

WGNE – evolution of systematic errors in Earth system modelling

Nils Wedi

co-chair WGNE

European Centre for Medium Range Weather Forecasts (ECMWF)





**38 years of
WGNE in
WCRP & CAS !!
Earlier structure
since 1967 ...**

Working Group on Numerical Experimentation (WGNE)

WGNE co-chairs: Nils Wedi (ECMWF), Ariane Frassoni (INPE) WCRP Secretariat: Nico Caltabiano

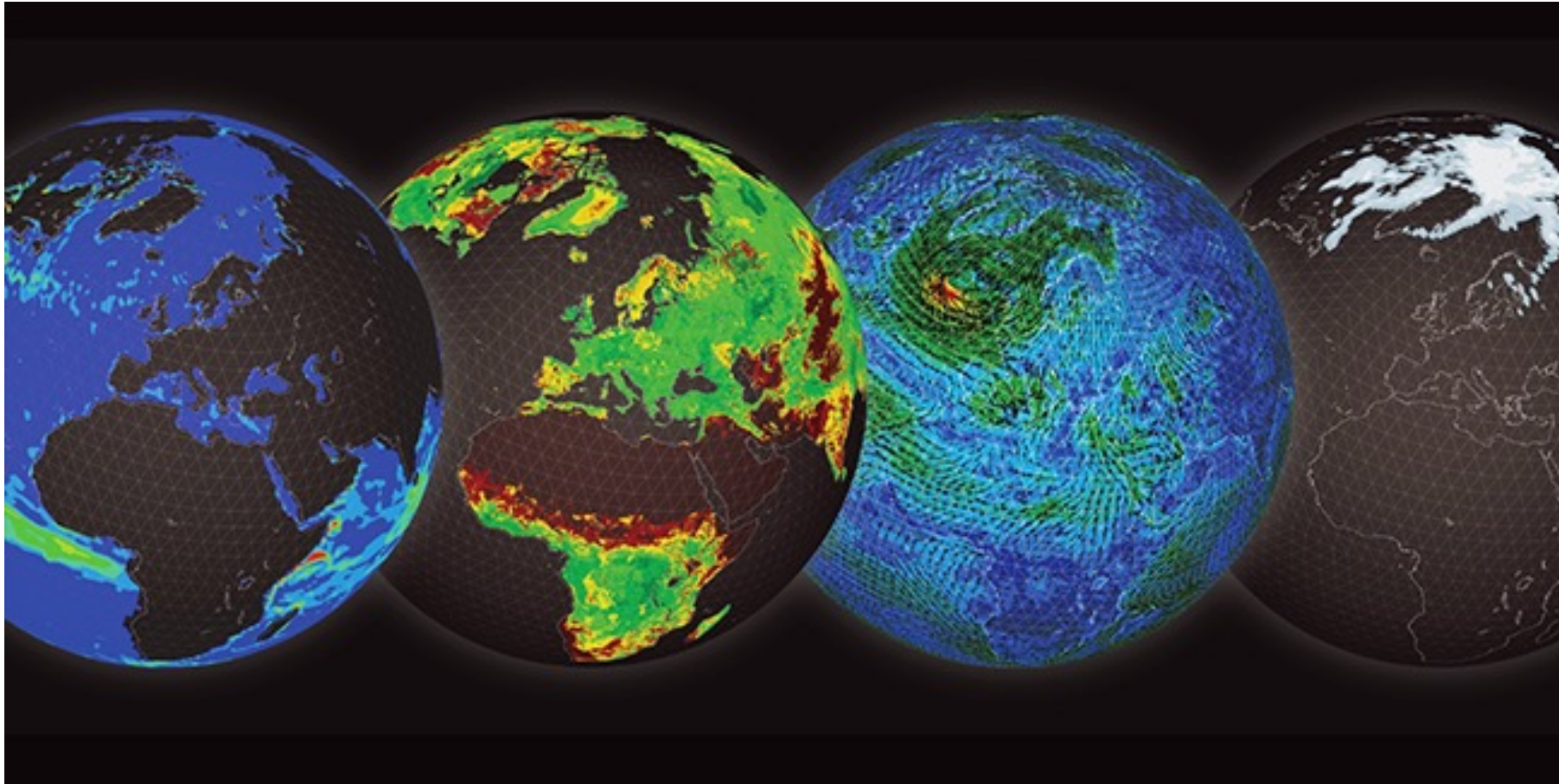
Members

- *Tim Graham-Met Office (UK)*
- *Romain Roehrig-CNRM/MeteoFrance (France)*
- *Gunther Zaengl-DWD (Germany)*
- *Peter Lauritzen-NCAR (USA)*
- *Fanglin Yang-NOAA/NCEP/EMC (USA)*
- *Masashi Ujiie-JMA (Japan),*
- *Ron McTaggart-Cowan-ECCC (Canada),*
- *Oscar Alves-BOM (Australia),*
- *Jian Sun-CMA (China),*
- *Elena Astakhova-Rushydromet (Russia) still supporting wgne.net*

