

September 3, 2021

# Exascale Computing & Data

*Opportunities and challenges for weather, climate and environmental prediction*

WEATHER CLIMATE WATER  
TEMPS CLIMAT EAU



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World Meteorological Organization  
Organisation météorologique mondiale

Kris Rowe

*Argonne Leadership Computing Facility*

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*NOAA, Global Systems Laboratory*



WORLD  
METEOROLOGICAL  
ORGANIZATION



Met Office



Environment and  
Climate Change Canada  
Environnement et  
Changement climatique Canada



JÜLICH  
Forschungszentrum

JÜLICH  
SUPERCOMPUTING  
CENTRE



University of  
Reading



気象庁  
Japan Meteorological Agency



NCAR



AIMS



GEOMAR

Helmholtz Centre for Ocean Research Kiel

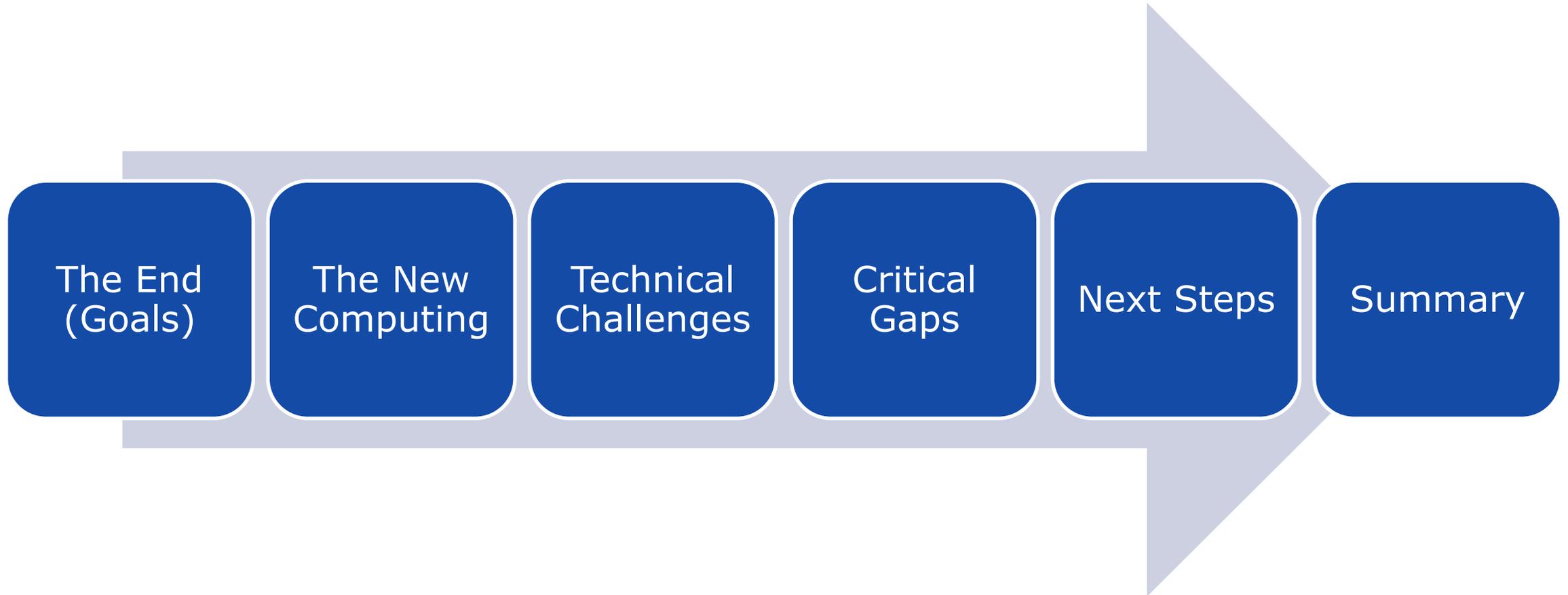
# The Team

WMO Research Board  
Task Team on Exascale,  
Data, and AI



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# The Big Picture



# The End

(Goals)

## Faster Prediction

- *Strong scaling*
- Weather\*

### Benefits

- Prevent loss of life and property
- Assimilate more recent data into predictions

