

The rate of information transfer as a measure of ocean-atmosphere interactions



David Docquier

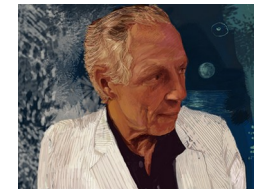
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EGU General Assembly 2023

Vienna, 24 April 2023



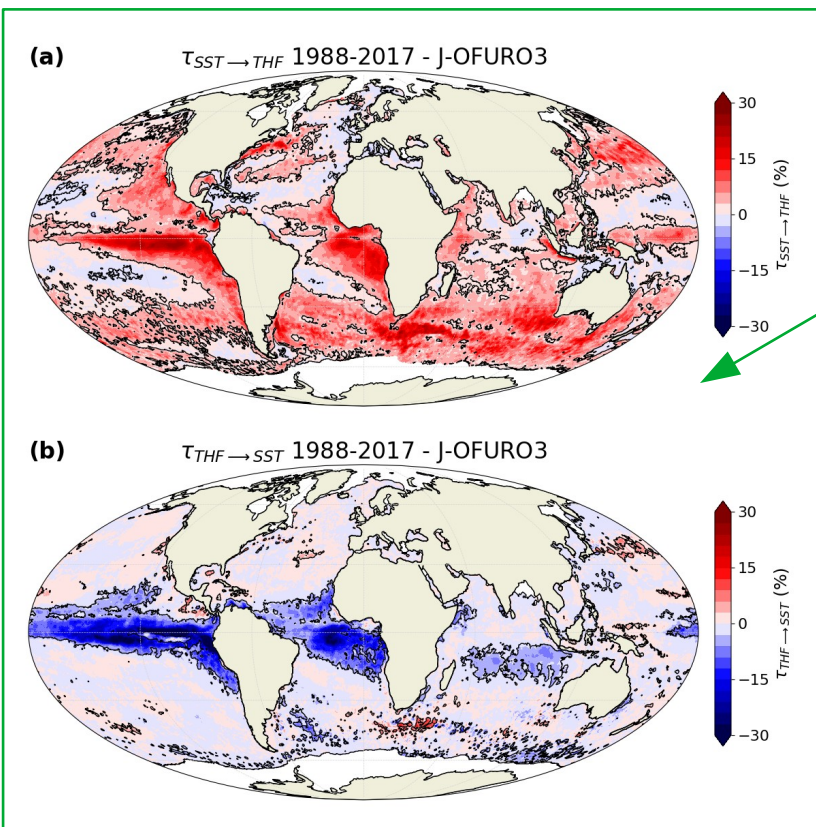
1st case study:

SST - Turbulent heat flux
Satellite data

Key results

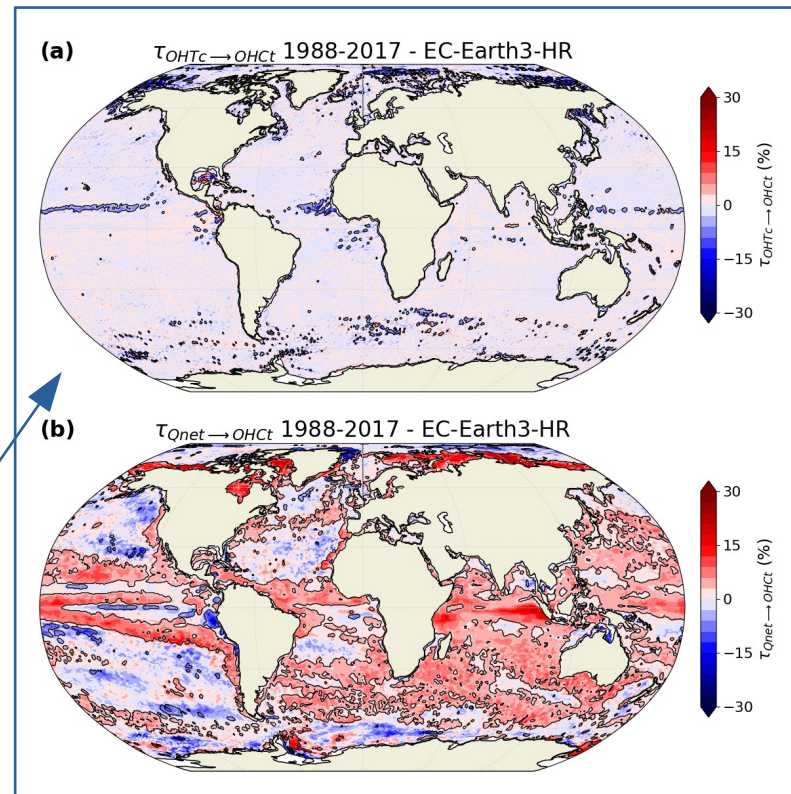
2nd case study:

Ocean heat budget upper 50m
Climate models



1. Stronger ocean influence (vs. atmosphere) in many regions of the world based on satellite data

2. Lower number of regions with a significant ocean dynamical influence in high-resolution models vs. low-resolution models



Docquier et al. (in review), Interactions between ocean heat budget terms in HighResMIP climate models measured by the rate of information transfer, *EGUphere*

Docquier et al. (accepted), The rate of information transfer as a measure of ocean-atmosphere interactions, *Earth System Dynamics*

Liang-Kleeman rate of information transfer

- Correlation (even if lagged) does not necessarily imply causation!
- Causality measured by the **rate of information** flowing from X_j to X_i :

