

Narrowing Uncertainties of Climate Projections using Data-driven Methods: the AMOC Case Study

Pierre Tandeo, Pierre Le Bras, Florian Sévellec, Pierre Ailliot, Juan Ruiz

▶ To cite this version:

Pierre Tandeo, Pierre Le Bras, Florian Sévellec, Pierre Ailliot, Juan Ruiz. Narrowing Uncertainties of Climate Projections using Data-driven Methods: the AMOC Case Study. NHM 2023: 6th International Workshop on Nonhydrostatic Models, Aug 2023, Sapporo, Japan. pp.1-13. hal-04193133

HAL Id: hal-04193133 https://imt-atlantique.hal.science/hal-04193133v1

Submitted on 4 Sep 2023

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.











Narrowing Uncertainties of Climate Projections using Data-driven Methods The AMOC case study

Pierre Tandeo¹, Pierre Le Bras^{1,2}, Florian Sévellec², Pierre Ailliot³, and Juan Ruiz⁴

The 6th International Workshop on Nonhydrostatic Models (NHM-WS 2023)

31 August - 2 September 2023

Hokkaido University Conference Hall, Sapporo, Japan

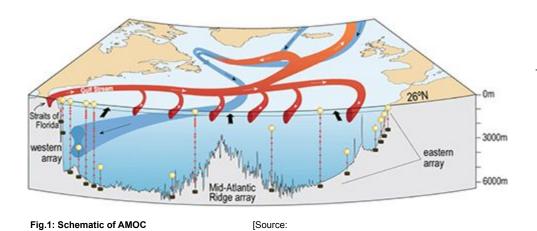
¹ Lab-STICC, IMT Atlantique, FRANCE

² Laboratoire d'Océanographie Physique et Spatiale, IUEM, FRANCE

³ Laboratoire de Mathématiques de Bretagne Atlantique, UBO, FRANCE

⁴ Centro de Investigaciones del Mar y la Atmósfera, Univ. Buenos Aires, ARGENTINA

Context: Atlantic Meridional Overturning Circulation (AMOC)

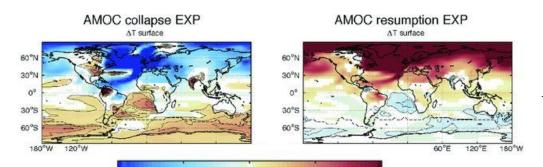


AMOC

→ Circulation in the Atlantic ocean

Shallow warm water northward

Deep cold water southward



http://www.noc.soton.ac.uk/rapidmoc/

→ Influence on climate of Atlantic regions

→ Essential to monitor its evolution

Fig.2: Experiences of AMOC collapse/resumption and influence on surface temperature anomalies [Source: Pedro et al., 2018]]

Context: AMOC observations and AMOC projections in climate models

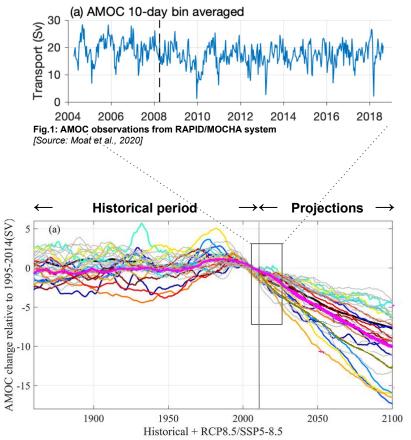


Fig.2: AMOC anomalies from different models (SSP5-8.5 in CMIP5-6) [Source: Lyu and Church, 2020]

Observations:

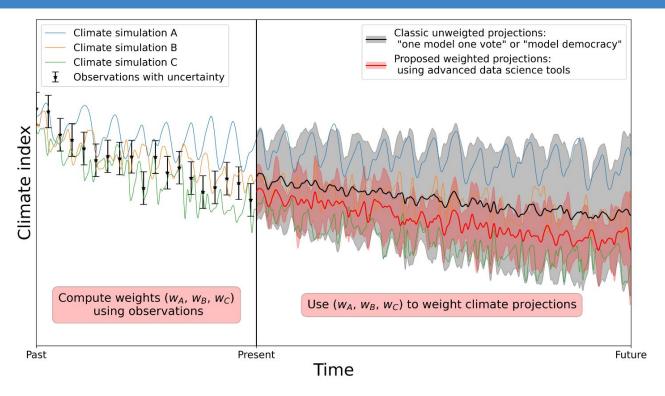
- → Seasonal and interannual variability
- → Hard to measure, lot of noise

CMIP Model projections:

- → Models predict severe decrease of the AMOC
- → Wide spread (model uncertainty)

How to improve accuracy and reliability of AMOC projections?

