

Climate-relevant imprints and observational implications of the chaotic/intrinsic ocean variability: lessons from the OCCIPUT oceanic Ensemble.

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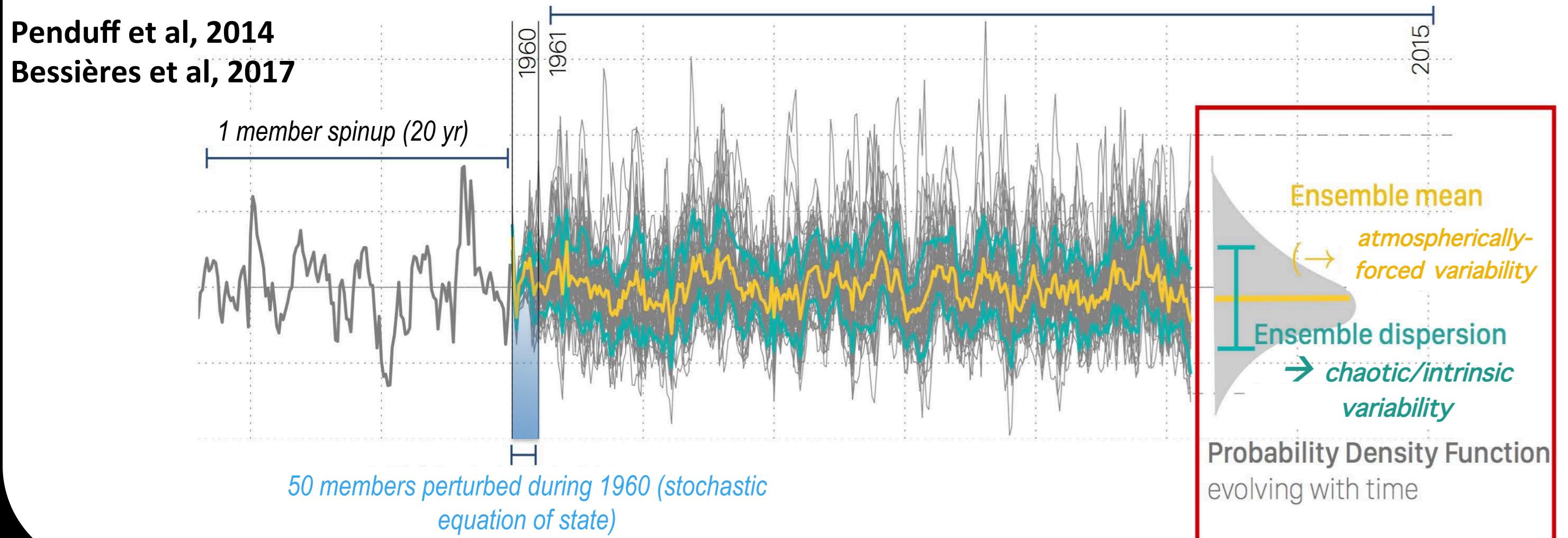
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The OCCIPUT Ensemble: 50 turbulent ocean hindcasts

