

AI for Weather Task Team

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Context

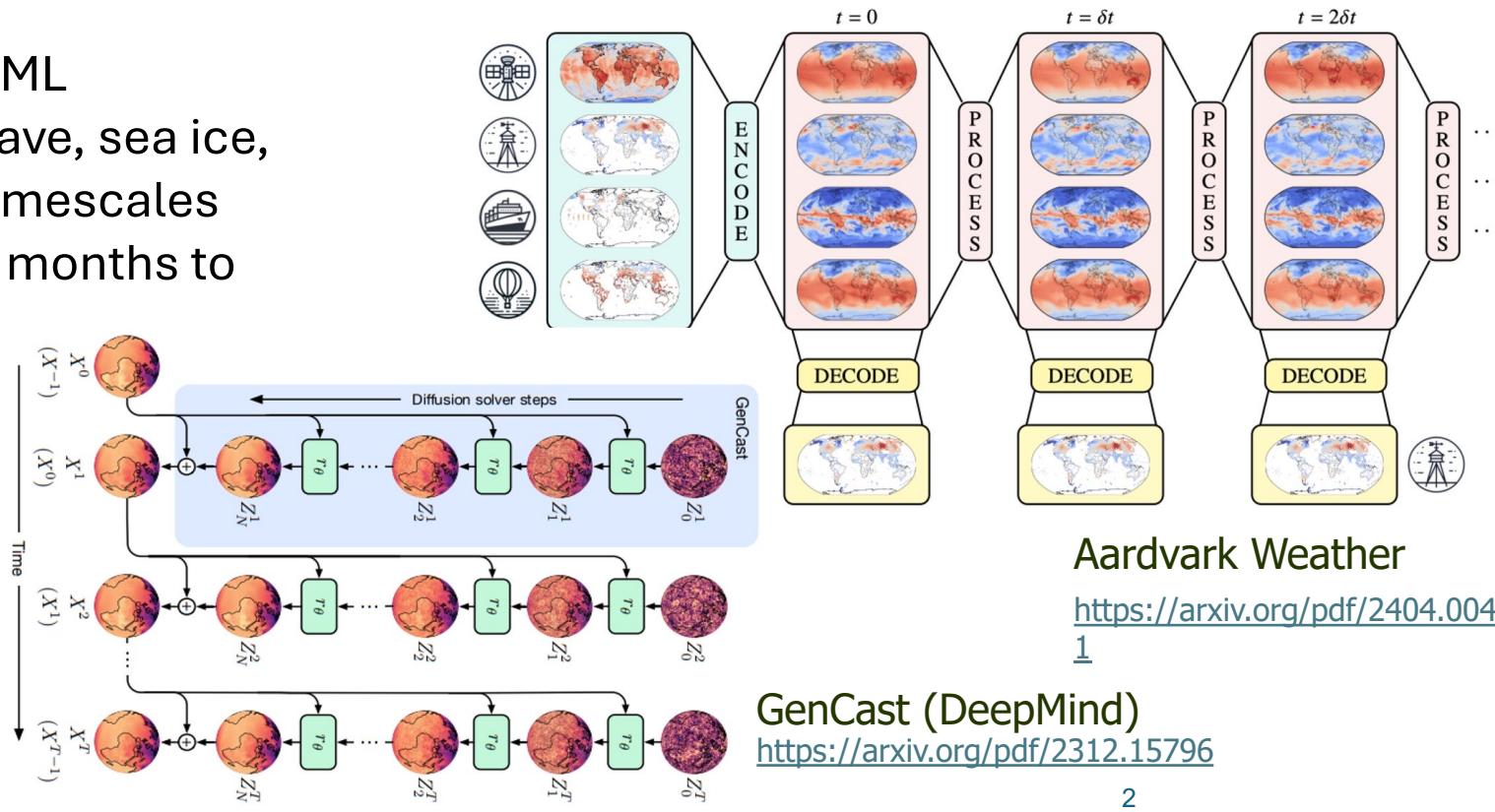
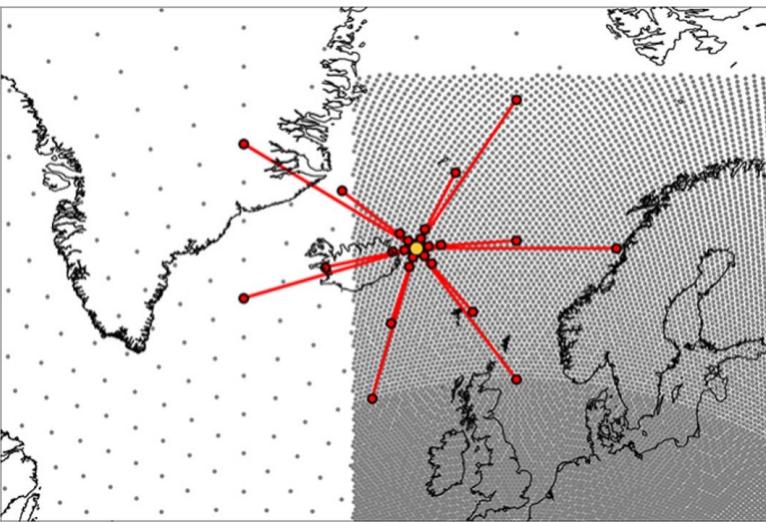
Background