Aim of this study: compute weights to constraint climate projections



State of the art

- → Model democracy (same weights)
- → Questionable (e.g., Knutti et al., 2019)

Our goal → Weight the projections

- → <u>Using available simulations and observations</u>
- → Reduce the uncertainty

Proposed methodology: combination of data assimilation and machine learning

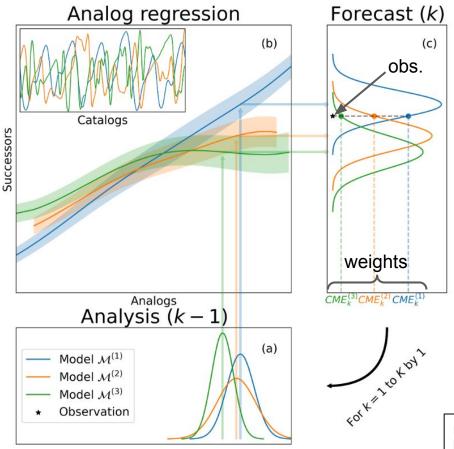


Fig.1: Scheme of the methodology

[Source: Ruiz et al., 2022]

Step (a): Data assimilation

- → start from an accurate initial condition
- → Ensemble Kalman filter (EnKF)

Step (b): Data-driven forecast

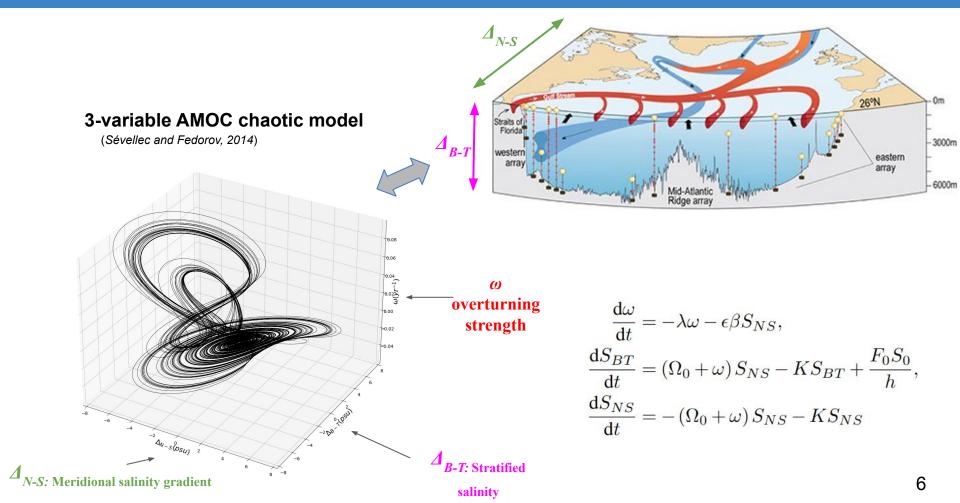
- → make forecasts without equations
- → analog method

Step (c): Weighting procedure

- → compute observation-forecast distance
- → contextual model evidence (CME)

Ruiz, J., Ailliot, P., Chau, T. T., Le Bras, P., Monbet, V., Sévellec, F., & Tandeo, P. (2022). Analog data assimilation for the selection of suitable general circulation models. *Geoscientific Model Development*, *15*(18), 7203-7220.

