

A simple ocean performance metrics applied to historical CMIP5 simulations



Motivation and introduction

- For atmospheric models performance metrics very common, for example Reichler and Kim (2008)
- Not for ocean models !!!**
- Here we define simple ocean performance metrics in a similar way as Reichler and Kim (2008) did for the atmosphere.

Performance metrics

- For each 3D grid point of the PHC climatology the absolute error for potential temperature T and salinity S is calculated and averaged over ocean basins / the global ocean
- The mean absolute error over all CMIP5 models for an ocean basin / the global ocean serves as a reference and a specific model can be compared to the CMIP5 ensemble
- A performance index (PI) of 1 indicates same performance as CMIP5 ensemble
- A PI of less than 1 indicates better performance than CMIP5 ensemble, of greater than 1 worse performance

Example output: performance of AWI-CM

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global S DJF 0.800658287      pacoce S DJF 0.956561685      atloce S DJF 0.746122181
global S JJA 0.792248607      pacoce S JJA 0.922508359      atloce S JJA 0.750263929
global T DJF 0.809991717      pacoce T DJF 0.936220169      atloce T DJF 0.711102664
global T JJA 0.756092548      pacoce T JJA 0.839840889      average:atloce 0.739893973
average:global 0.789754510      npacoce S DJF 1.02797759      natloce S DJF 0.633922219
southoce S DJF 0.713550925      npacoce S JJA 0.972718418      natloce S JJA 0.636811793
southoce S JJA 0.642232180      npacoce T DJF 1.03628802      natloce T DJF 0.729878187
southoce T DJF 0.582666814      npacoce T JJA 0.929830909      natloce T JJA 0.813130796
southoce T JJA 0.625893474      average:npacoce 0.991703749      average:natloce 0.703435799
average:southoce 0.641085863      spacoce S DJF 0.893189490      satloce S DJF 0.896557152
indoce S DJF 0.637605846      spacoce S JJA 0.878341854      satloce S JJA 0.903289855
indoce S JJA 0.651327014      spacoce T DJF 0.847720480      satloce T DJF 0.691242218
indoce T DJF 0.717630625      spacoce T JJA 0.753928125      satloce T JJA 0.685272276
indoce T JJA 0.591852546      average:spacoce 0.843169987      average:satloce 0.7940900390
average:indoce 0.649603963
  
```