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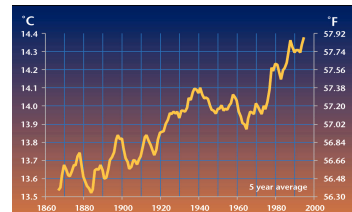
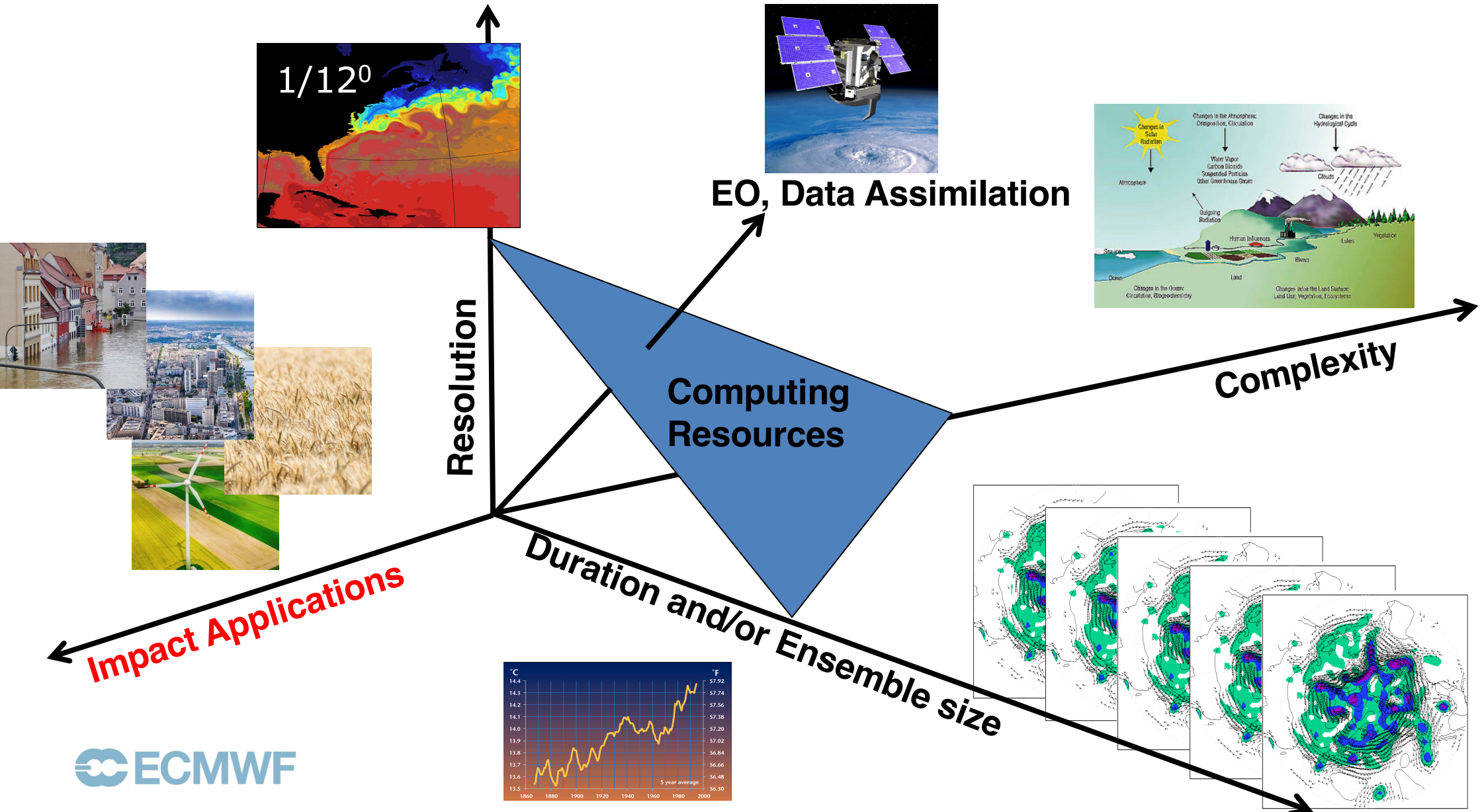
1. Earth system modelling
2. Review from 6th WGNE workshop
3. Weather parameters and high-impact weather indicators & EW4All
4. Rise of data driven forecasting
5. Interactivity & novel data delivery

Multi-spheres concept in modelling & prediction



<https://www.ecmwf.int/en/about/media-centre/news/2021/ecmwf-strategy-2021-30-and-machine-learning-roadmap-launched>

A NEW AXIS CONSTRAINING SCALABILITY, BIG DATA HANDLING AND COMPUTING RESOURCES



OpenIFS

EC-Earth

Integrated Forecasting System (IFS)

ECLand

Multiscale parameters regionalisation (MPR)

Seasonal Vegetation cover

Prognostic LAI

Photosynthesis

Soil carbon

Fire module

Dry deposition

....