Pre-2022: Two ML atmosphere models in the literature, neither with performance competitive with traditional NWP models

Post 2022 to today: more than 40 AI/ML atmosphere, ocean, land surface, wave, sea ice, coupled, foundation, etc. models. Timescales from DA to hours to weeks (and now months to years...)

From the ECMWF AIFS Blog:

<https://www.ecmwf.int/en/about/media-centre/aifs-blog/2024/data-driven-regional-modelling>



Aardvark Weather

<https://arxiv.org/pdf/2404.00411>

GenCast (DeepMind)

<https://arxiv.org/pdf/2312.15796>

Context

TT-AI4Wx - Overview

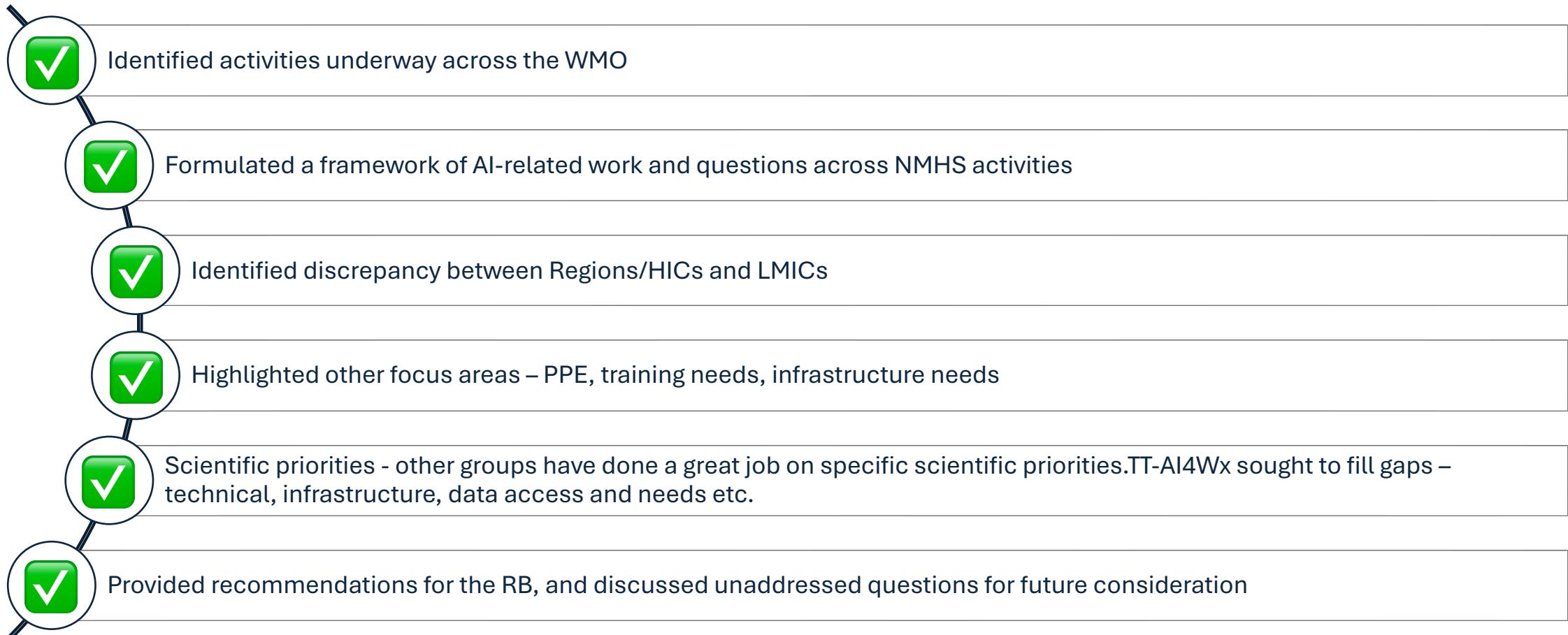
- A first response to the AI weather forecasting revolution
- Scope –