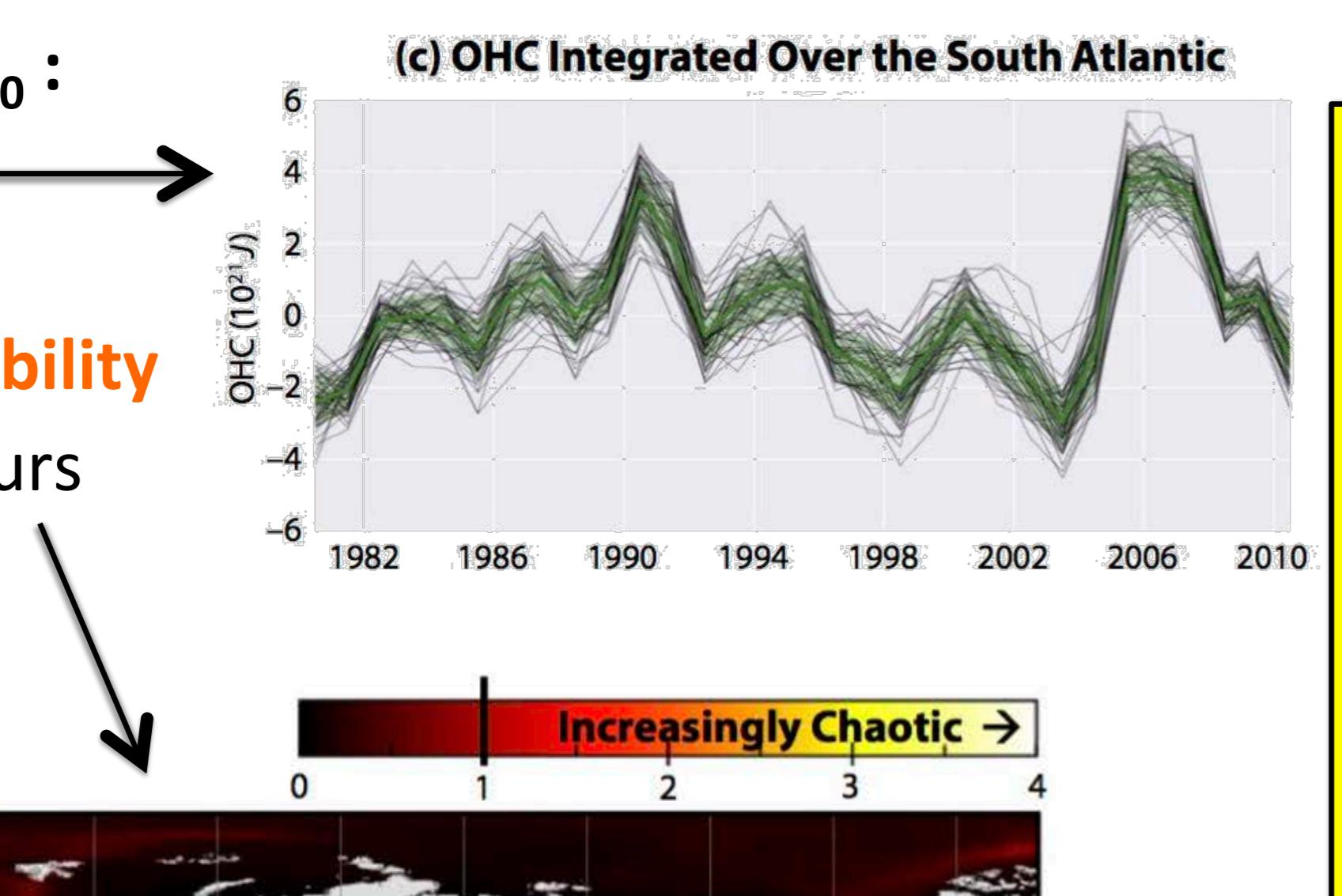
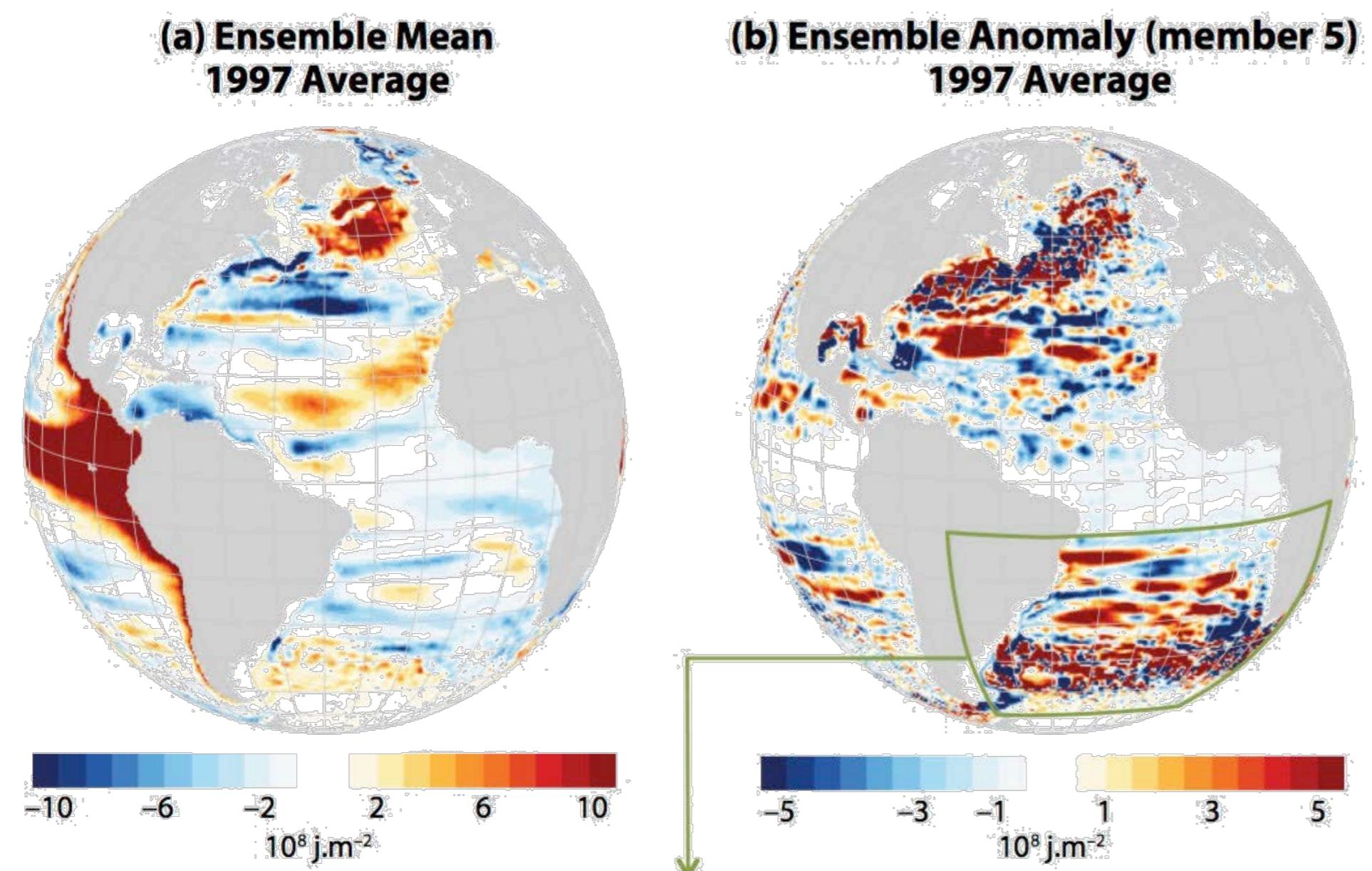
- NEMO global ocean model
- Resolution $\frac{1}{4}^\circ$ — 1960-2015
- 50 members with :
 - Same ERA-Interim atm. forcing
 - Slight initial perturbations