Snow Multi Layer (5ML)

Soil 4Layers (4L ~3m)

A-gs (Carbon)

Jarvis (Energy)

Soil moisture stress + rooting depth

Lake

CaMa-Flood hydrodynamic model

Climate fields: soil, vegetation, ...

Vertical resolution

Soil N Layers (10L~8m)

Snow N Layers (MLx)

Urban module

CH4 module



Working Group on Numerical Experimentation (WGNE)

6th WGNE workshop on **systematic errors in weather and climate models** at ECMWF in 2022 (<https://events.ecmwf.int/event/241>)

Systematic Errors in Weather and Climate Models: Challenges and Opportunities in Complex Coupled Modeling Systems, *Frassoni, Reynolds, Wedi et al 2023*

<https://doi.org/10.1175/BAMS-D-23-0102.1>



Systematic Errors in Weather and Climate Models

Selected qualitative conclusions for the 2024-2027 time-scale:

Constraining errors on troposphere-stratosphere coupling and improved predictability

Amplitude of diurnal cycle of precipitation over land remains a challenge

Reduction in systematic errors of upper ocean (SST, salinity, Gulf stream separation) and of some deep ocean properties

Substantial errors in high-latitudes remain

Substantial MJO simulation errors (and convective boundary layers in coupled models) remain

Substantially improved tropical cyclone track and intensity forecasts, in part through improved air-sea coupling

Improved hydrological and flood prediction and improved representation of vegetation and soil, and snow, in part based on more up-to-date mapping information

Increased complexity of very-high resolution simulations within coupled ocean-atmosphere-land systems give also rise to new systematic errors...

Bias correction of systematic errors through ML/AI advances

Recommendations to advance on systematic error reduction including data assimilation, machine learning, and a hierarchy of models supported by standardised and widely available observational data.



Forecast Errors in Weather and Climate Models 2013 → 2023

Met Office

Area	Parameter	Forecast range								
		T+24 RMSE	T+24 10-year RMSE change	% difference	T+72	T+72 10-year RMSE change	% difference	T+120	T+120 10-year RMSE change	% difference
NH	pmsl	82,4417	67,0675	-22,9	211,6148	173,4242	-22,0	398,8079	354,9571	-12,4
	500 hPa GPH	6,5992	5,244	-25,8	20,7134	17,0331	-21,6	41,8865	37,211	-12,6
	250 hPa wind	3,3205	2,8675	-15,8	7,5854	6,7353	-12,6	12,5156	11,6188	-7,7
	250 hPa temp	0,6437	0,5743	-12,1	1,4889	1,3209	-12,7	2,3724	2,1715	-9,3
TR	850 hPa wind	1,6936	1,5059	-12,5	2,643	2,4518	-7,8	3,2955	3,1333	-5,2
	250 hPa wind	3,0121	3,1105	3,2	5,3702	5,5818	3,8	7,0235	7,3234	4,1
	250 hPa temp	0,3827	0,4454	14,1	0,6646	0,7205	7,8	0,8236	0,8538	3,5
SH	pmsl	96,6029	69,2929	-39,4	266,07	206,771	-28,7	507,1582	435,204	-16,5
	500 hPa GPH	8,343	6,1211	-36,3	26,2052	20,1679	-29,9	51,8982	44,0892	-17,7
	250 hPa wind	3,3525	2,952	-13,6	8,1274	6,9988	-16,1	13,6093	12,2837	-10,8
	250 hPa temp	0,665	0,5999	-10,9	1,5812	1,3605	-16,2	2,5117	2,2335	-12,5

Tim Graham & colleagues

Based on a range of global models



High Impact weather error reduction

Thomas Haiden

Forecast Errors in Weather and Climate Models 2013 → 2023

Feature	Current error or score (2023)	Error or score 10 years ago (2013)	Approximate improvement in 10 years	Comments
Tropical cyclone position	MAE (D+3) = 160 km	MAE (D+3) = 180 km	11%	
	MAE (D+5) = 250 km	MAE (D+5) = 350 km	29%	
Tropical cyclone intensity (central pressure)	MAE (D+3) = 11 hPa	MAE (D+3) = 15 hPa	27%	
Strong wind	ROCS (D+5) = 0.77	ROCS (D+5) = 0.72	6%	EFI (95th percentile) ROC skill in Europe
Significant wave height	SI (D+3) = 20%	SI (D+3) = 23%	13%	SI = Scatter Index (error standard deviation divided by obs) in %
	SI (D+5) = 30%	SI (D+5) = 33%	9%	
High temperatures	ROCS (D+5) = 0.92	ROCS (D+5) = 0.88	5%	EFI (95th percentile) ROC skill in Europe
Heavy rainfall	ROCS (D+5) = 0.68	ROCS (D+5) = 0.63	8%	EFI (95th percentile) ROC skill in Europe
Heavy rainfall	ETS (D+3) = 0.155	ETS (D+3) = 0.125	24%	Equitable Threat Score (ETS) for >50mm/24h in N. Extratropics
	ETS (D+5) = 0.100	ETS (D+5) = 0.075	30%	

EW4All Priority hazards:

- Flash-floods
- Drought/Dry spell
- Riverine Floods
- Tropical cyclone
- Thunderstorms/Squall lines
- Heat wave

The landscape ...