The TT-AI4WX will:

- *Generate a list of existing activities within WMO programs and regional associations;*
- *Review activities which can be expanded with limited resources to include AI/ML systems;*
- *Identify main gaps and science questions of immediate relevance from the community of experts making up the TT; and*
- *Develop a report on the above including recommendations on immediate actions.*
- Meetings initially fortnightly (~6 months) then monthly (~6 months)
- Membership from RB, SAP, WGNE, WWRP, SERCOM, INFCOM, GAW, and PPE office
- Note lacking RA presence – an issue for LMIC representation
- Concluded at 2025 RB meeting in September – Final Report accepted by RB vote

Context

Activities



Outcomes

Deliverables

- Interim report and open data concept note
- Paper database (~600 and counting...)

Accessibility Benchmark Datasets Climate CNN Coupled
 Data Assimilation Data Driven Models Downscaling
 Ensembles/Uncertainty Evaluation Explainability Finetuning
 Foundation Gappy data interpolation Geometric Deep Learning
 GNN Graph neural network Graph Transformer Graphs Group
 Hybrid Hydrology/Land Ice/Sea Ice LAM/Regional LLM
 Neural Operator Nowcasting Object Identification Ocean
 Other Parameterisation Perspective Platforms
 Postprocessing Review RNN Seminal Paper
 Single variable/process Space Weather Subseasonal/seasonal
 Survey Transformer Underpinning Theory/Speculative Research
 Visual Transformer Waves

- Final Report and framework of considerations
- Informed guidance on what next
- All outputs are here:
<https://community.wmo.int/en/task-team-artificial-intelligence-weather-tt-ai4wx>

Concept Note: Maintaining the open principles of the WMO in an AI/ML world

Context: With the advances in Machine Learning (ML)/Artificial Intelligence (AI) approaches in observations, predictions and services for weather, climate, and related Earth system information, profound transformations in the development, delivery, and access to hydrometeorological services are underway.

