

Ensemble approaches to enhance climate prediction

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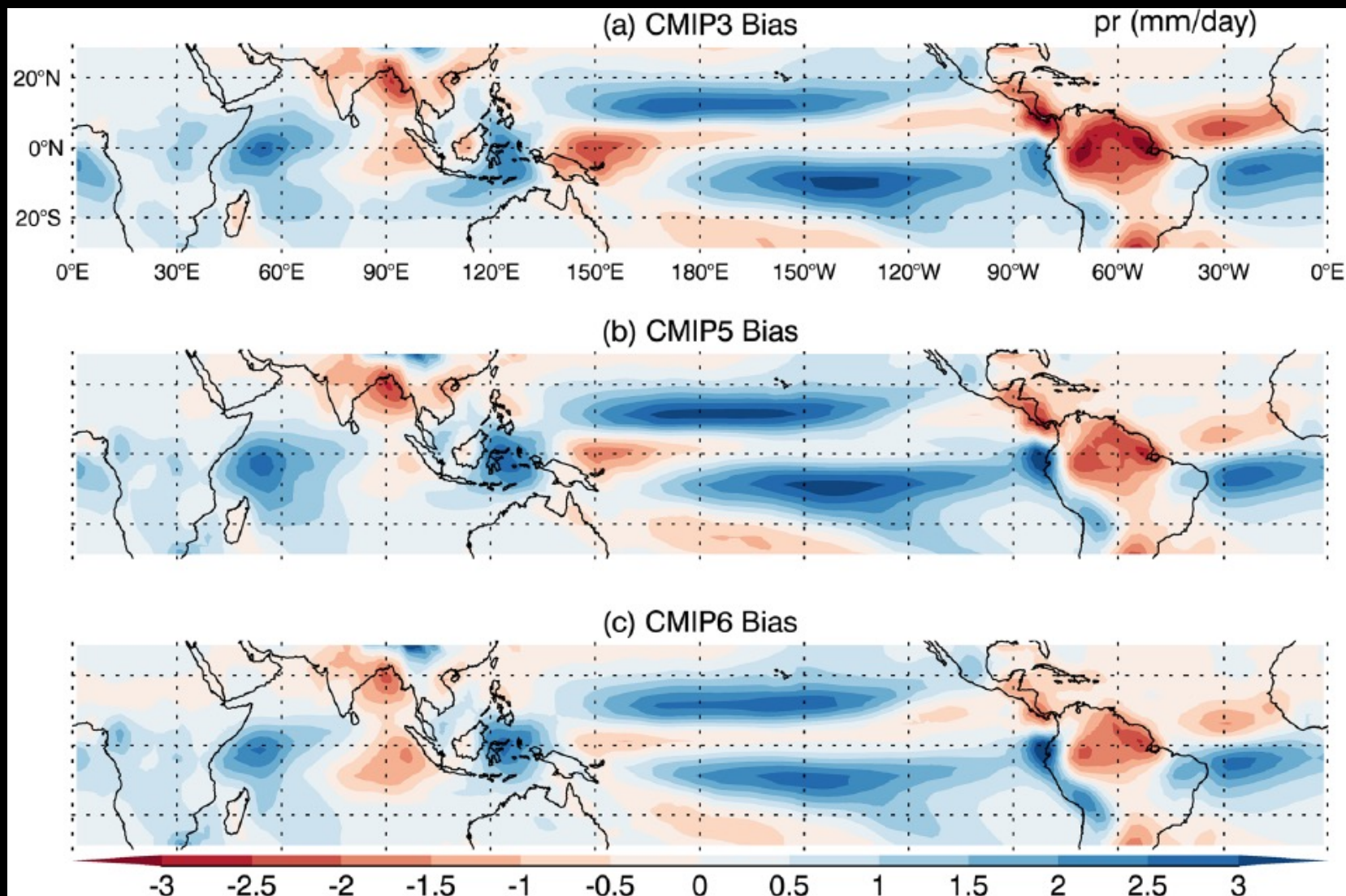
A.M. Obukhov Institute of Atmospheric Physics RAS