<https://community.wmo.int/en/news/exploring-possibilities-artificial-intelligence-areas-water-weather-and-climate>

DeepMind & Google's ML-Based GraphCast Outperforms the World's Best Medium-Range Weather Forecasting System



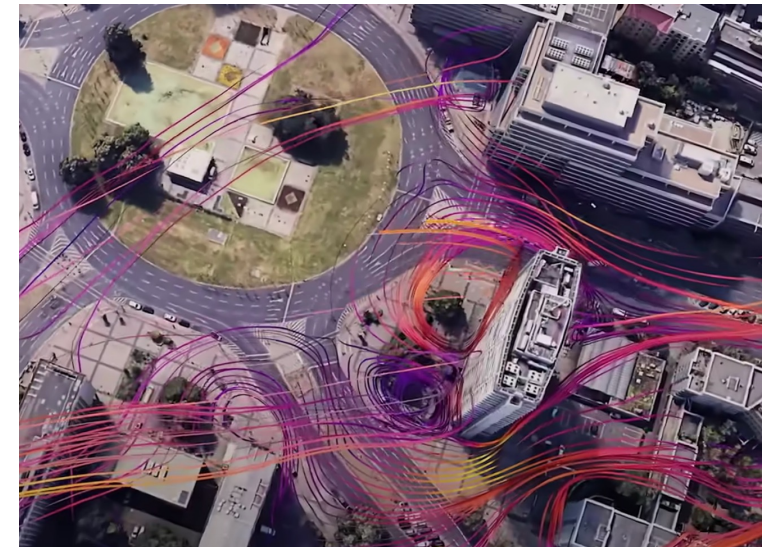
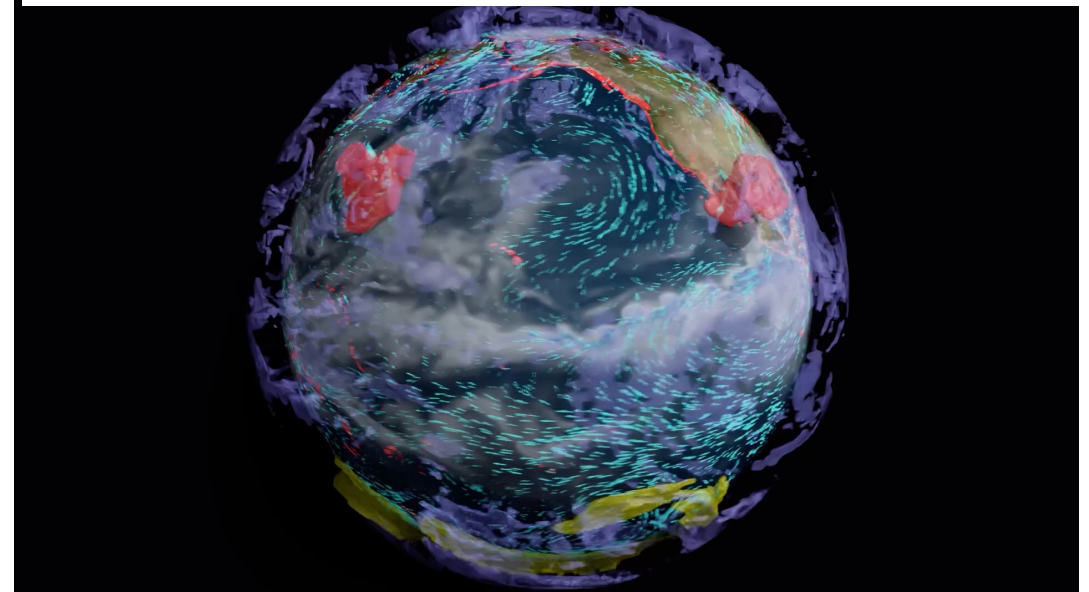
<https://www.science.org/doi/10.1126/science.adi2336>