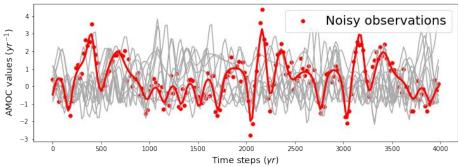
Experimental set up: idealized AMOC chaotic model



Experimental set up: generation of different versions of the idealized AMOC model

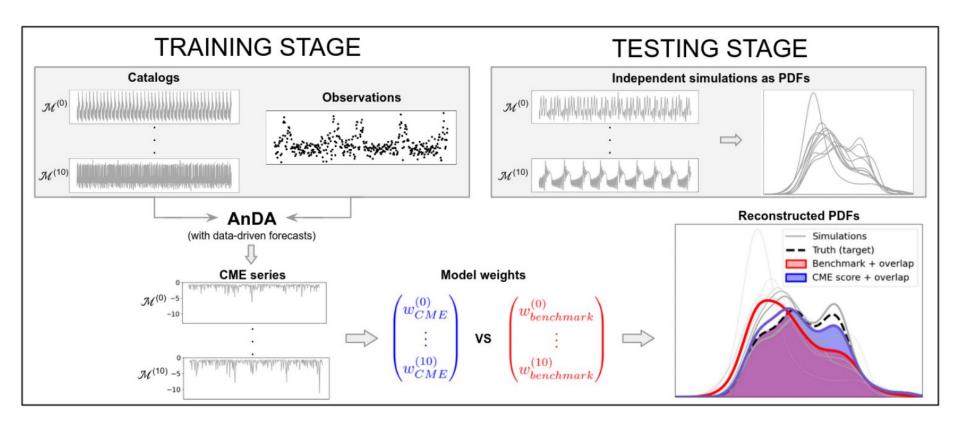


- → 11 perturbed-parameter versions
- → Leave-One-Out strategy
- → Mimic 10 climate simulations and 1 observation

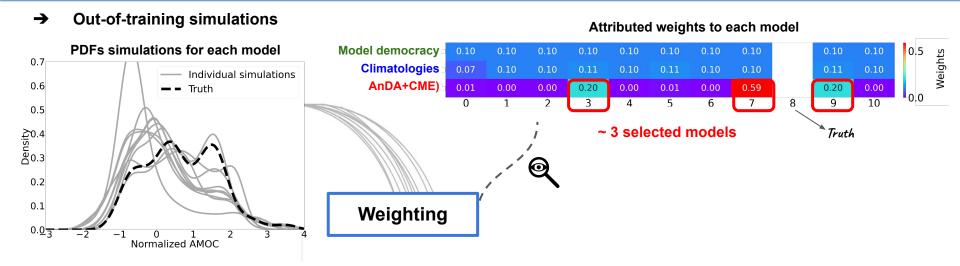


Schematic of the Leave-One-Out strategy

- → Compute 10 weights
- → Approximate observation distribution

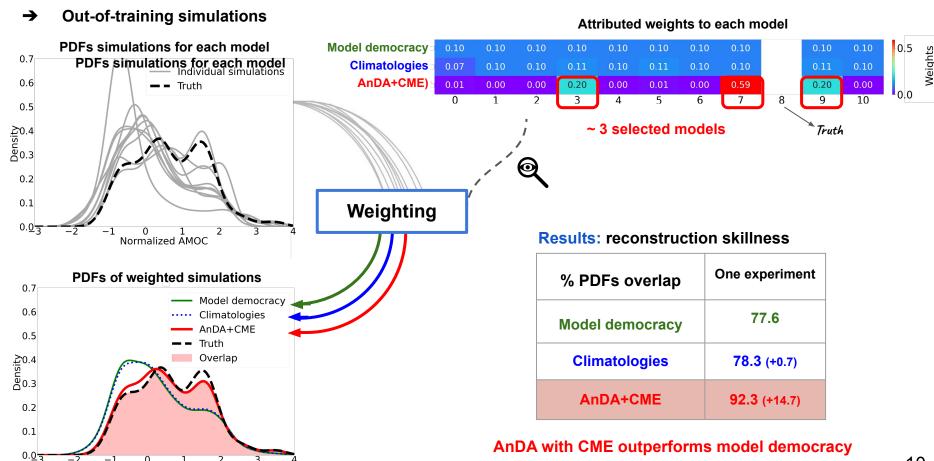


Application results for one experiment: applying the weights to reconstruct the truth