## Higher Fidelity

- *Weak scaling*
- Climate\*

### Benefits

- Model coupled systems
- Resolve critical small-scale processes
- Improve accuracy of long-time simulations



# THE NEW COMPUTING

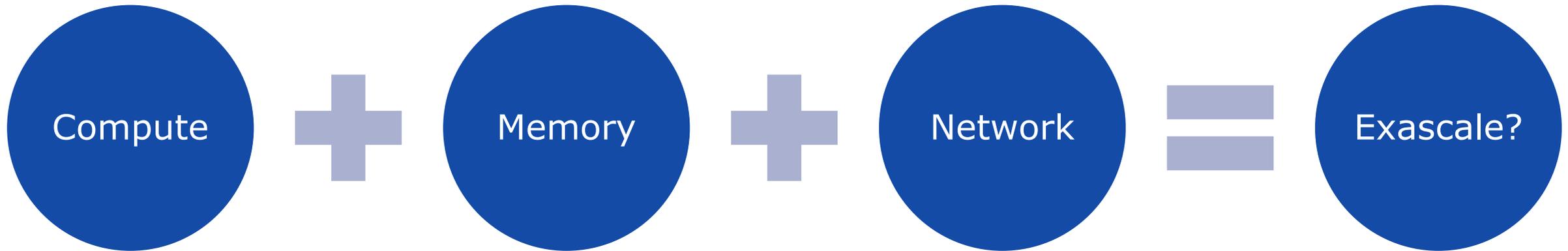
Form follows function



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# What is “Exascale”?

≥1 Exaflops/s (double precision) on the HPL benchmark



What about  
“AI Flops”?

CPU? GPU?  
Total?

Does cloud  
count?

The rigorous definition is not as important as  
the dramatic change in computer architecture!



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# Examples of Exascale Systems

## Key Details

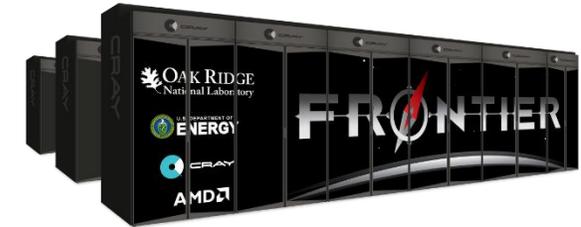
- Multi-CPU + Multi-GPU
- Multiple HPC GPU vendors
- Network and storage must complement compute

## ALCF



- > 1 Exaflops/s
- Per node:
  - 2 Intel "Sapphire Rapids" CPUs
  - 6 Intel "Pointe Vecchio" GPUs
  - 8 fabric endpoints (NICs)
- Cray Slingshot
  - 3 hop Dragonfly topology
- DAOS Storage:  $\geq 230$  PB,  $\geq 25$  TB/s

## OLCF



- > 1.5 Exaflops/s
- Per "blade":
  - 2 AMD EPYC CPUs
  - 8 AMD Radeon Instinct GPUs
  - Multiple fabric endpoints (NICs)
- Cray Slingshot
  - 3 hop Dragonfly topology
- Lustre Storage:  $\geq 250$  PB, 5-10 TB/s

## Why this architecture?

Multiple GPUs

HBMe

Multiple NICS



## Design Requirements

Flops, Bandwidth

Power Envelope

Cost Constraints

# TECHNICAL CHALLENGES

New computers, old problems



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# New Computers, Old Problems

- Traditionally end-users of prediction models focus on **scientific challenges**
- Concurrent with these are numerous **technical challenges** related to hardware, software, and human-factors
- Many of these challenges are not new, however, their difficulty and complexity are amplified in the exascale context
- Challenges are not independent: addressing or failing to address one may reduce or increase the difficulty of another



# Challenge #1

## Cost

### Hardware

- U.S exascale systems in the range \$300m-\$600m
- Electricity costs for 30 MW are more than \$12m USD
- Facilities and maintenance costs are also significant

### Software

- Development and maintenance costs are often overlooked
- Funding a team of research software engineers can cost millions/yr.
- Example: ECP has made significant investment in this area

### Environmental Impact

- Carbon footprint for 30 MW is over 100 Mt per year!
- EU Green New Deal requires data centres to be carbon neutral by 2030