These include:

- AI/ML-based weather and climate models and emulators that could make it feasible to generate forecasts at very low computing costs
- Acceleration in the availability and broad access to data-driven prediction systems with improved quality and accuracy
- Data-driven systems that address the prediction needs at the local to very-local scales

It is the opinion of the Task Team participants that in order to realize these opportunities for all weather services and Members of WMO, the data, code, and intellectual property that

Model/ application Development	Fundamental research/ development	Scoping/strategic considerations	NHMS customization/ implementation	Operationalization	Postprocessing/ Downstream applications	Use of Information
Evaluation and Verification	<ul style="list-style-type: none"> Foundation models, Explainable AI (XAI), ESM extension (ocean, ice, land, hydrology, etc.), nowcasting Regional and global AI models, Ensembles, Nudging methods 	<ul style="list-style-type: none"> Evaluation of different AI architectures Reduction in cost/increase in timeliness/versatility of prediction systems Product focus (parameter/ global/regional/ensemble) 	<ul style="list-style-type: none"> Regional AI modelling/fine-tuning (Global) model fine-tuning Forecast nudging Measurement-model fusion by AI 	<ul style="list-style-type: none"> Path to operations - R2O processes, workflows for ML-ops (repeatable training and results) Online learning 	<ul style="list-style-type: none"> Downscaling AI for forecast augmentation and forecaster support 	<ul style="list-style-type: none"> Monitoring of forecaster use of AI support and augmentation
Observations (obs), Data Assimilation (DA), Initial Conditions	<ul style="list-style-type: none"> AI based observation QA, correction and monitoring AI for DA and observation operators AI forecasts directly from observations AI model performance in data sparse regions and outside training area 	<ul style="list-style-type: none"> Future obs infrastructure and design How much more of a priority will obs become? Optimal obs sites for AI Reduce cost/complexity of global DA Exploit regional/local knowledge 	<ul style="list-style-type: none"> AI based observation QA, correction and monitoring Develop regional AI-based forecast from obs 	<ul style="list-style-type: none"> AI based observation quality assurance, correction and monitoring Develop global, regional and local impact AI-based forecast from obs 	<ul style="list-style-type: none"> Verification of models AI based uncertainty quantification 	<ul style="list-style-type: none"> AI based evaluation of forecast outcomes User response analysis using AI, and of AI models Measure trust in AI forecasts
AI tooling and accessibility	<ul style="list-style-type: none"> Development of tooling for improved accessibility to training data and data/methodology openness Big data handling, geographically distributed data use 	<ul style="list-style-type: none"> AI infrastructure and software: <ul style="list-style-type: none"> - Cost and accessibility - Open-source frameworks Specific needs of developing countries 	<ul style="list-style-type: none"> Sharing of open access AI tools and experience Sharing of data Access to tooling and models in NHMSs in developing countries 	<ul style="list-style-type: none"> How to ensure data and models remain open Ongoing maintenance and support of tooling 	<ul style="list-style-type: none"> Integration of tooling to include postprocessing and downstream applications 	<ul style="list-style-type: none"> Routine assessment of state of the art and knowledge of AI tools and methodologies Assessment of the uptake of AI tools and technologies in developing countries
Training (especially for developing countries)	<ul style="list-style-type: none"> (People) Training in use of AI tools, ML model research and development 	<ul style="list-style-type: none"> Identifying training/ knowledge sharing needs Levels of complexity/expertise Specific needs of developing countries 	<ul style="list-style-type: none"> Establish platform for sharing experience and training materials Training in ML model finetuning, customisation and evaluation 	<ul style="list-style-type: none"> Training in ML model operational considerations and technical requirements 	<ul style="list-style-type: none"> Training in ML model postprocessing, and ML models for postprocessing 	<ul style="list-style-type: none"> Training in communication/user engagement around ML User education
Data/code access and availability	<ul style="list-style-type: none"> Data for training and retraining: collections, access, use guidance 	<ul style="list-style-type: none"> Data access and sharing guidelines on use Catalogue of collections Guidelines on public-private partnerships Ensuring ongoing open data and models as default 	<ul style="list-style-type: none"> Sharing Data for AI/ML training or retraining Sharing of inference weights or data 	<ul style="list-style-type: none"> Data for training or retraining Licensing of model models and model training data – implications for commercial use of forecasts 	<ul style="list-style-type: none"> Data for training or retraining 	<ul style="list-style-type: none"> Understanding obstacles to data access from users of information

Outcomes

Scientific Priorities

Interpretability and explainability

Improve AIWP performance on rare and high-impact events while enhancing trust and understanding through physical constraints, case sharing, and targeted benchmarking.

