necessary to have WRFPLUS installed first – HPC needed and a form of this methodology is used by Weather forecasting Centres
- Start with the test data available in the tutorial starting with a simple 3DVAR



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- User guide --- tutorial --- free input data for DA <u>https://www2.mmm.ucar.edu/wrf/users/wrfda/</u>
- Pain point: If you have your own data to include in the DA, you have to prepare the little-R file on your own: <u>https://www2.mmm.ucar.edu/wrf/users/wrfda/OnlineTutorial/Help/littler.html</u>

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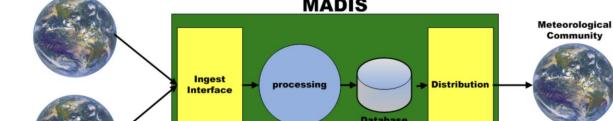
Use only data that have passed quality control and you trust. Quality-controlled weather stations:

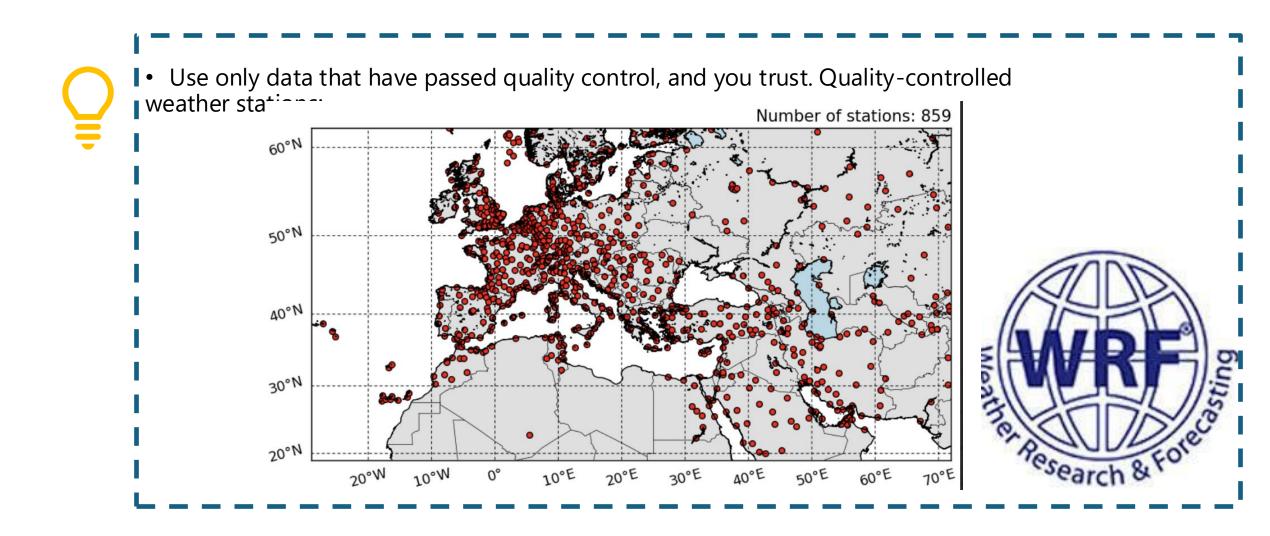
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Meteorological Assimilation Data Ingest System MADIS is a meteorological observational database and data delivery system that provides observations that cover the globe.





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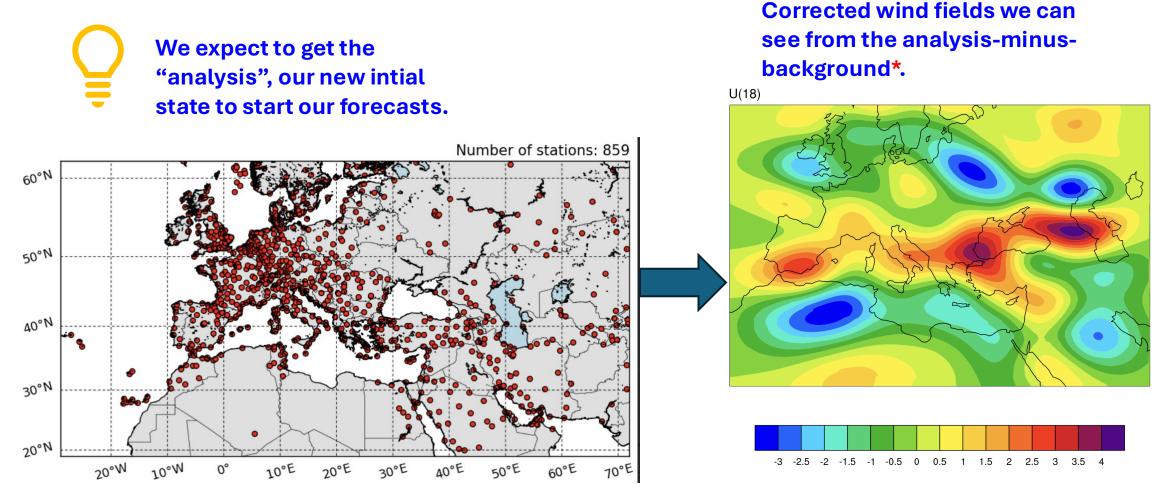
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Use only data that have passed quality control, and you trust. Quality-controlled weather stations.