## Challenge #2

# Data

- Estimated **0.5 PB** storage for 10-day forecast with
  - 3km resolution
  - 192 vertical
  - 3-hour output interval
- Storage for climate simulation will be significantly larger
- Data-in-place strategies now are fundamental
- Data loss or corruption must be addressed at this scale
- *In situ* analysis and visualization are essential tools
- With 5G/IOT volume of data to be assimilated will continue to grow



## Challenge #3

# Performance

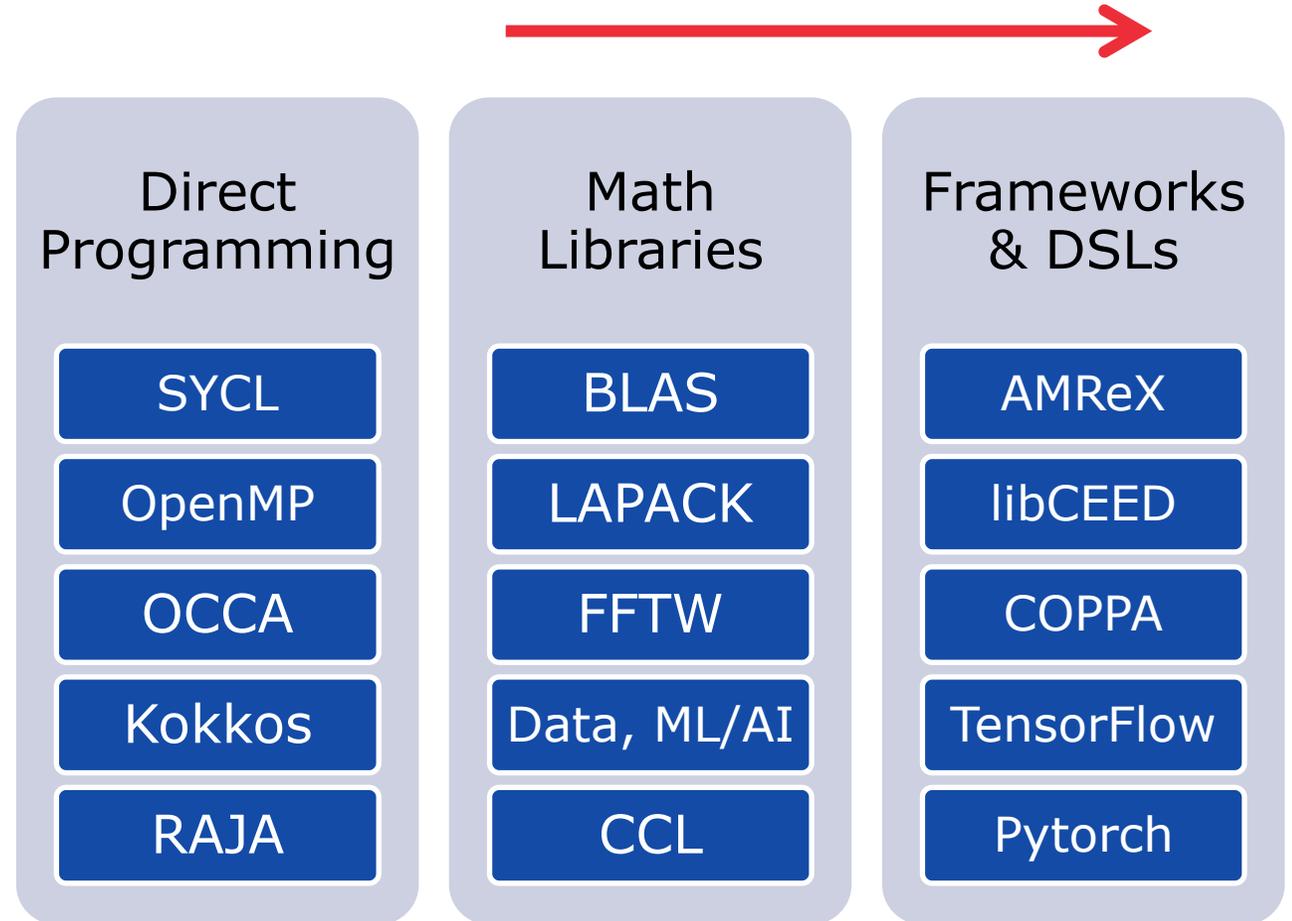
- Runtime is still the main performance metric
  - *Energy-to-solution* also key consideration
- Internode communication is still an issue for parallel scaling
  - e.g., halo exchange, global reductions
- GPUs require different data-layouts and algorithms which expose more parallelism
  - Performance tuning can be notoriously difficult
  - Subtle differences between vendor microarchitecture can be important
- Denser compute nodes require greater focus on intranode communication and optimization
- Algorithmic changes can sometimes provide the greatest benefit
- Mixed-precision techniques have significant momentum