Upstream of the model - observations, DA and initial conditions

Enhance AIWP by integrating AI with data assimilation, optimizing observations, and evaluating model sensitivity to initial conditions, particularly in observation-sparse regions.

Downstream of the model - post-processing

Use AI to improve forecast skill and analyze extreme weather drivers, while managing high data volumes and ensuring careful quality control.

Model complexity and computational cost

Maintain or improve AIWP performance with smaller, efficient, physics-informed models using fine-tuning, regional models, or model distillation to reduce computational demands.

Infrastructure and data design and development

Enable equitable and efficient AI-enabled forecasting through lightweight models, shared platforms, improved data infrastructure, and reproducible standards.

Recommendations

Key Takeaways



Coordination & Agility

Strengthen cross-body collaboration; enable rapid AI adoption



Equity & Capacity Building

Empower underrepresented regions; provide training, infrastructure, and access to AI tools



Research & Innovation

Focus on AI methods, high-impact events, distributed computing; promote carbon-efficient workflows; address data-sparse regions and LMIC needs



Workforce Development

Build skills in AI implementation, interpretation, and ethics; include societal data expertise



Infrastructure & Standards

Ensure robust observation/reanalysis networks; establish shared platforms, data standards, open AI tools, and trustworthy evaluation

Looking forward

What does the WMO need to do differently/more?



Be faster, and more adaptable, and more able to work seamlessly across arms of the WMO

Emphasize supporting and advocating for positive PPE that ensures the mandated role of NMHSs is protected

Double down on explicit training, advocacy and support for LMICs – draw on and unify existing material

Looking forward

Future focus areas

ML for sub-seasonal, seasonal and climate



- Coupled models
- Insufficient evaluation data
- Impact of nonstationary climate on training
- Model stability
- Emulators and projections – climate change signal?

Equity and accessibility for LMICs



- Fairness in ML models
- Impact of observations sparsity in LMICs
- Need for evaluation over LMICs
- How to support needs of LMICs in terms of infrastructure?

ML for, and from, observations and DA



- The impact of ML on the design and scale of observation networks
- The future role reanalyses – design considerations for ML
- Progress in ML for DA
- Straight from observations ML models – implications for DA and reanalyses

Social sciences and ML

- The role of ML in modulating engagement how people engage with warnings and impacts
- How ML can be used in social sciences

Training and capacity building - should include academics



Foundation models

- One model for all timescales?

Evaluation and benchmarking



- One-shot and zero-shot models?