Estimate the oceanic :

- Atmospherically-forced variability ($\sigma_F = \text{STD of ensemble mean}$)
- Chaotic/intrinsic variability ($\sigma_C = \text{mean of ensemble STD}$)



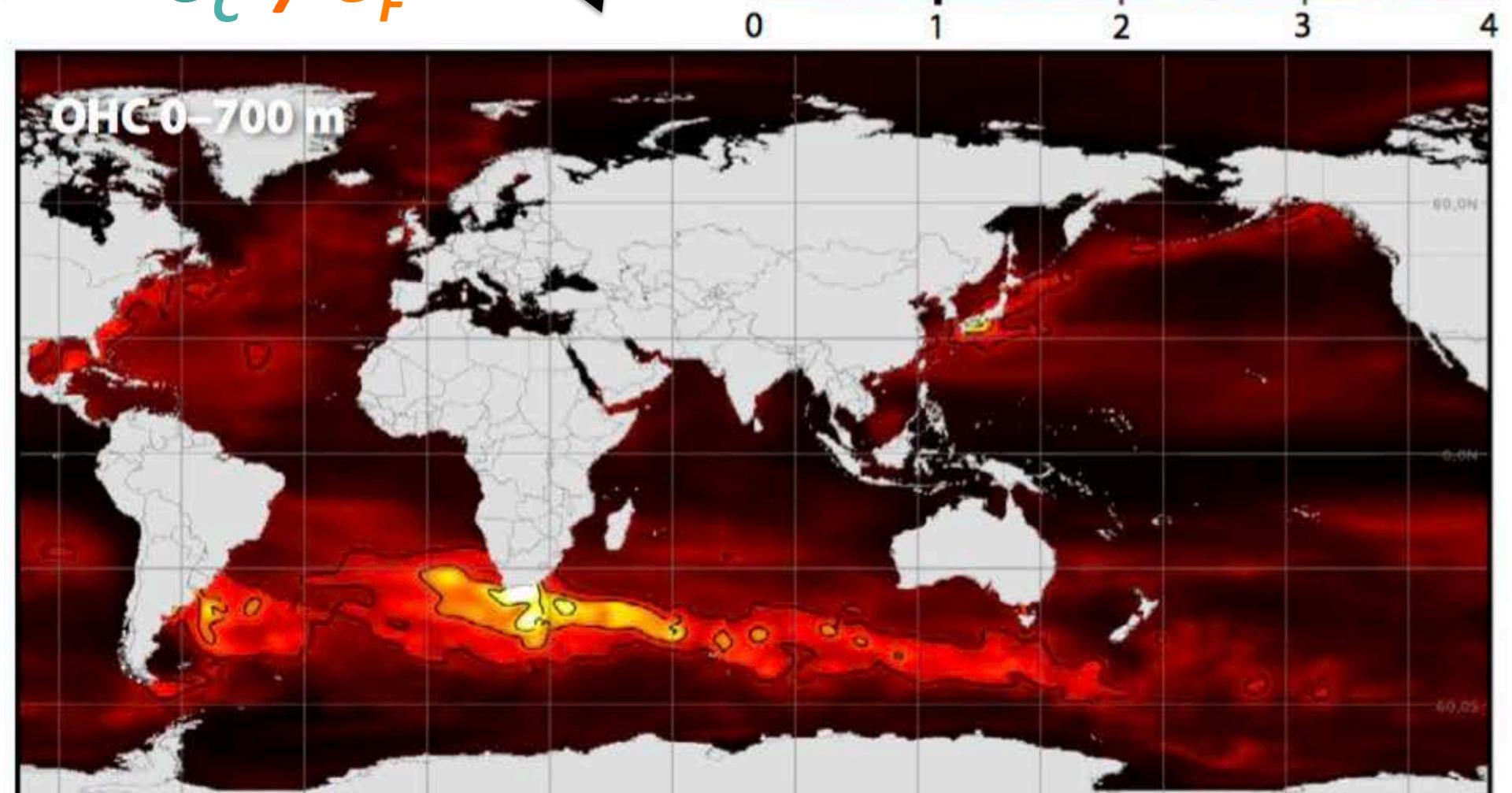
Interannual variability of Ocean Heat Content (0-700m)



LFCIV imprint on OHC_{0-700} :

- up to basin scales
- dominates **Forced variability** in regions within contours

This strong 1-30yr chaotic variability of OHC & AMOC (almost absent without eddies) is expected to impact the atmosphere in coupled models with eddying oceans