## Challenge #4

# Portability

- Portable software can run on
  - Different types of hardware
  - Different vendors' hardware
- Goal is to minimize
  - Lines of source code needed to achieve portability
  - Effort to run existing code on new and future types of hardware
- Want turnkey performance
  - Otherwise with minimal (automatic) parameter tuning



## Challenge #5

# Productivity

The *Better Scientific Software* website is a great resource from the US ECP  
<https://bssw.io/>

- Ease with which software is developed, tested, shared, maintained, documented
- Following best practices is critical for creating high-quality scientific software
- Software which is modular, composable, and extensible retains greater value, can be more easily ported/adapted
- A **co-design approach** is optimal
  - Scientists and research software engineers working collaboratively, communicating effectively
  - Examples:
    - ECMWF's Scalability Programme
    - German Climate Computing Center + Max Planck Institute for Meteorology
    - US ECP Co-design Centers: CEED, AMReX, COPA



# CRITICAL GAPS

And ways to address them



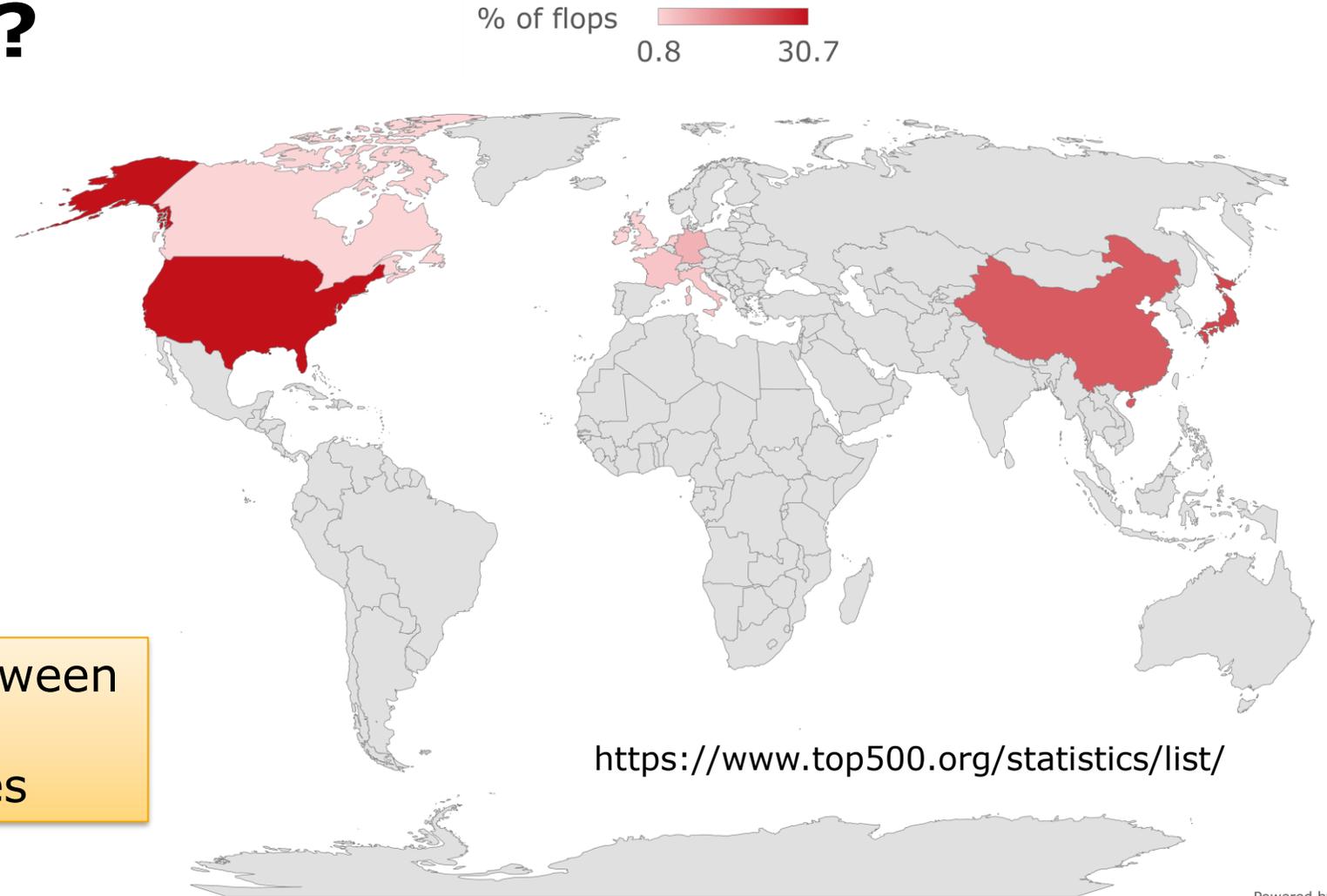
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# The Benefits of Exascale—for Who?