Bjerknes Centre for Climate Research, Norway



Persistent model biases

Multi-model precipitation biases across model generations



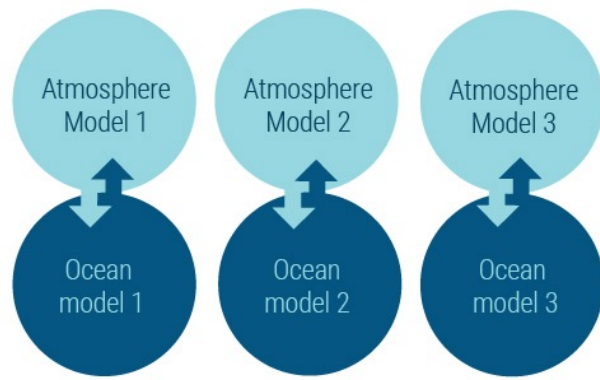
Tian & Dong 2020

Bias is often larger than the signal we analyze or predict



Standard approaches to deal with model error – unconnected multi-model ensemble

Standard modelling

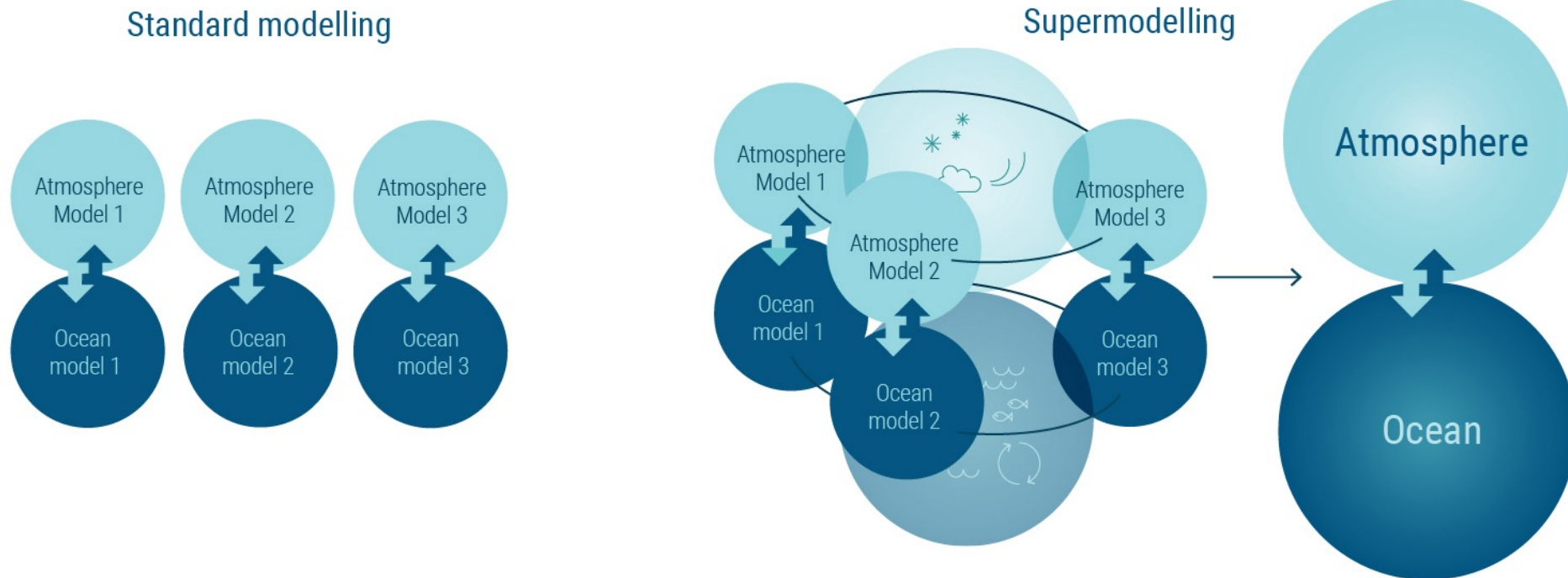


Long-term strategy – improve models to represent key processes, through e.g., increasing resolution

Current practical solution – combine outputs of different models in large-ensembles



A smarter ensemble approach – the supermodel



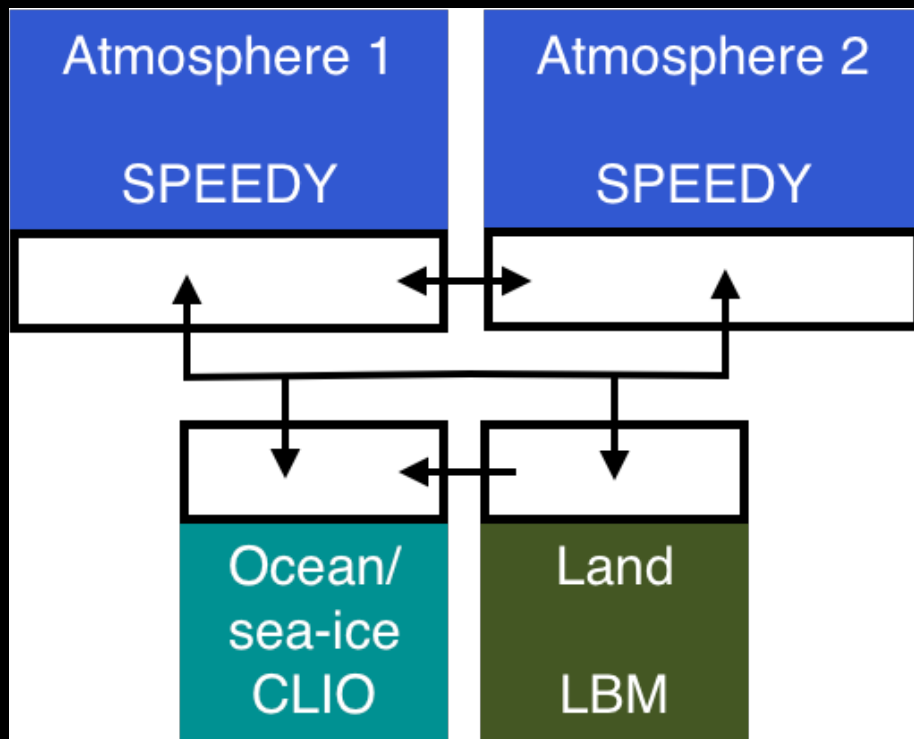
A supermodel is an optimal dynamical combination of models that is superior to its individual constituent models



Two different supermodel implementations

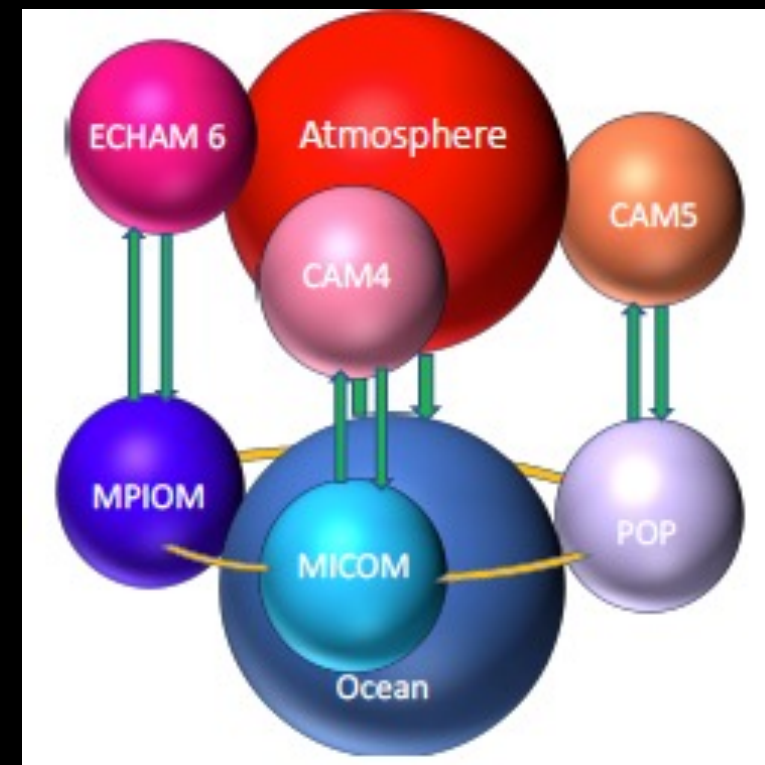
Multiple “imperfect” models are connected via atmospheric components and trained to reproduce a “true” model