One sees straight away in which area / parameter / season the model performs **better** / **worse** than CMIP5 average. In this example: AWI-CM very good!

Conclusions

- A simple ocean model performance metrics has been defined and applied to CMIP5 and prototype HighResMIP simulations
- Allows to quickly diagnose in which ocean basin and in which depth the model drift is strongest
- State-of-the-art ocean models show large errors which exceed the interannual variability and from 500 m depth downwards even the climate change signal
- Shows that in ocean models there is still much room for improvements

Vertical profiles

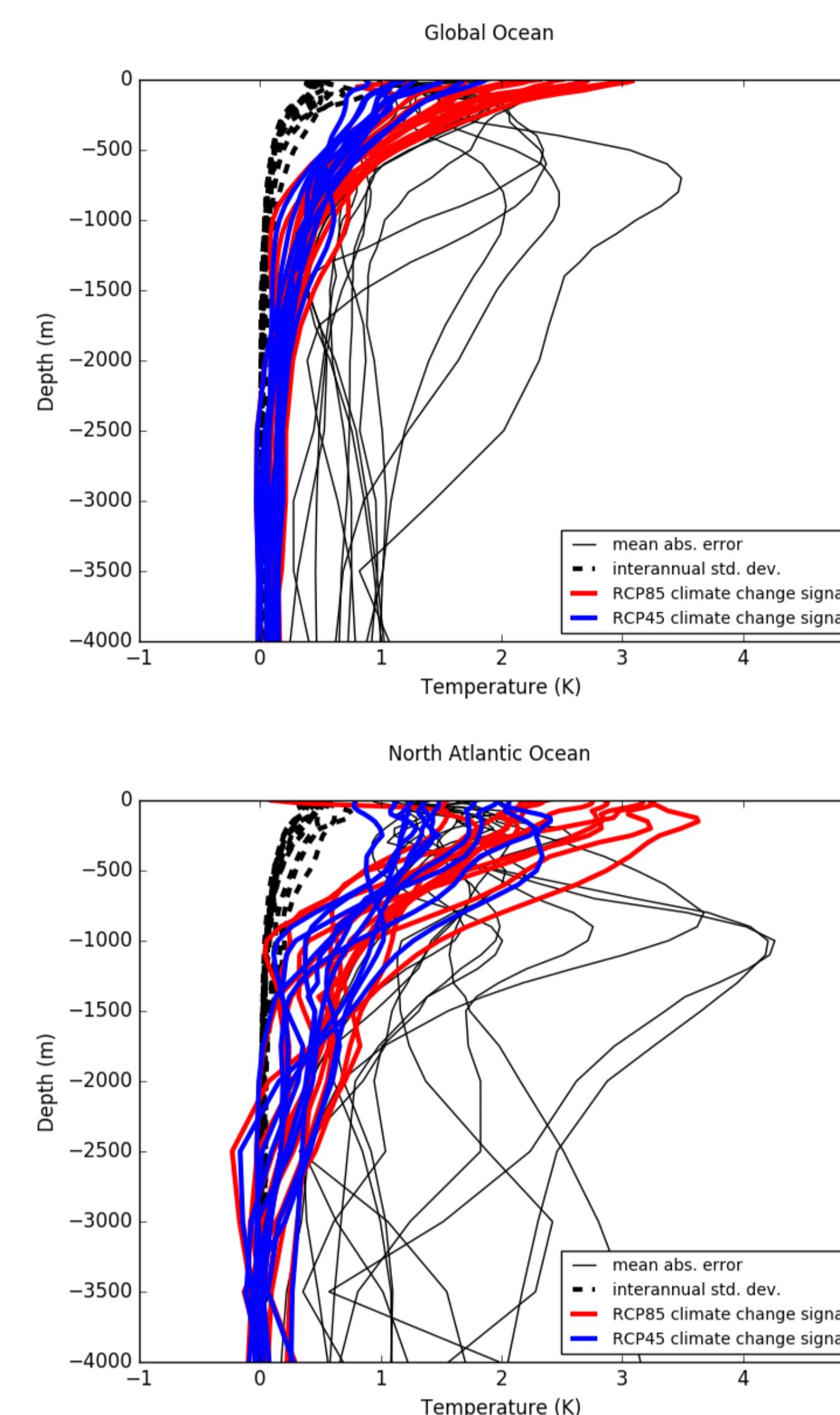


Fig. 1: Profiles of potential temperature mean absolute error for years 1971–2000 of the historical simulations from 13 CMIP5 models contrasted to model interannual standard deviation and climate change signal 2071–2100 minus 1971–2000.

Above: global ocean
Below: North Atlantic ocean

- The mean absolute error is everywhere and in all models larger than the interannual standard deviation.
- Except for near-surface layers it is even larger than the climate change signal.

Potential temperature bias 1000 m