Natural language translation

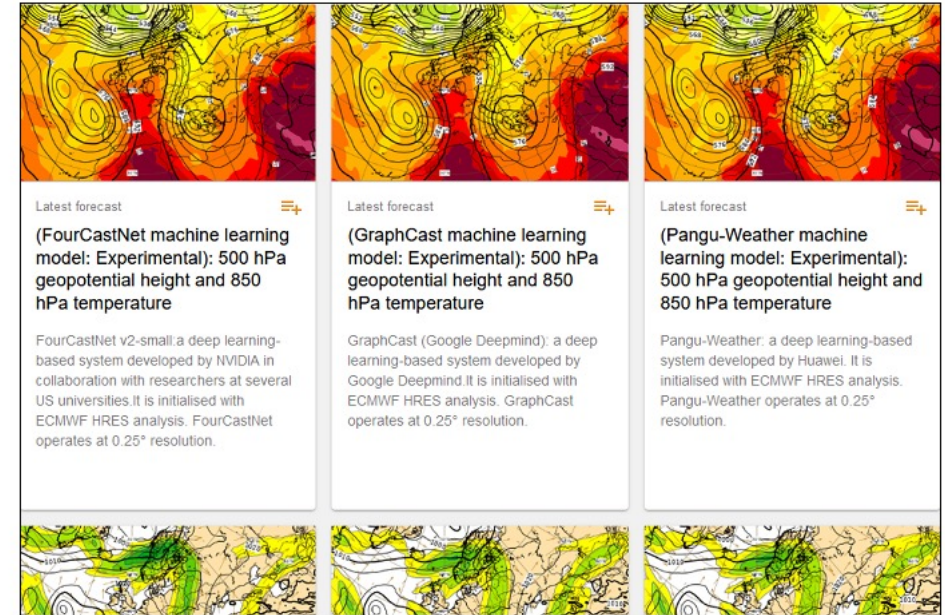
NVIDIA to Build Earth-2 Supercomputer to See Our Future

November 12, 2021 by JENSEN HUANG



The rise of data-driven forecasting in 2023

charts.ecmwf.int



THE RISE OF DATA-DRIVEN WEATHER FORECASTING

A FIRST STATISTICAL ASSESSMENT OF MACHINE LEARNING-BASED WEATHER FORECASTS
IN AN OPERATIONAL-LIKE CONTEXT

A PREPRINT

Zied Ben Bouallègue, Mariana C A Clare, Linus Magnusson, Estibaliz Gascón, Michael Maier-Gerber, Martin Janoušek, Mark Rodwell, Florian Pinault, Jesper S Dramsch, Baudouin Raoult, Florence Rabier, Matthieu Chevallier, Irina Sandu, Peter Dueben, Matthew Chantry, Florian Pappenberger

ECMWF

ABSTRACT

Data-driven modeling based on machine learning (ML) is showing enormous potential for weather forecasting. Rapid progress has been made with impressive results for some applications. The uptake of ML methods could be a game-changer for the incremental progress in traditional numerical weather

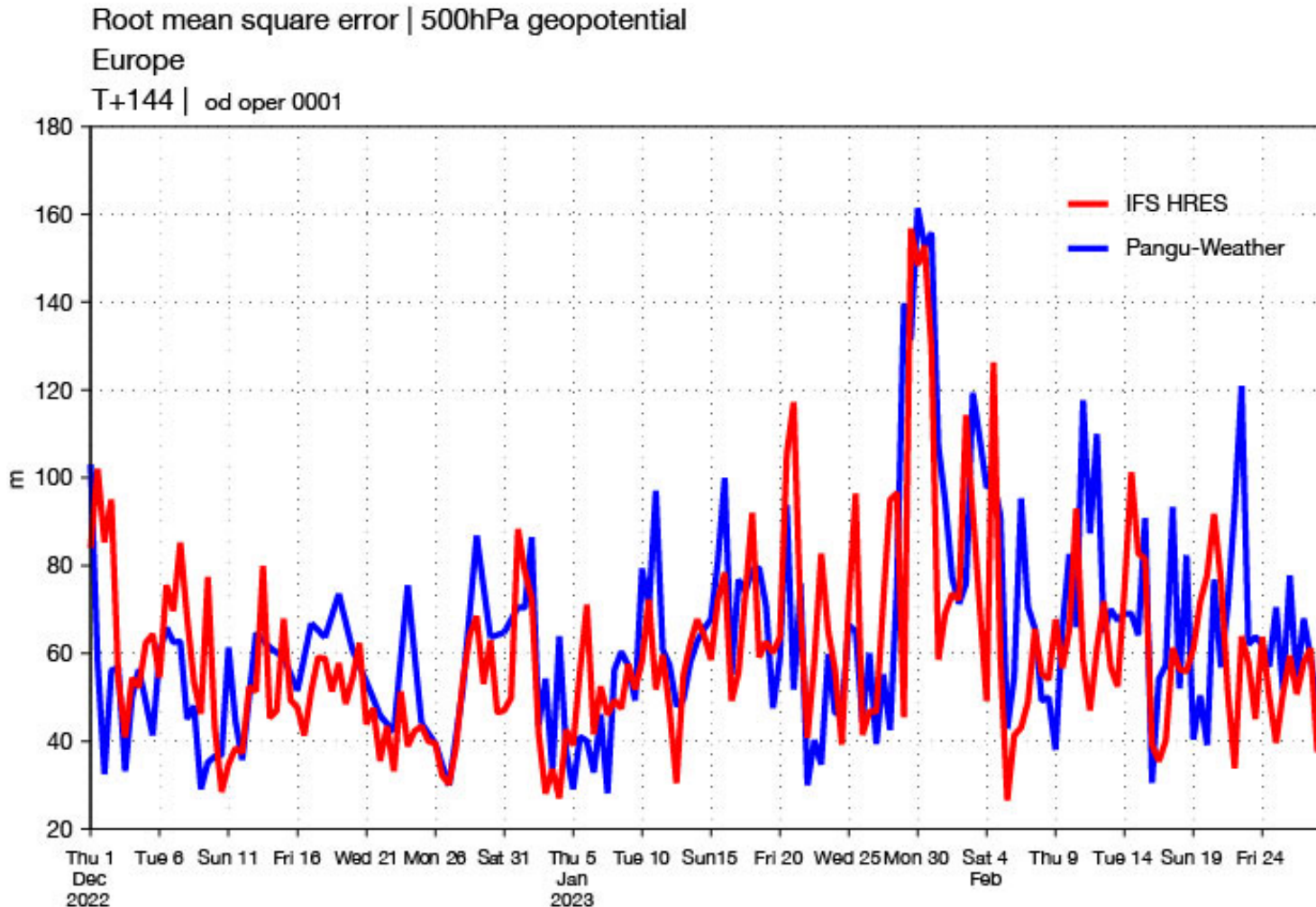
ECMWF unveils alpha version of new ML model