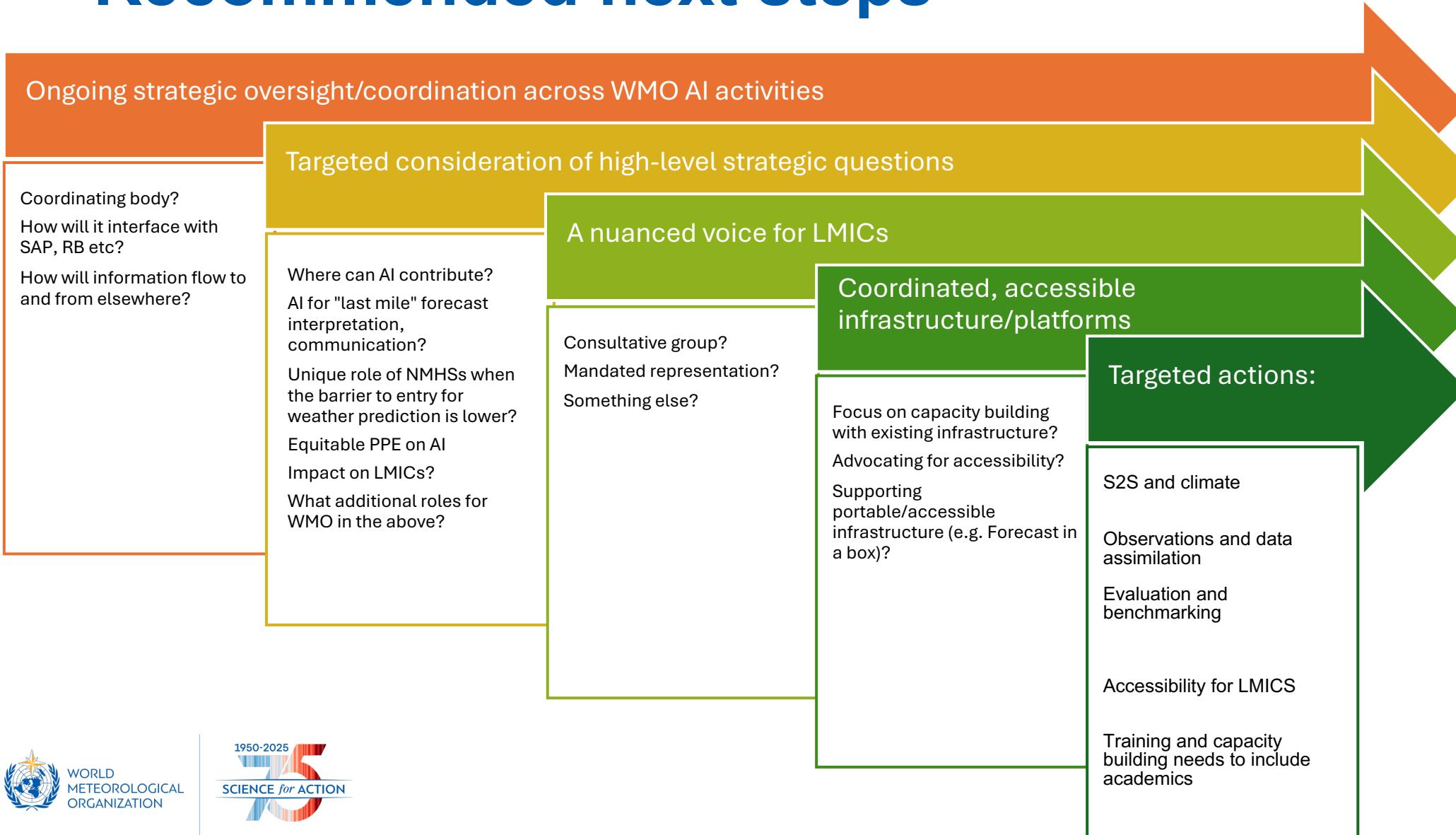
Infrastructure, tooling, and standards

- How to ensure open and accessible technologies, code, weights and data?
- How to support the development of accessible and affordable tooling?
- Best practice/standards for evaluation and benchmarking