Based on the SPEEDO intermediate climate model



Schevenhoven et al. 2019

Three state-of-the-art ESM are connected via ocean components and trained to reproduce observations

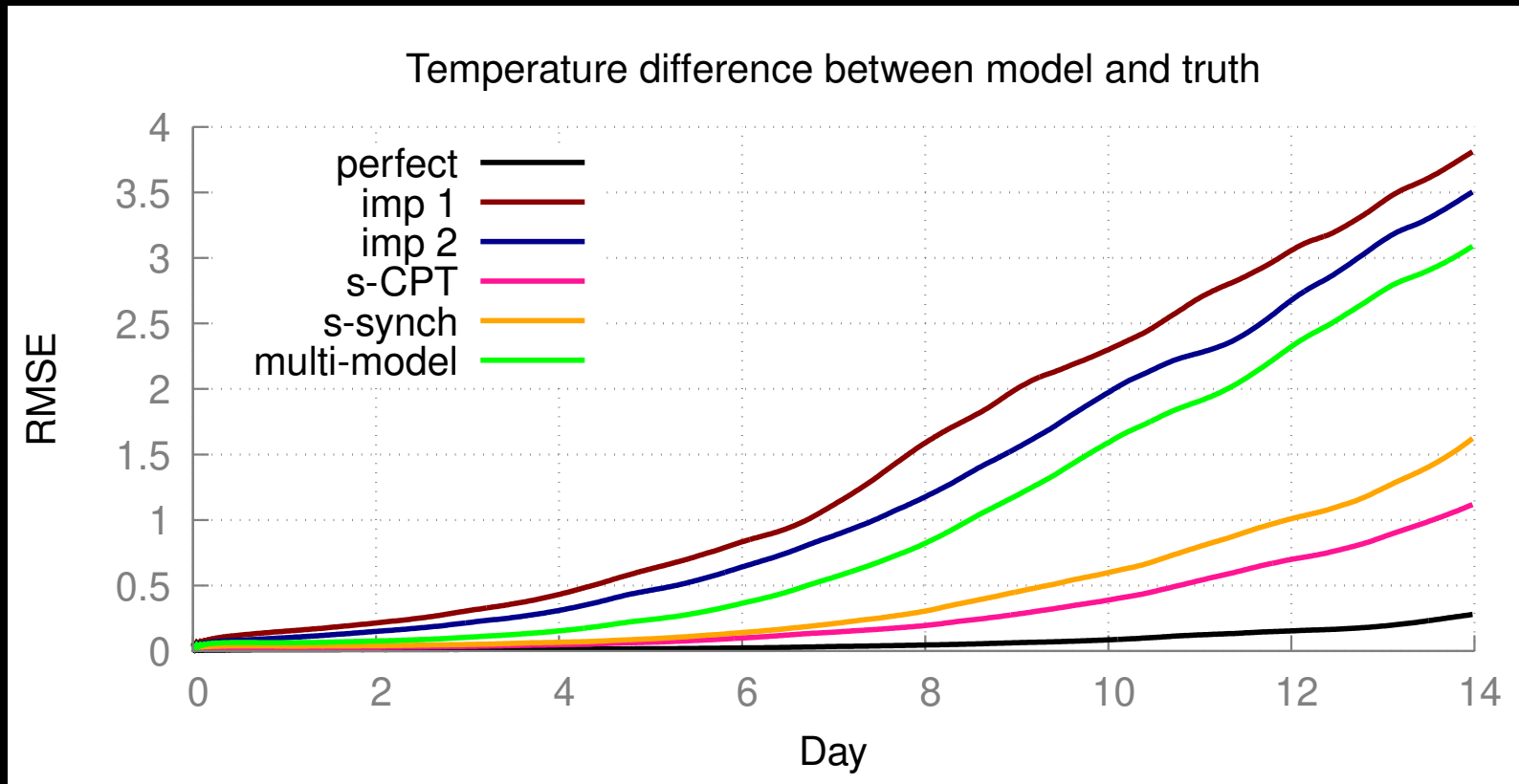


Counillon et al. in prep.

Supermodel improves weather forecasts

Results from intermediate complexity GCM super model
SPEEDO model with horizontal resolution of 3.75°

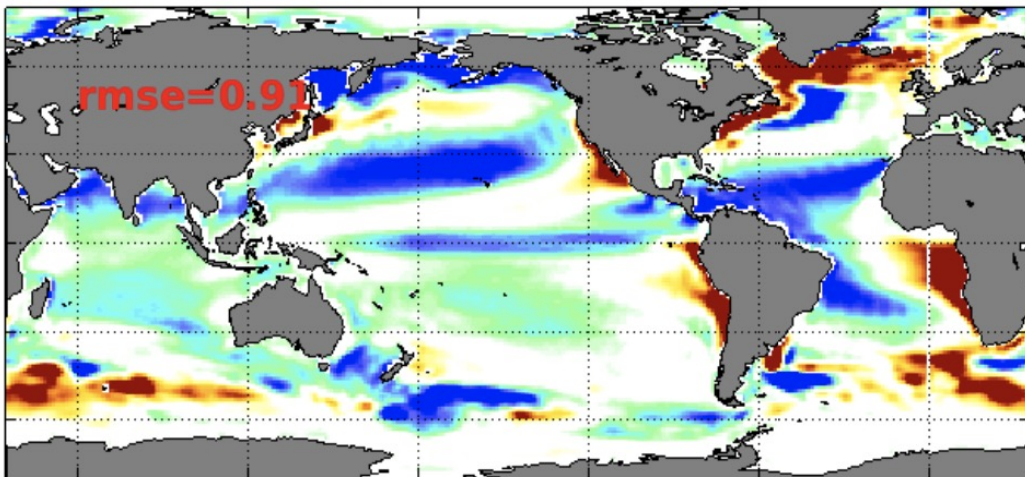
RMSE for 2-week forecasts of surface temperature, 25 forecasts