13 October 2023
The AIFS team

<https://www.ecmwf.int/en/about/media-centre/aifs-blog>

What results are showing

Time-series of RMSE over Europe, day 6



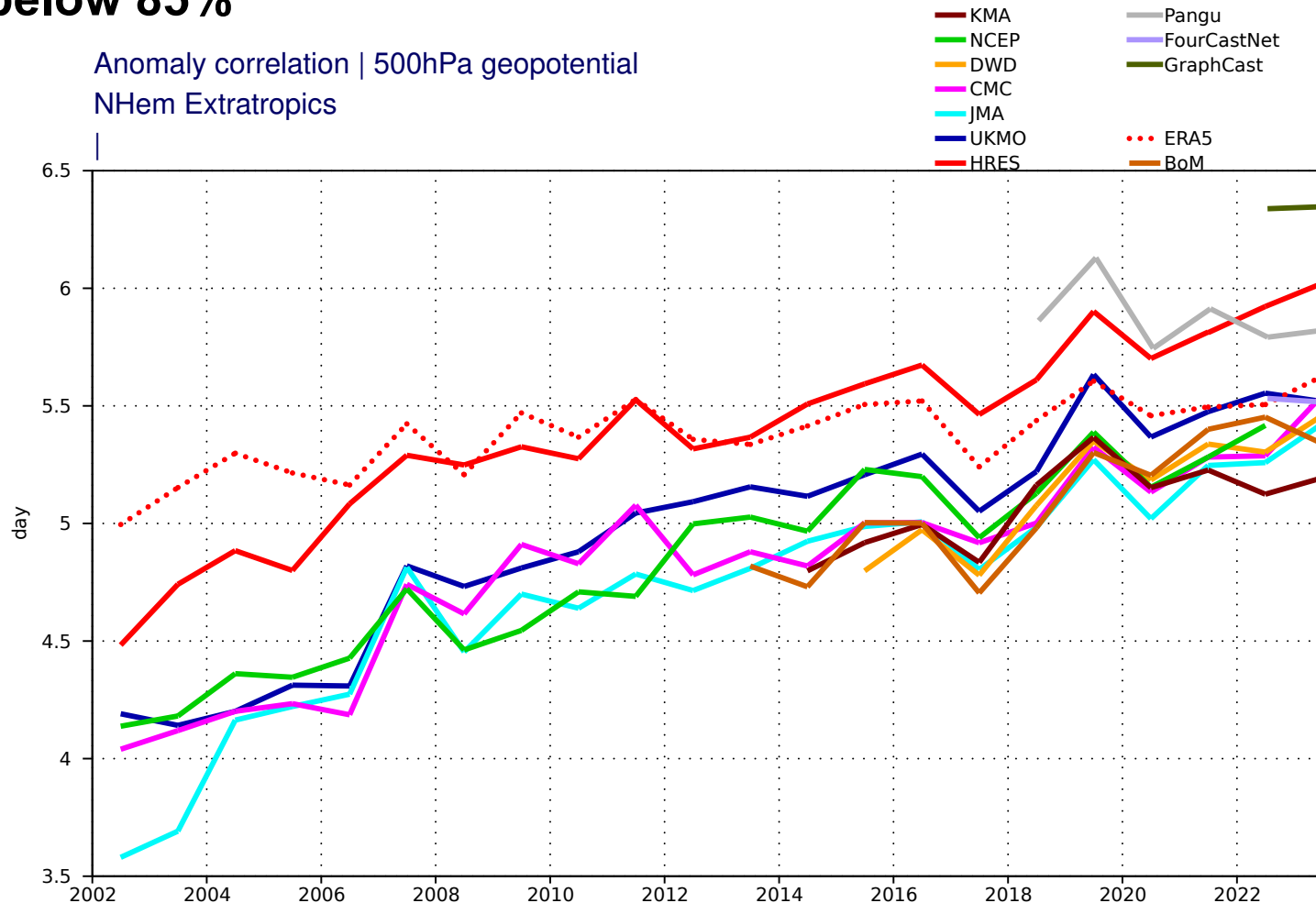
PanguWeather (initialised with ECMWF operational analysis) vs ECMWF HRES forecasts