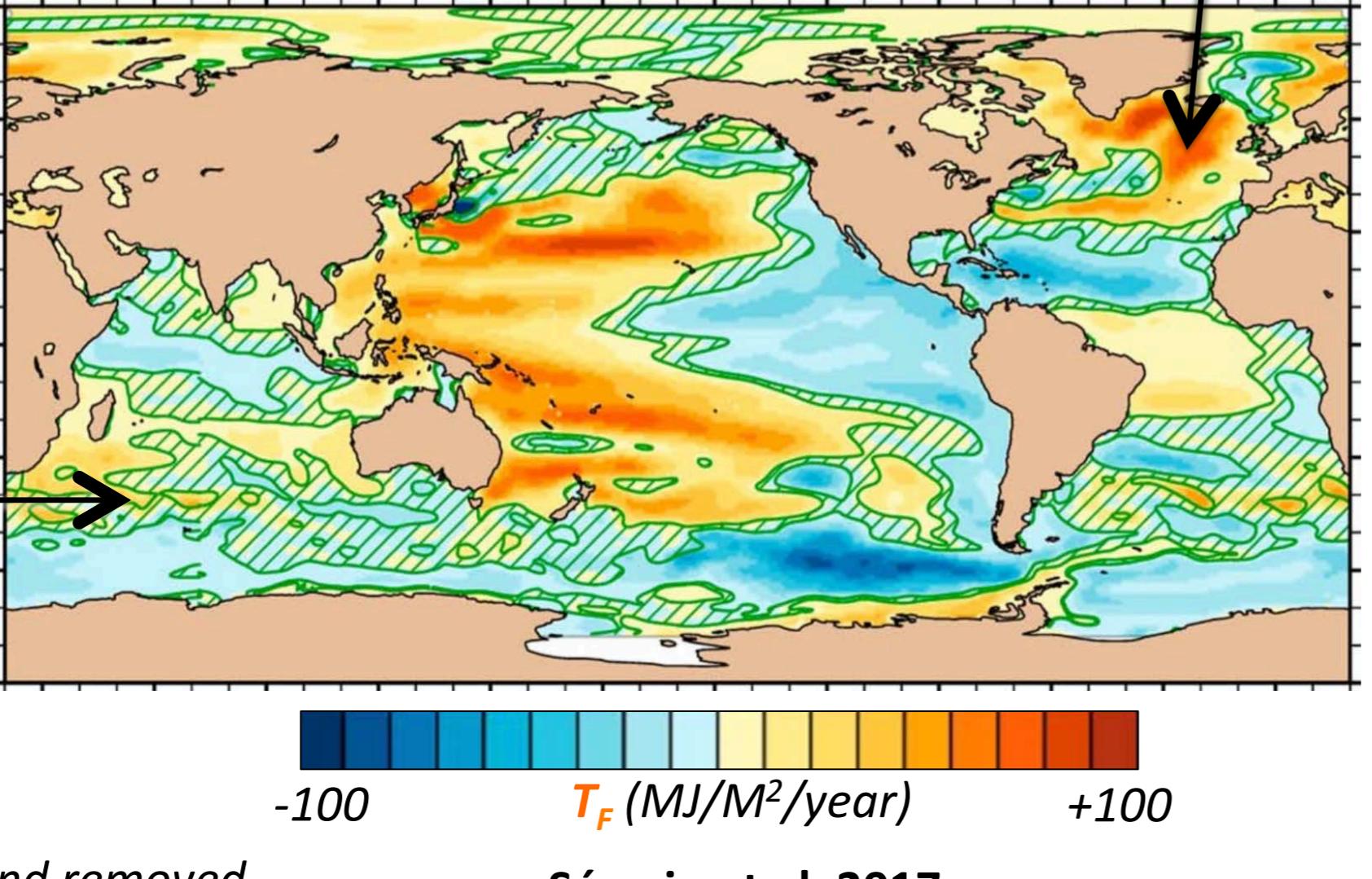


Sérazin et al, 2017
Penduff et al, 2018

21-year trends* (1980-2010) of Ocean Heat Content

- [atmospheric + anthropogenic drivers] → **Forced** OHC trends T_F (colors)
- LFCIV → **Random** OHC trends T_R

In regions where $|T_F| < 2 \cdot T_R$ (green shading) OHC trends can NOT be surely attributed (95% confidence) to [atmospheric + anthropogenic drivers]