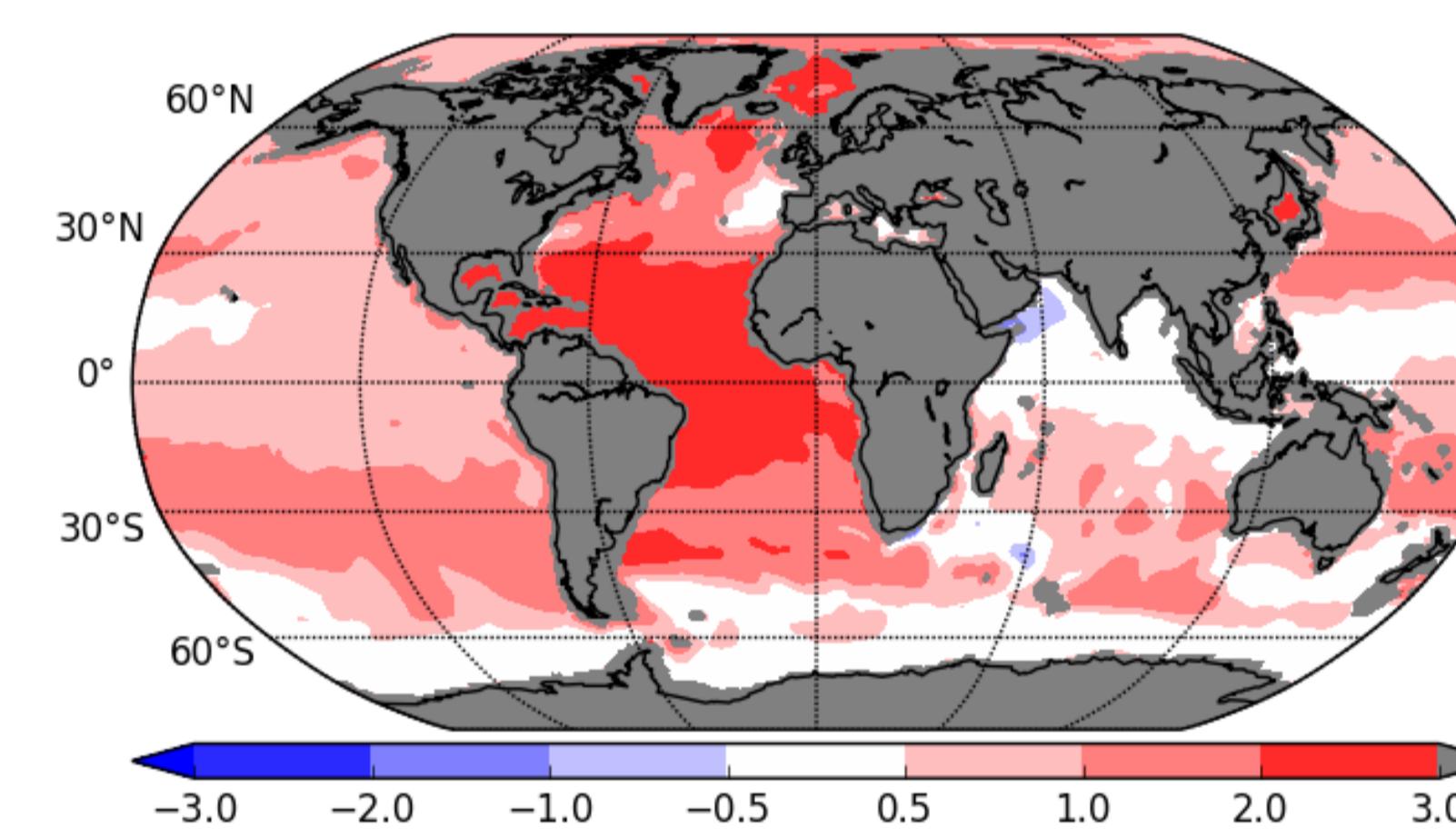


Fig. 2: Potential temperature bias for years 1971–2000 of the historical simulations from 13 CMIP5 models

Strong warm bias in 1000 m depth especially in Atlantic

Example application: error growth

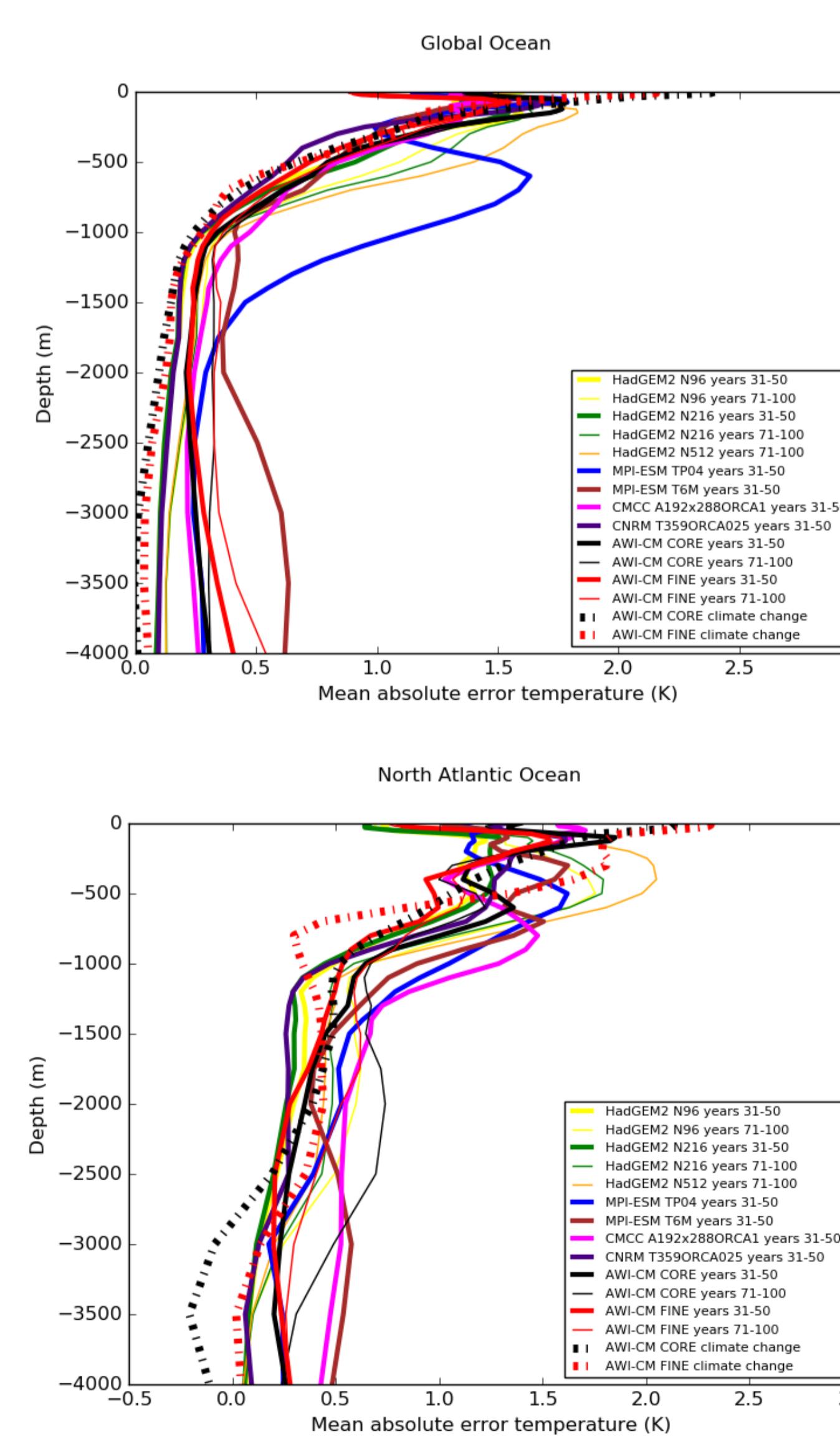


Fig. 3: Profiles of potential temperature mean absolute error averaged over 31–50 years and over 71–100 years after initialization with PHC climatology from different PRIMAVERA HighResMIP prototype simulations.

Above: global ocean
Below: North Atlantic ocean.

- Error growth from years 31–50 to years 71–100 visible
- Already after such a short time from initialization the error is often larger than climate change signal

References for AWI-CM:

- Sidorenko, D., Rackow, T. et al. (2015): Towards multi-resolution global climate modeling with ECHAM6-FESOM. Part I: model formulation and mean climate. *Clim. Dyn.* Vol. 44, Issue 3, pp 757–780, doi: 10.1007/s00382-014-2290-6
- Rackow, T. et al. (2016): Towards multi-resolution global climate modeling with ECHAM6-FESOM. Part II: climate variability. *Clim. Dyn.*, doi: 10.1007/s00382-016-3192-6