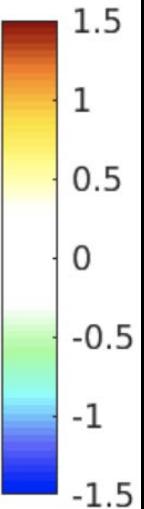
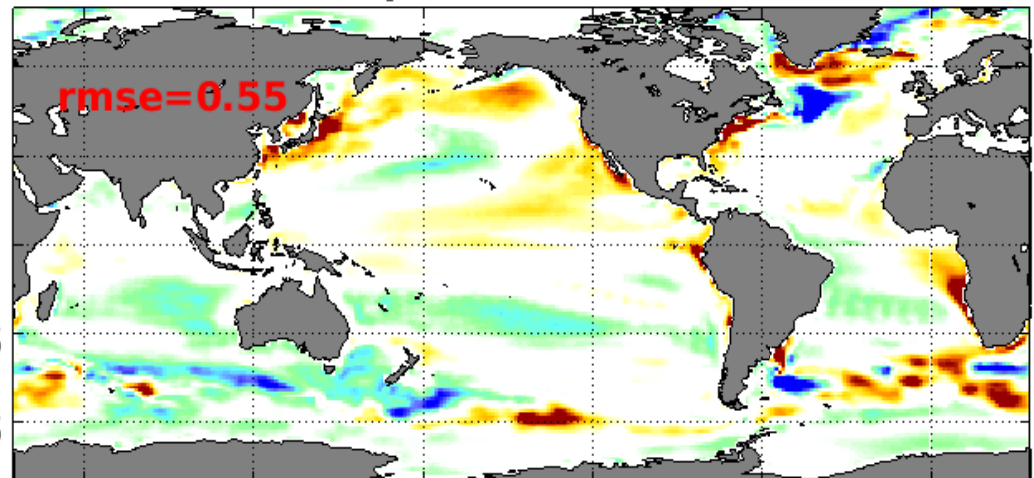
Supermodel reduces SST biases

NorESM-MPIESM-CESM supermodel
connected by assimilating optimally weighted model SST
25 years of simulation

SST bias – standard ensemble average



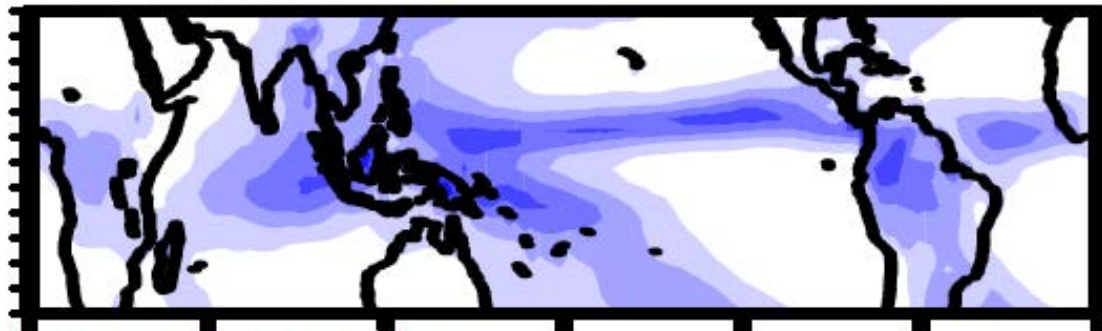
SST bias – supermodel



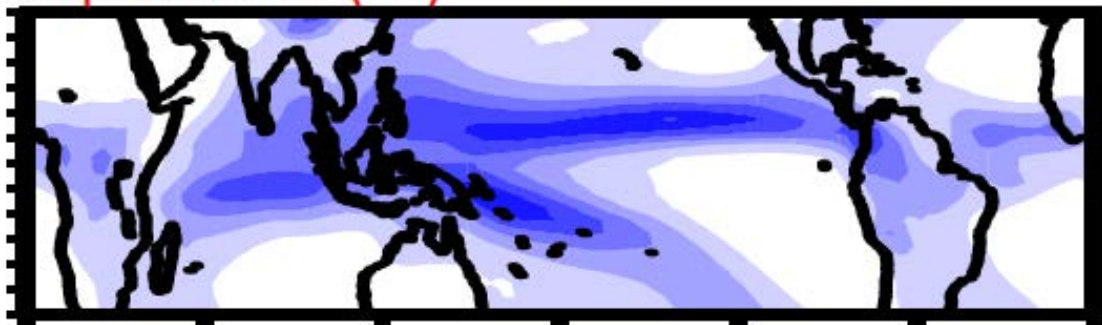
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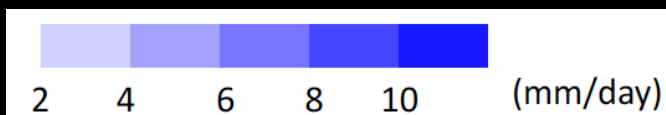
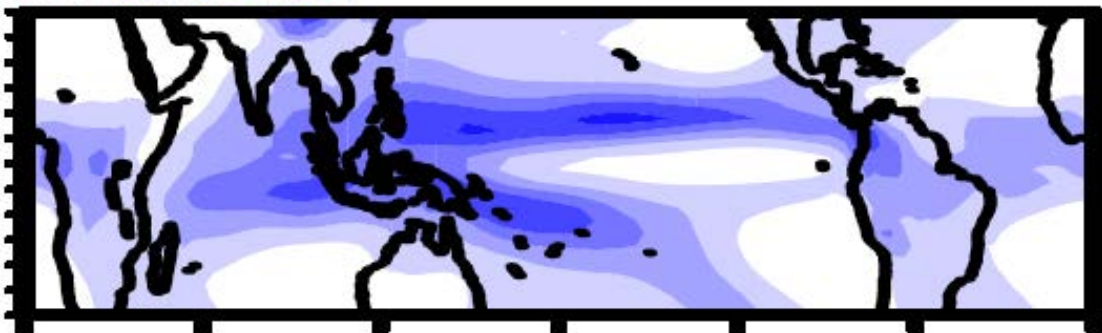
Observation