- Results extremely close...
- 'forecast busts' at similar time
 - more 'physical' than one might think

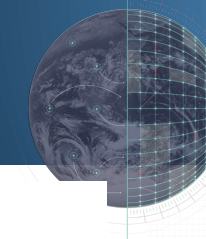
What results are showing

Headline score - 500hPa geopotential

Anomaly correlation of 500hPa geopotential over Northern Hemisphere Extratropics, falling below 85%



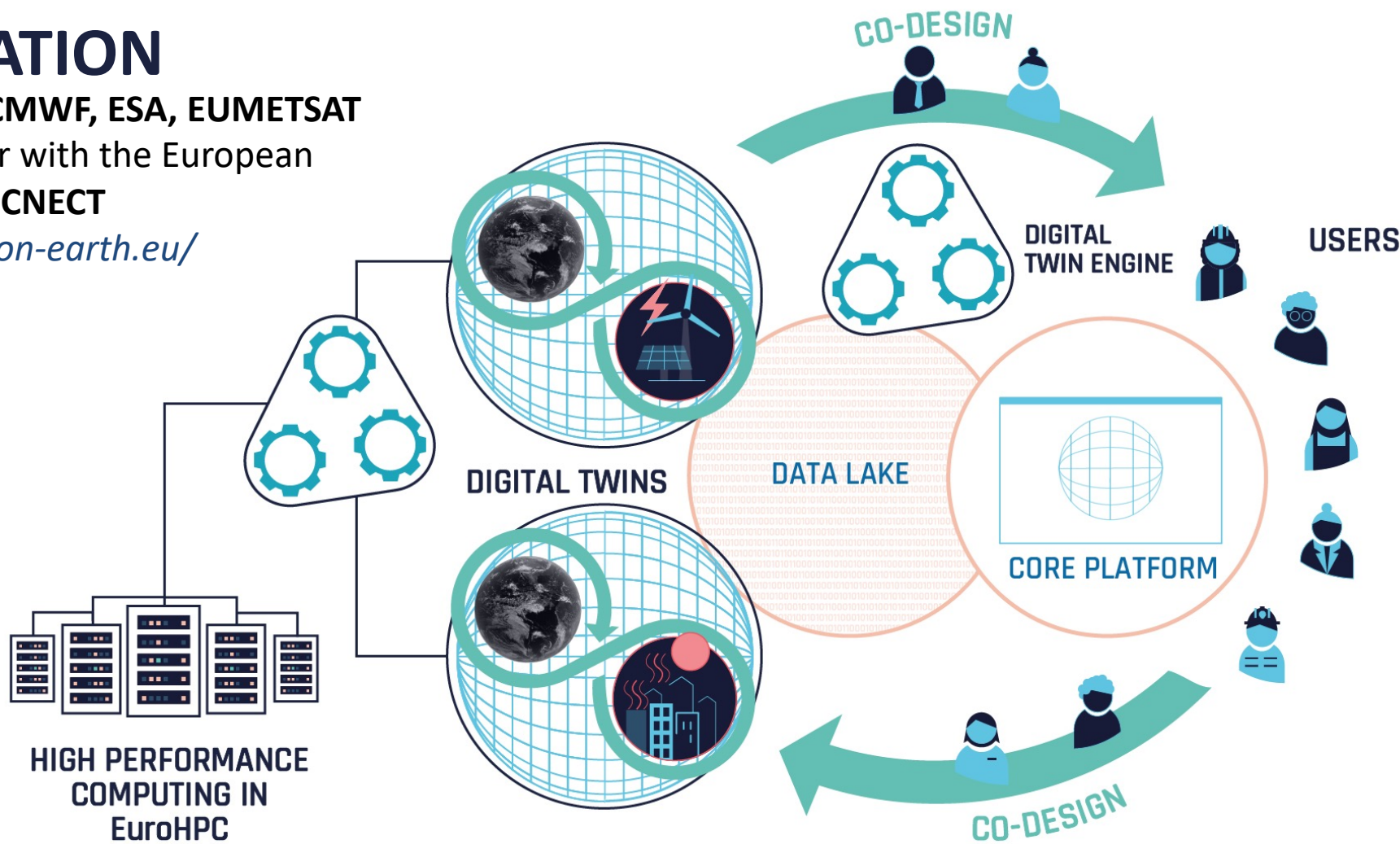
ML models (initialised with ECMWF HRES analysis)