 Tools to convert the data for MADIS to little-R: <u>https://www2.mmm.ucar.edu/wrf/users/wrfda/download/madis.html</u>

## WHY WE NEED DATA ASSMILATION AND WHY HPC IS VITAL? WRF-WRFDA We have compiled- now what do we expect?



https://github.com/wrf-model/WRFDA\_TOOLS

\* backround= intial field estimation

# Challenges and tips

Containerisation in the HPC



- HPC stuff can pull docker images from Docker Hub registry and convert them into singularity.
- What does this mean?
- The user can just "load" the module from command line.
- Ask always the technical team if they can do this for you --- Especially if you have an oficial image (e.g MET)





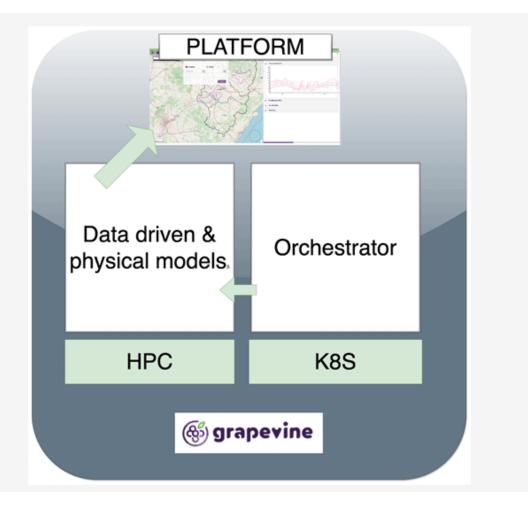
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singularity build <output\_image.sif> docker://<docker\_image>

# Challenges and tips

#### Operational set-ups with HPCs



- 1. A need to orchestrate the job submisions to the HPC.
- 2. Orchestrators like cloudify can do this by connecting the cloud (k8s) with the HPC through croupier plugin.
- 3. The job-submisions are linked to a user with a certain amount of credits in the HPC.
- 4. The job can be submited but it cannot start if the user has not reserved the computation time in the HPC.
  - Automated reservations/configurations for oprational set-ups?
  - Should large and small jobs grouped under one reservation?  $-\rightarrow$  Not optimal for the HPC usage as we reserve the max amount of cores but for some time these resources remain idle.

### Future of WRF and HPC – AI challenge

**Key Examples:** 

- **GraphCast (DeepMind):** Outperforms ECMWF HRES for 10-day forecasts, near real-time.
- FourCastNet (NVIDIA/MIT): High-res (0.25°) forecasts in <2 seconds.
- **Pangu-Weather (Huawei):** Transformer-based, cheaper and faster than classical NWP.
- **ECMWF AI4Weather:** Integrating AI into operational workflows (hybrid future).

### Future of WRF and HPC – AI challenge

#### **Challenges:**

- AI struggles to fully replicate complex physics (e.g., convection, microphysics).
- Data assimilation workflows (4D-Var, EnKF) hard to integrate with pure AI.
- Al generalization and reliability for extreme events still uncertain.
- Training requires massive datasets and HPC resources.
- Trust and interpretability issues with black-box AI models.

### Future of WRF and HPC – AI challenge

#### **Opportunities:**

- Al surrogate models can accelerate expensive WRF physics (e.g., radiation).
- Al improves post-processing: bias correction, downscaling, uncertainty quantification.
- Hybrid WRF + AI systems: physical consistency + AI speedup.
- Reduced computational cost  $\rightarrow$  high-res forecasts and large ensembles feasible.

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# Thank you!!

#### AI will not replace WRF overnight — but it will change how we forecast. HPC remains vital to support both next-gen NWP and AI models.

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**Contact Us** 