Supermodel (PF)



Unconnected



Rainfall climatology for the period 1980-2005

*Rainfall patterns in the tropics
are improved
using optimal weights*

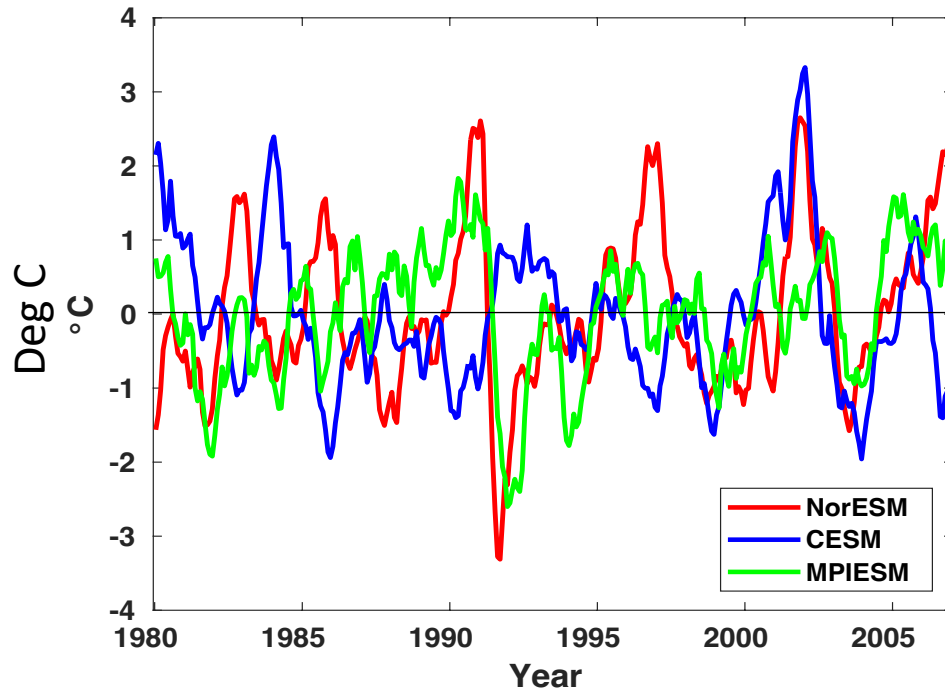
*Improvement beyond
standard multi-model mean*



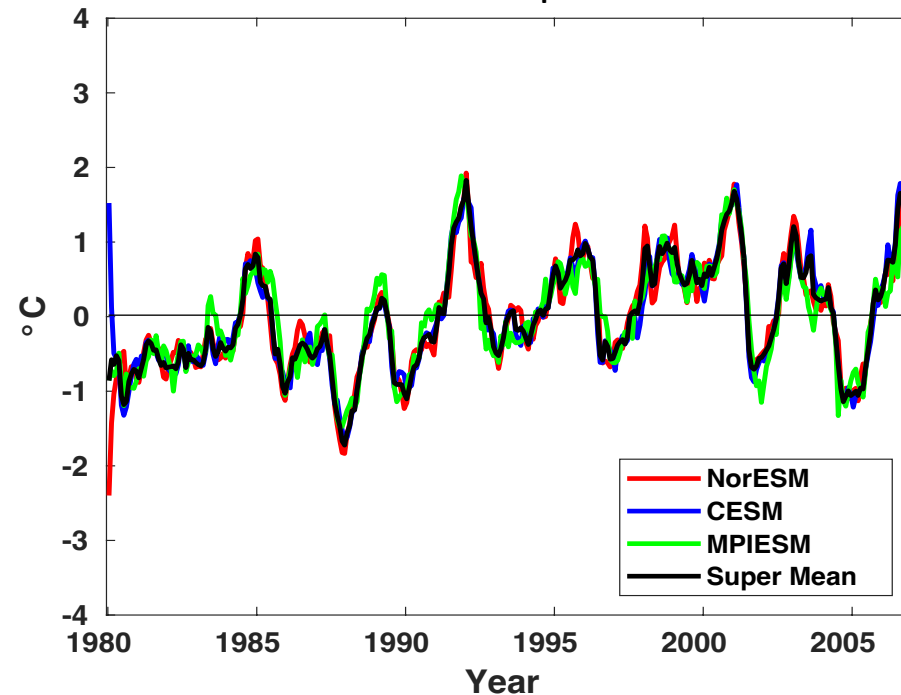
Supermodel leads to synchronized ENSO variability

NorESM-MPIESM-CESM supermodel
connected by assimilating optimally weighted model SST
25 years of simulation