- Over **98%** of computing power worldwide is in Europe, Asia, and North America
- Over **72%** belongs to China, Japan, and the U.S.
- Regions most at risk from climate change have few or no computing resources

Goal: Identify critical gaps between members with and without significant computing resources

## TOP 500 Share of Performance



# Gap #1: Access to Sufficient HPC Resources

## Observations

- Many HPC centers can be accessed worldwide
- Remote visualization and data analysis require only a laptop
- Significant software development can be completed remotely
- Compute requirements cannot be avoided!

## Potential Solutions

- International allocation programs (e.g., INCITE) exist for research
  - Create awareness of these programs; connect researchers
  - Not available for operation forecasting needs
- Collaboration between Members with & without HPC resources
- Possible service opportunities for industry



# Gap #2: Access to Data Resources, Storage, Analysis Tools

## Observations

- Data now often too big to transfer, remains on-site
- Learning to use specialized or niche tools can be a barrier
- Consumers have different use cases, need different aspects
  - Potentially at-odds with *in-situ* analysis, data reduction

## Potential Solutions

- Large centers can provide shared external access to
  - Data sets
  - Storage
  - Visualization & analysis nodes
- Advocate for free and open-source community-based tools
- Develop models to accept external analysis “plugins”
- Include at-risk regions in the creation of data standards



# Gap #3: Access to Specialized Knowledge & Skills

## Observations

- Developing models for large-scale HPC requires advanced knowledge in
  - applied mathematics
  - computer science
  - software engineering
- Allocation programs require that researchers and their codes can make effective use of resources
- Members with significant computing resources have pipelines to train researchers
  - E.g., workshops, summer schools, internships/fellowships

## Potential Solutions

- Create broader awareness of existing resources:
  - Training material from hardware vendors, major computing centers, government funded projects
  - Workshops/summer schools
    - E.g., ATPESC
- WMO develop shared, specialized training resources focused on applying HPC, AI, data techniques in the in the weather/climate
- International technical meeting using [ECP Annual Meeting](#) as a prototype