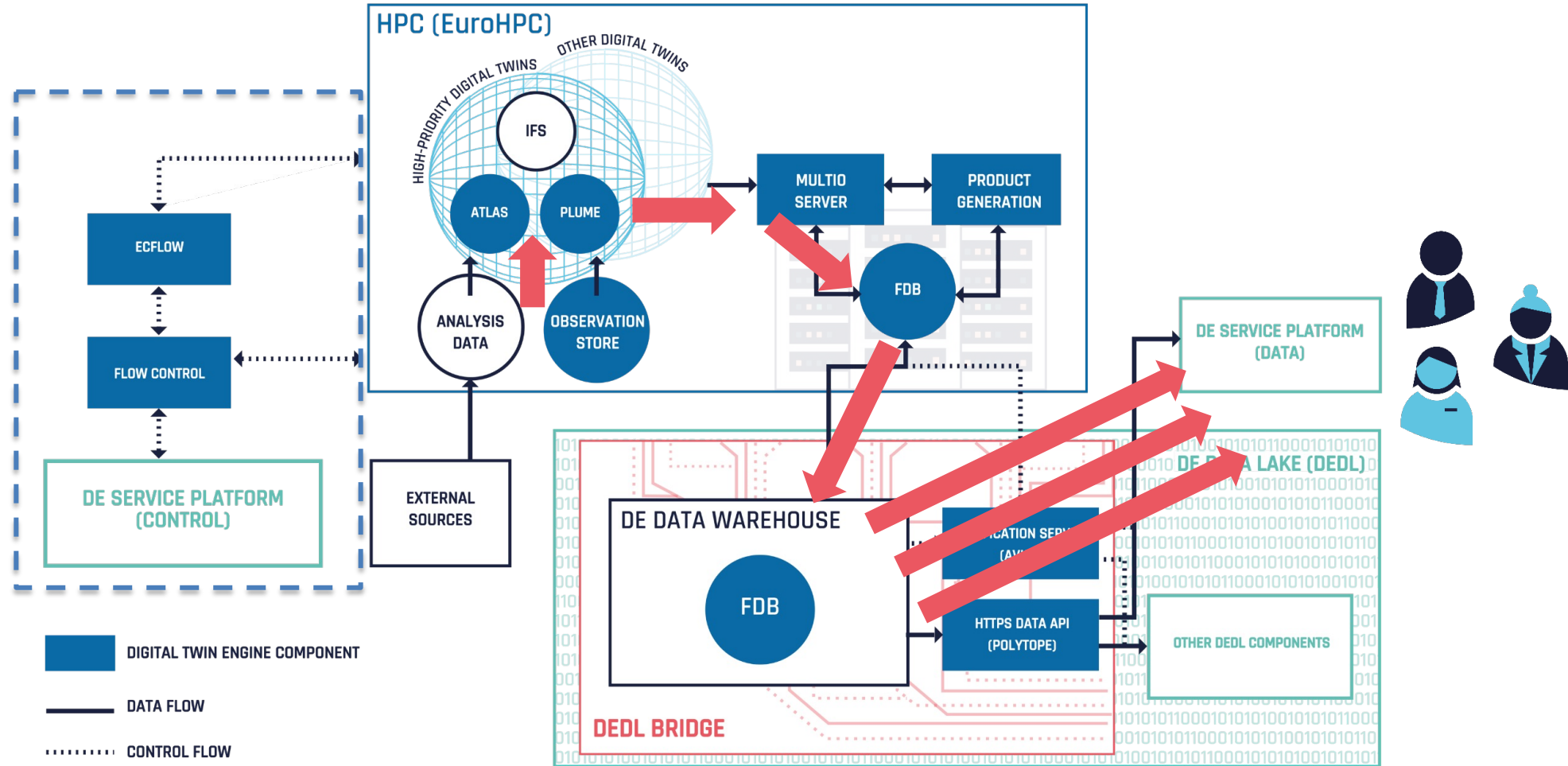
INNOVATION

Three entities **ECMWF, ESA, EUMETSAT** working together with the European Commission **DG-CNECT**

<https://destination-earth.eu/>



Digital Twins, managing Big Data & interacting with bespoke information resources



e.g. WIS2.0 compatible data access

<https://pygeoapi.io/>

<https://polytope-client.readthedocs.io/en/latest/>

Systematic Errors in Weather and Climate Models

Conclusions for 2024-2027 timeframe

1. **Significant improvements over last decade** on high-impact weather and multi-model hemispheric scores
2. Identified hazards **benefit from progress in land-surface modelling, higher horizontal resolution and integration with hydrological modelling** within land-surface schemes
3. **ML/AI can improve systematic errors, timeliness of delivery and uncertainty estimation**
4. Challenges WMCs readiness to make available **open Earth-system data and accelerate novel access patterns (including compute)**

EW4All Priority hazards:

- Flash-floods
- Drought/Dry spell
- Riverine Floods
- Tropical cyclone
- Thunderstorms/Squall lines
- Heat wave

Thank you



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