Niño3.4 SST – standard ensemble



Niño3.4 SST – supermodel



SPEEDO supermodel better captures response to CO₂ doubling

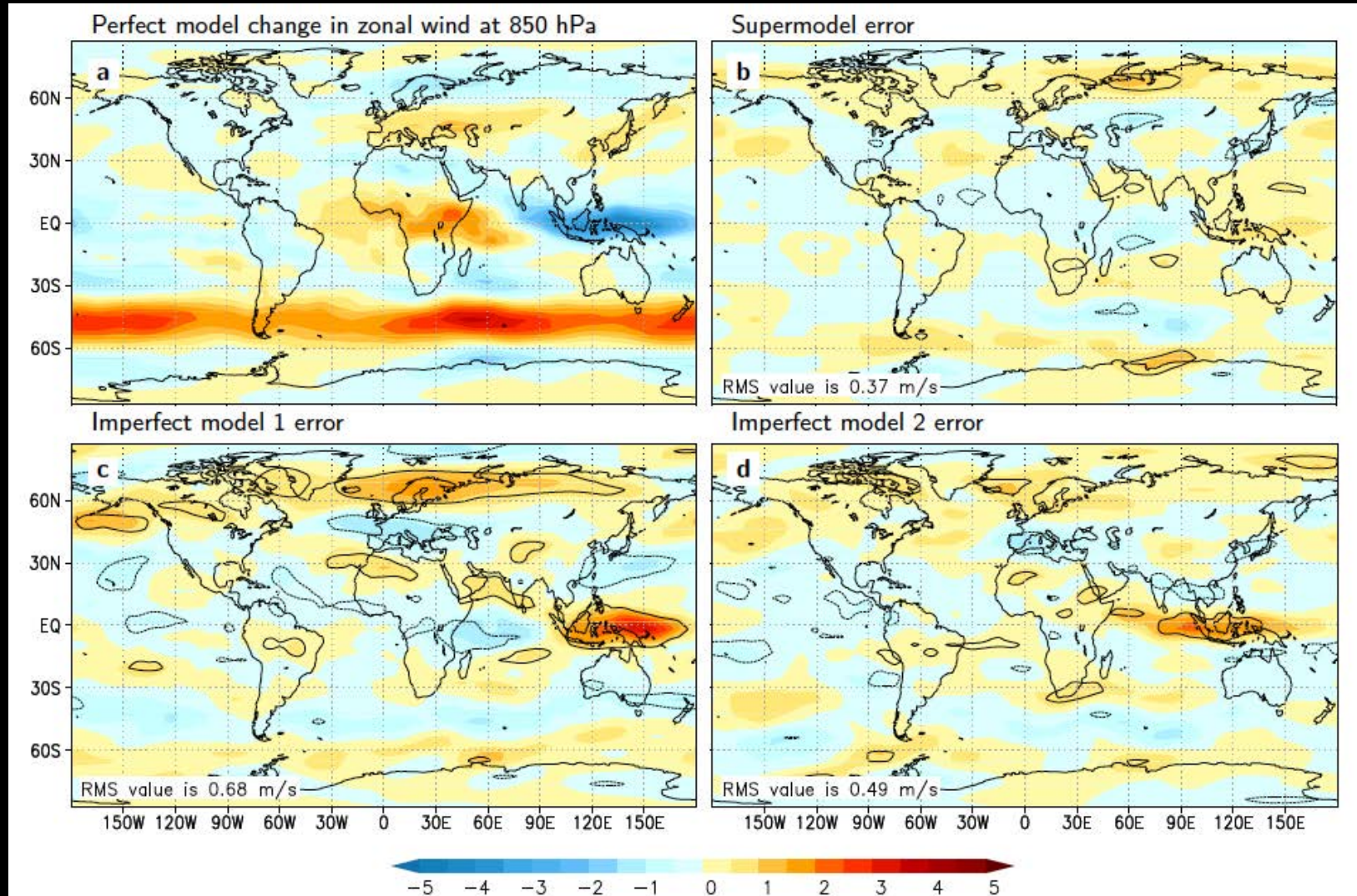
Climatological 850 hPa zonal wind (m/s)

Perfect model
response

Error of
supermodel

Error or
imperfect
model 1

Error or
imperfect
model 2





New avenue to improve climate prediction

- Supermodelling – interactive combination of models
- Compensating errors at early stage can reduce structural errors, which would otherwise be enhanced by non-linearity
- *Improvements greater than the standard ensemble approaches*
- *Can improve short-term and long-term predictions*