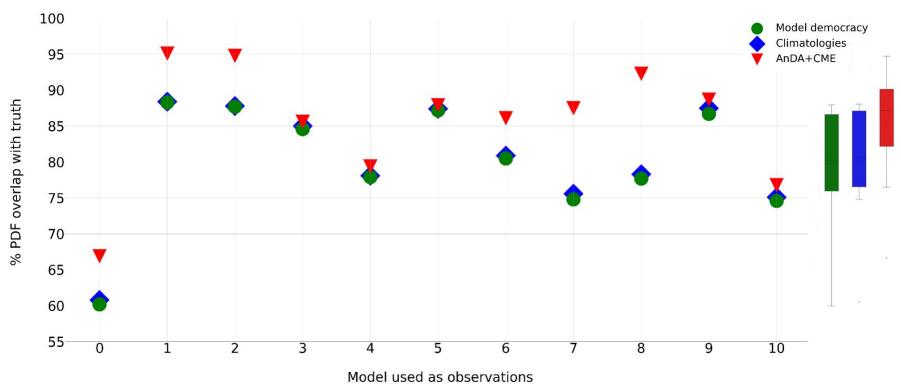
Application results for one experiment: applying the weights to reconstruct the truth

Normalized AMOC



Application results: overall results on Leave-One-Out (LOO) experiments

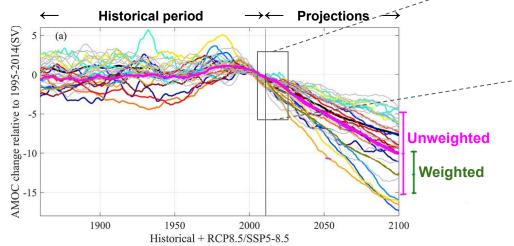


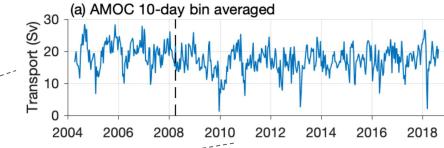


AnDA with CME performs at least as good as model democracy

Conclusion & perspectives

- → Weight climate simulations using noisy observations
- → Data-driven approach (no need to run climate models)
- → Combination of Data Assimilation & Machine Learning
- → Application to synthetic data of the AMOC
- → The proposed method outperforms "model democracy"





Reference



Geoscientific Model Development

🔐 ARTICLES & PREPRINTS 🔻 SUBMISSION POLICIES 🕶 PEER-REVIEW PROCESS 🔻 EDITORIAL BOARD ABOUT 🕶 EGU PUBLICATIONS 🖒

Article Articles / Volume 15, issue 18 / GMD, 15, 7203-7220, 2022 Search https://doi.org/10.5194/gmd-15-7203-2022 Download @ Author(s) 2022. This work is distributed under Article (6128 KB) Article Peer review Metrics Related articles Assets the Creative Commons Attribution 4.0 License. ▶ Full-text XML Methods for assessment of models | @() 26 Sep 2022 ▶ BIbTeX Analog data assimilation for the selection of suitable ▶ EndNote general circulation models **Short summary** We present a new Juan Ruiz ☑, Pierre Ailliot, Thi Tuyet Trang Chau, Pierre Le Bras, Valérie Monbet, Florian Sévellec, approach to validate and Pierre Tandeo numerical simulations of the current climate. The method... Abstract ▶ Read more

あいがとうこといき! arigato gozaimasu!

Data assimilation is a relevant framework to merge a dynamical model with noisy observations. When various models are in competition, the question is to find the model that best matches the observations. This matching can be measured by using the model evidence, defined by the likelihood of the observations given the model. This study explores the performance of model selection based on model evidence computed using data-driven data assimilation, where dynamical models are emulated using machine learning methods. In this work, the methodology is tested with the three-variable Lorenz model and with an intermediate complexity atmospheric general circulation model (a.k.a. the SPEEDY model). Numerical experiments show that the data-driven implementation of the model selection algorithm performs as well as the one that uses the dynamical model. The technique is able to select the best model among a set of possible models and also to characterize the spatiotemporal variability of the model sensitivity. Moreover, the technique is able to detect differences among models in terms of local dynamics in both time and space which are not reflected in the first two moments of the climatological probability distribution. This suggests the implementation of this technique using available long-term observations and model simulations.



ModE-Sim – a medium-sized atmospheric... Hand et al.



MESMAR v1: a new regional coupled climate... Storto et al.

Contact:

https://tandeo.wordpress.com/pierre.tandeo@imt-atlantique.fr