*** Evaluation, verification and benchmarking is essential across all these areas**

Looking forward

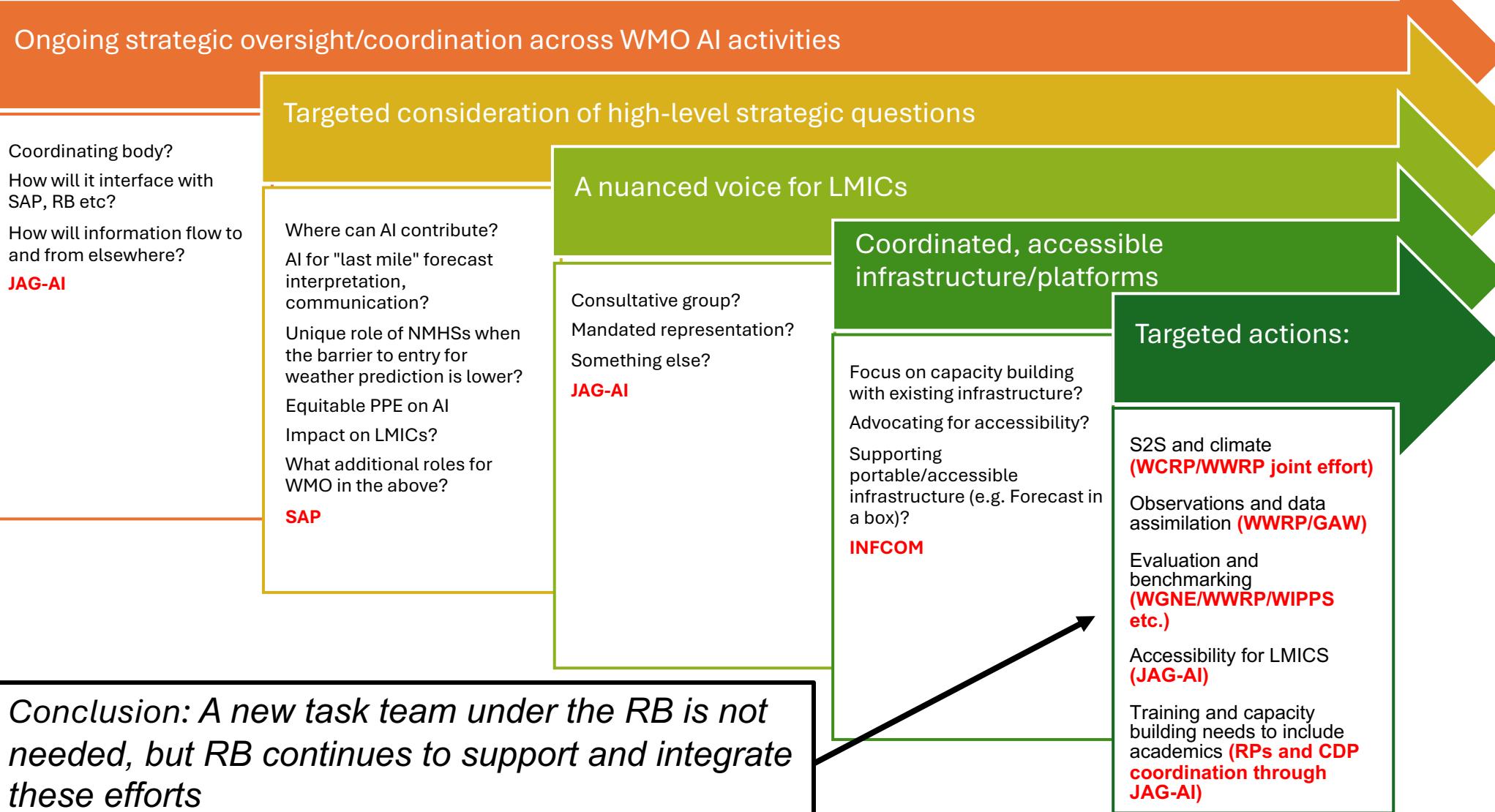
Recommended next steps



Risk:
expectations may be getting too high - How do we balance expectations against ability to deliver?

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Recommended next steps



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Summary

- The TT-AI4Wx has completed an important first step in the WMO's response to the AI revolution
- The WMO will need to continue to respond in full across all elements of NMHSs/member responsibilities
- The rapidly evolving nature of the field will require an equal (and thus arguably unprecedented) level of agility in the WMO
- The RB has noted the large number of activities now underway across the WMO and RB subsidiary bodies, and thus has not formed further AI-related TTs
- WGNE has a critical role to play in guiding and contributing to WMO activities on AI, and supporting WMO members