Thank you for your attention

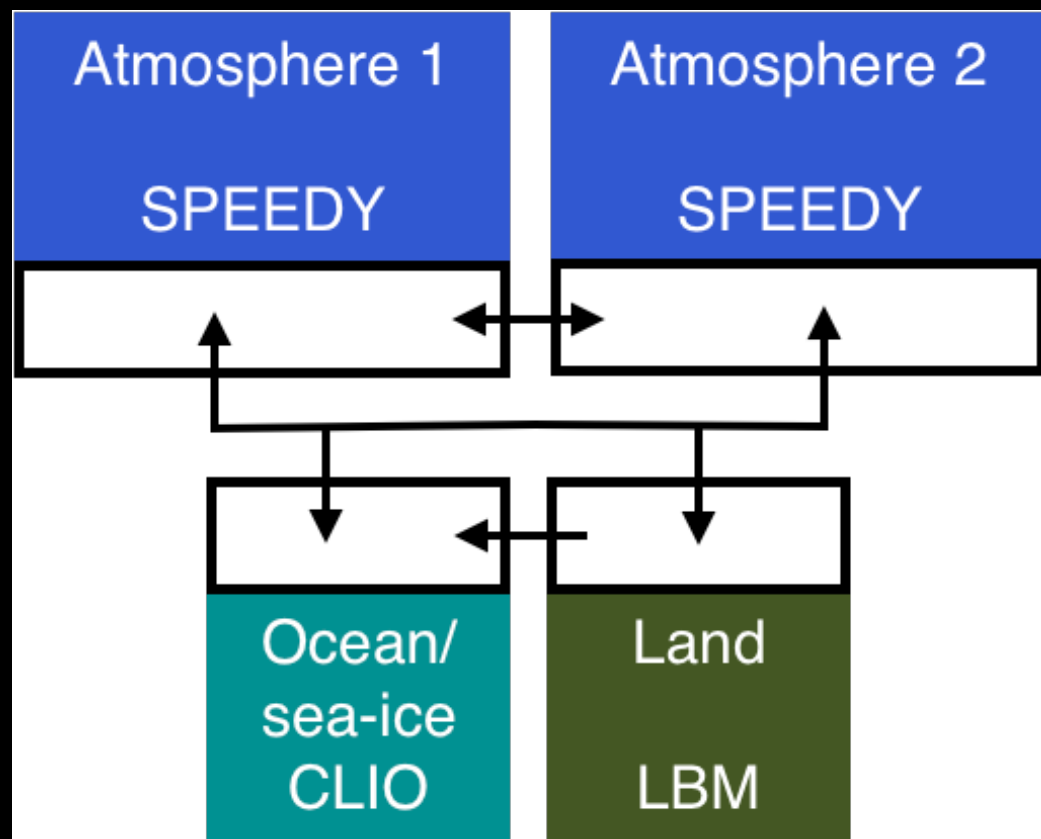


Methods



Supermodel with connected atmospheres – perfect model case

Two imperfect models are trained to reproduce a “true” model



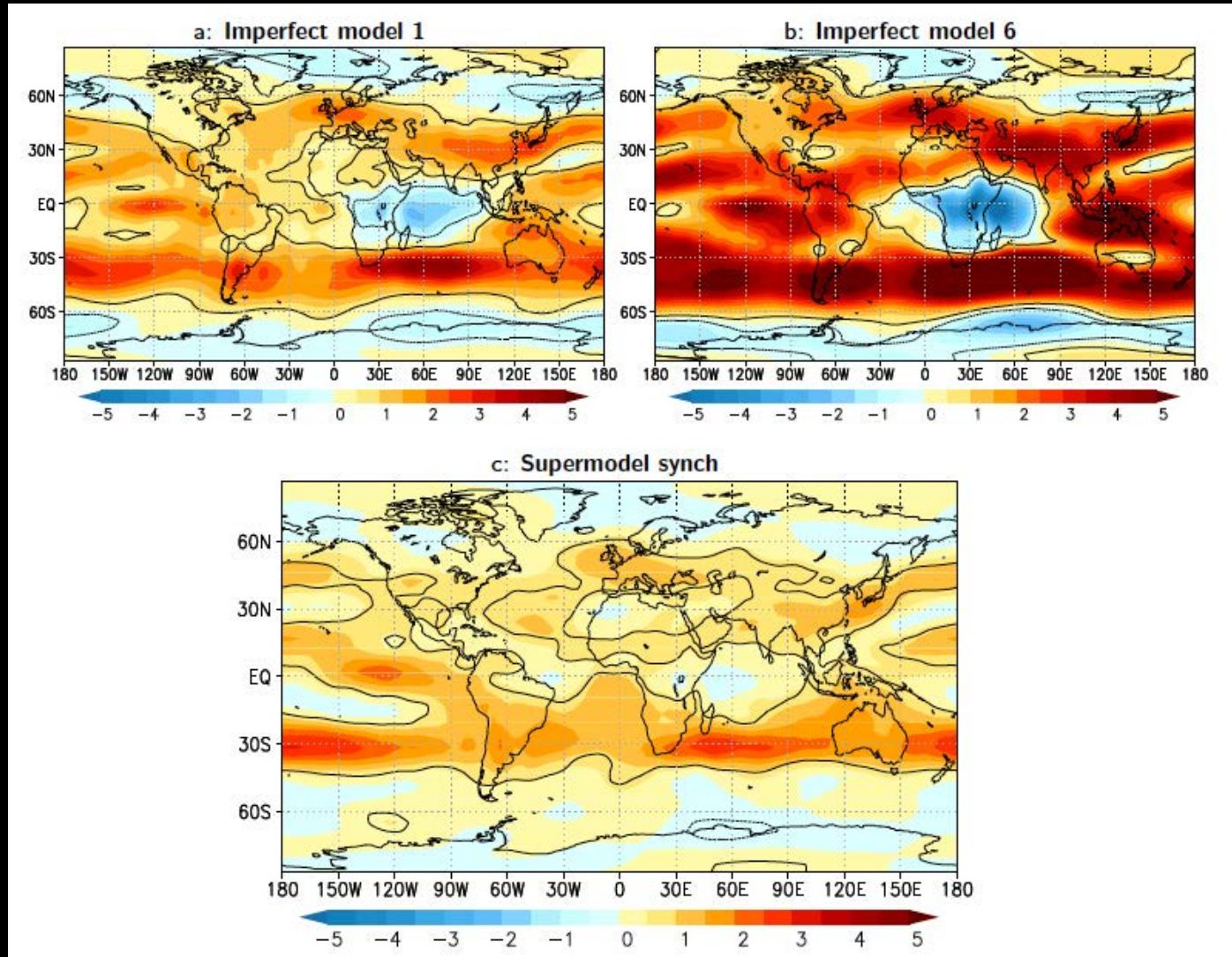
- Intermediate complexity climate model with $\sim 4 \times 4^\circ$ horizontal grid (SPEEDO)
- Connected by 3D states of temperature, vorticity and divergence every 15 minute timestep
- Global connections one value per field, per model
- Interactive ensemble with one ocean and multiple atmosphere (up to 4)
- Perfect model experiments, with imperfect models differing in parameters related to atmospheric convection
- Two new training (machine learning) methods: cross pollination in time, and synchronized based learning

Schevenhoven et al. 2019



Supermodel also improves climate simulations

Climatological error in 850 hPa zonal wind (m/s)