(*) corrected for model drift & global mean trend removed

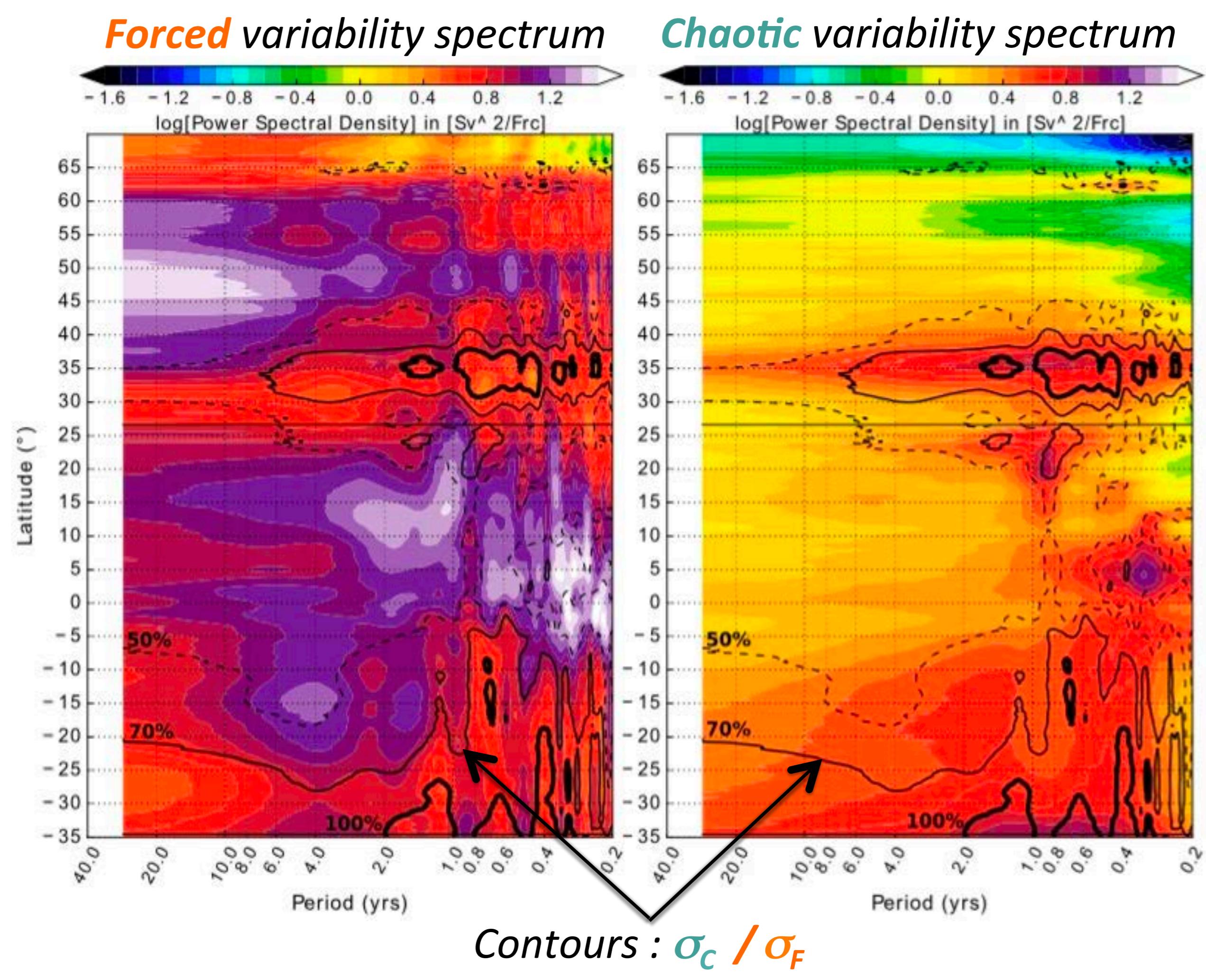
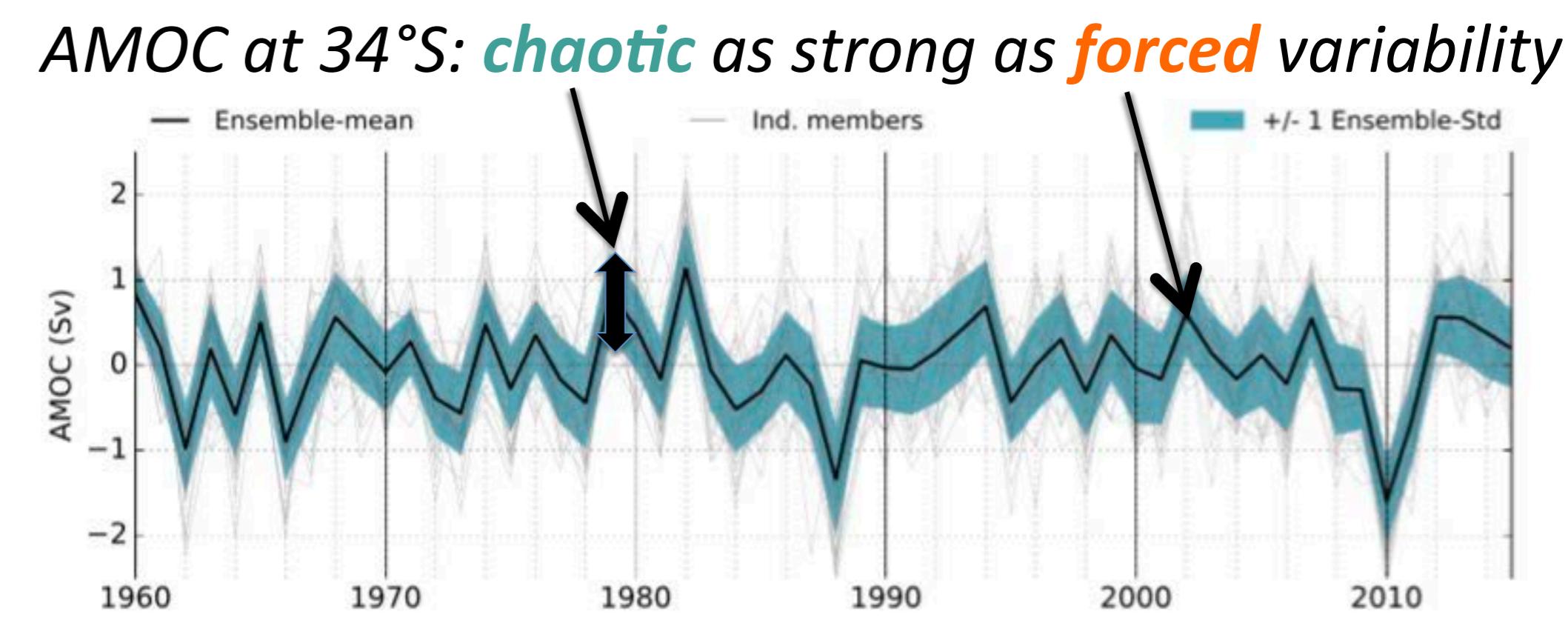
References

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Take-home messages

- In the presence of mesoscale, the ocean spontaneously generates a strong **low-frequency chaotic/intrinsic variability (LFCIV)**. It locally competes with the **atmospheric forcing** in driving the **interannual-to-multidecadal variability of key ocean climate indices (OHC_{0-700m} , AMOC, MHT, etc)**.
- Much weaker or absent in coarse-resolution ocean models used in most IPCC-class climate simulators, **this strong LFCIV**, might impact the **atmosphere in coupled models with turbulent oceans**.
- Over large regions, the **LFCIV** adds a **random component** to local **20/30-year trends of OHC and sea level**, hindering their unambiguous attribution to **[atmospheric+anthropogenic] drivers**.
- Ensemble model statistics can be used to **attenuate the signature of LFCIV in observational datasets** (via filters or Machine Learning), and **unveil the deterministic response of the real ocean to the atmosphere**.

Interannual variability of the Atlantic Meridional Overturning Circulation



Leroux et al, 2018

23-year trends (1993-2015) of Observed Sea Level

- In many regions, $\sigma_C > \sigma_F$ on space scales [110-800km], over all time scales
- Filtering out scales [110-800km] → **estimation of Forced observed trends**
- Currently developing an estimator based on Machine Learning (CNN)