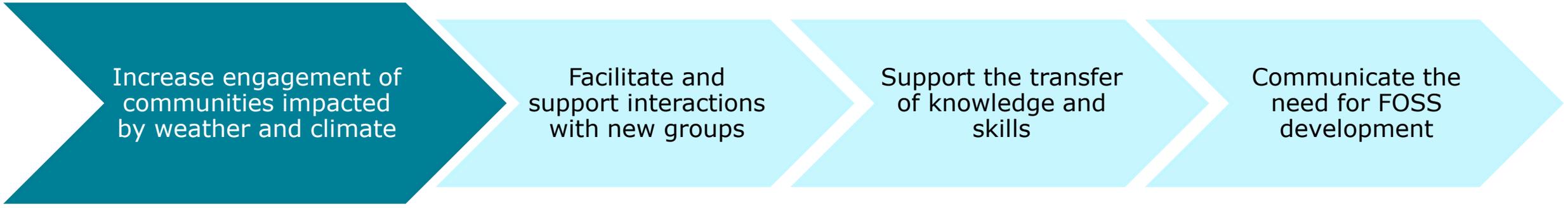
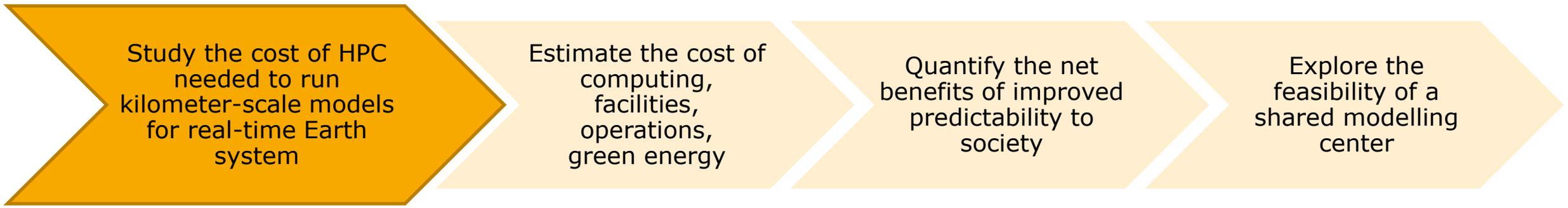
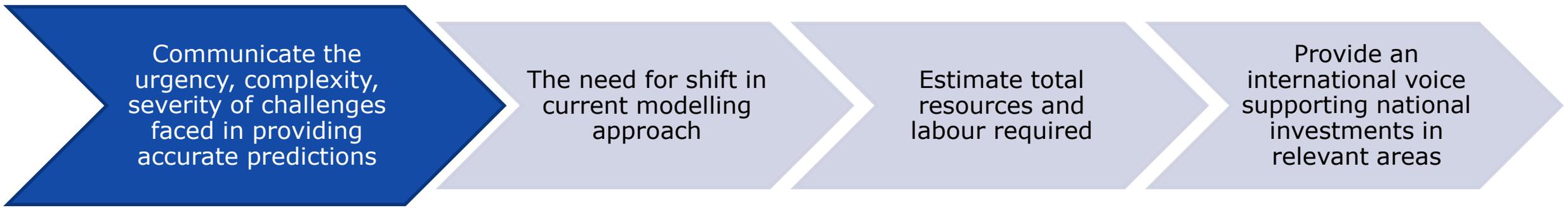


# NEXT STEPS

Concrete actions for the next 12-18 months



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# Summary of Recommendations

*For the Research Board and WMO Members*

We recommend urgency in dedicating efforts and attention to disruptions associated with evolving computing technologies that will be increasingly difficult to overcome, threatening continued advancements prediction capabilities.

The increasing scientific and computing complexity will require major efforts to adapt or rewrite earth system prediction models. In addition to scientific accuracy, models must be developed for performance, portability, and productivity.

The cost of computing resources, power consumption, and the related carbon footprint must be considered along with the benefit of improved predictability. Requirements to make data centers carbon neutral are already in force in a growing number of countries.

Scientists, model developers, computer scientists and software engineers need to work as equal partners on design, development, and maintenance of applications to overcome scientific, computing, and data challenges.

A data-in-place strategy is needed to support the increase in data volume from observations, model and ensemble output, and post processing. This will require co-location of HPC and data, with methods to access, extract, analyze, visualize, and store data by requesting processes & users.



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# Summary of Recommendations

## *For the Scientific Advisory Panel*

Few organizations will be able to fully address the software and data handling challenges, let alone provision the necessary supercomputing to continue to increase the scientific performance of their codes. A common, shared center could strengthen collaborative research on science, tools, software, and other development activities.

Scientists from regions lacking access to HPC resources face additional difficulties in adapting to this evolution. Large centers should be encouraged to provide open access to some shared resources as the most effective way for the community to collaborate, foster training and make improvements in all aspects of the prediction system.



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