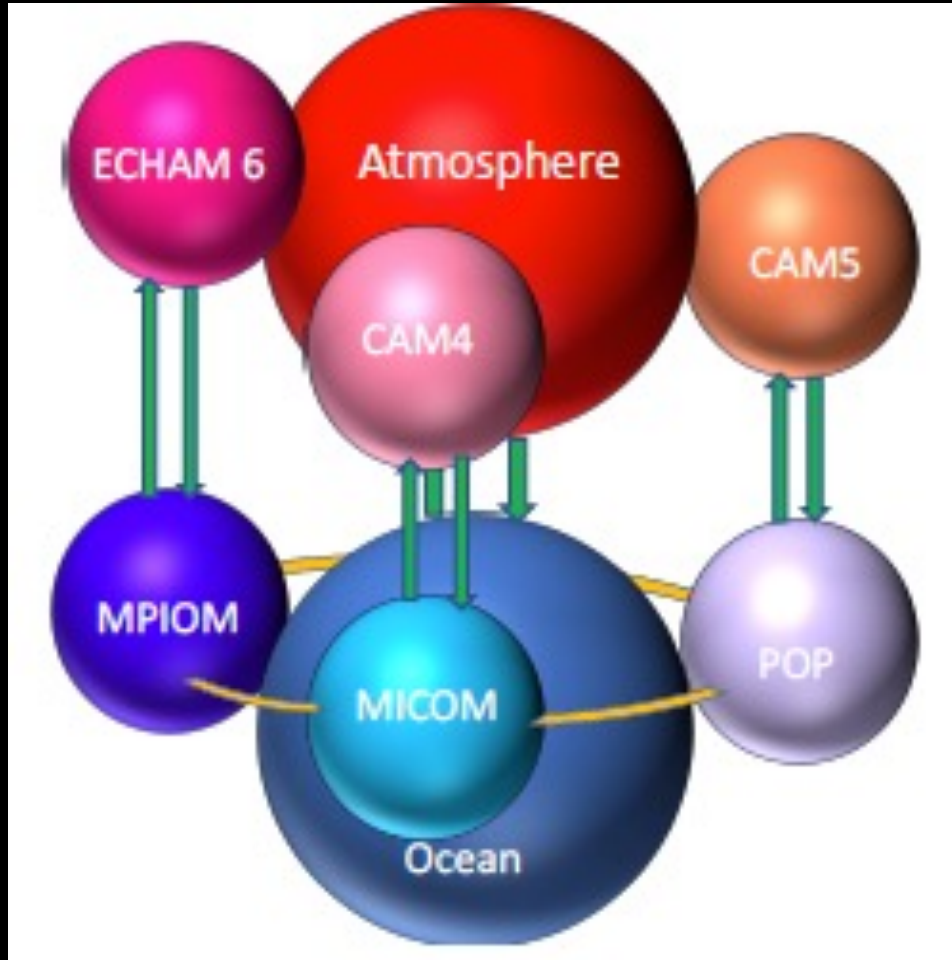
← Imperfect models

← Supermodel

Contours indicate where errors
significant at 95% confidence level

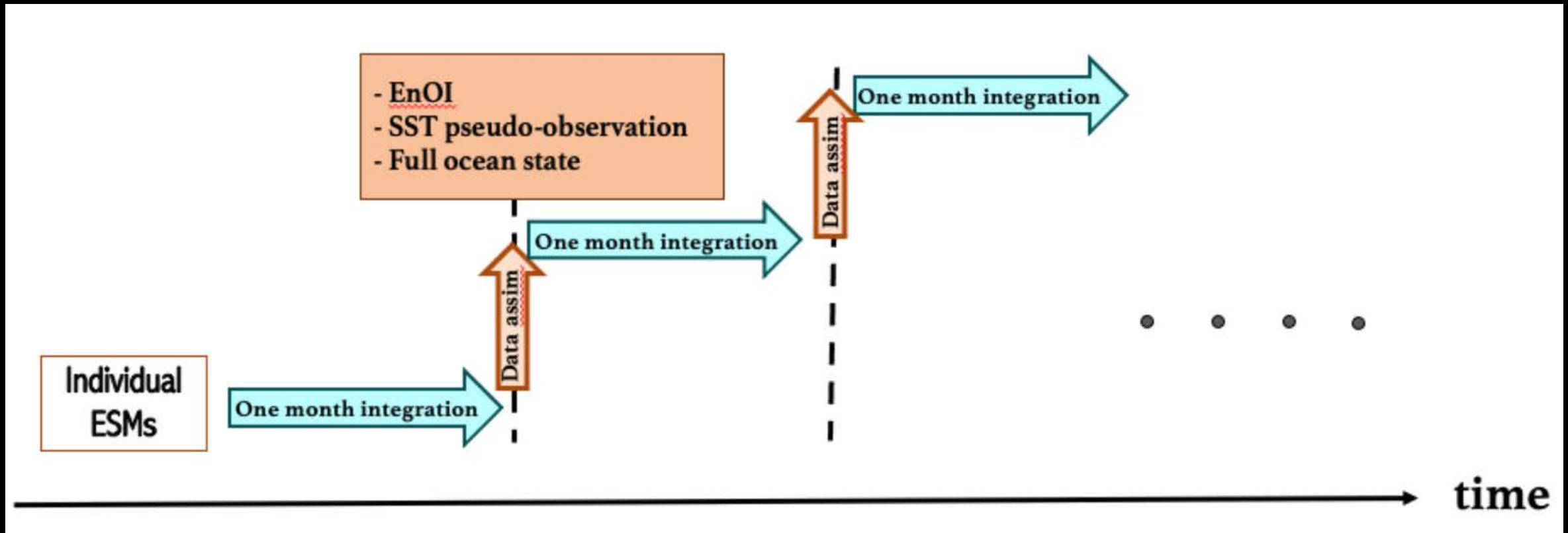
Schevenhoven et al. 2019

An ocean connected super-ESM – real world case



- Generate pseudo-observations from the three individual CMIP5/6 ESM
- Assimilate pseudo-observations back into each models every month (SST only)
- The three models are then propagated forward
- Several different offline approaches tested to train supermodel using observed SST data (all improve performance)
- Some results here for 25-year long simulations, training from one-month forecasts
- ~10 years per day, 11 nodes 1408 cpu

Connecting the ocean components of ESM



Pseudo-observations = a weighted combination of the individual model simulations