(Liang, 2021)

$$T_{j \rightarrow i} = \frac{1}{\det \mathbf{C}} \cdot \sum_{k=1}^d \Delta_{jk} C_{k,di} \cdot \frac{C_{ij}}{C_{ii}}$$

Annotations for the equation above:

- $\det \mathbf{C}$: covariance matrix
- $\sum_{k=1}^d$: number of variables
- Δ_{jk} : cofactors of \mathbf{C}
- $C_{k,di}$: covariance between X_k and $\mathbf{dX}_i / \mathbf{dt} = (X_{i,t+1} - X_{i,t}) / \Delta t$
- C_{ij} : covariance between X_i and X_j
- C_{ii} : variance of X_i

Normalization:

- If $|\tau_{j \rightarrow i}| = 0\%$: X_j does not influence X_i
- If $|\tau_{j \rightarrow i}| > 0\%$: X_j influences X_i
- If $\tau_{j \rightarrow i} > 0\%$: X_j causes X_i to be more uncertain
- If $\tau_{j \rightarrow i} < 0\%$: X_j causes X_i to be more certain

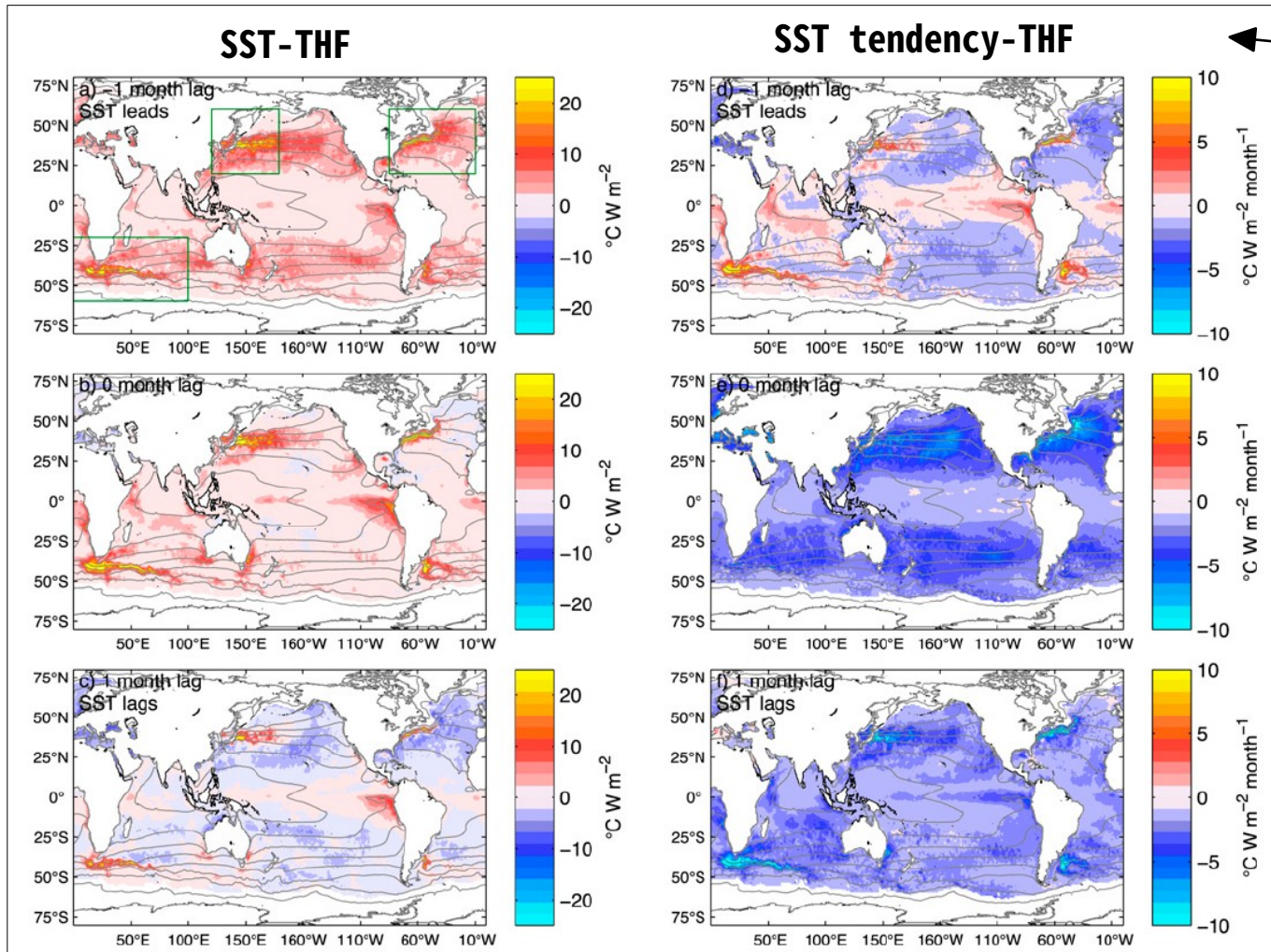
relative rate of information transfer from X_j to X_i

$$\tau_{j \rightarrow i} = \frac{T_{j \rightarrow i}}{Z_i}$$

$$Z_i = \sum_{k=1}^d |T_{k \rightarrow i}| + \left| \frac{dH_i^{noise}}{dt} \right|$$

noise term 3

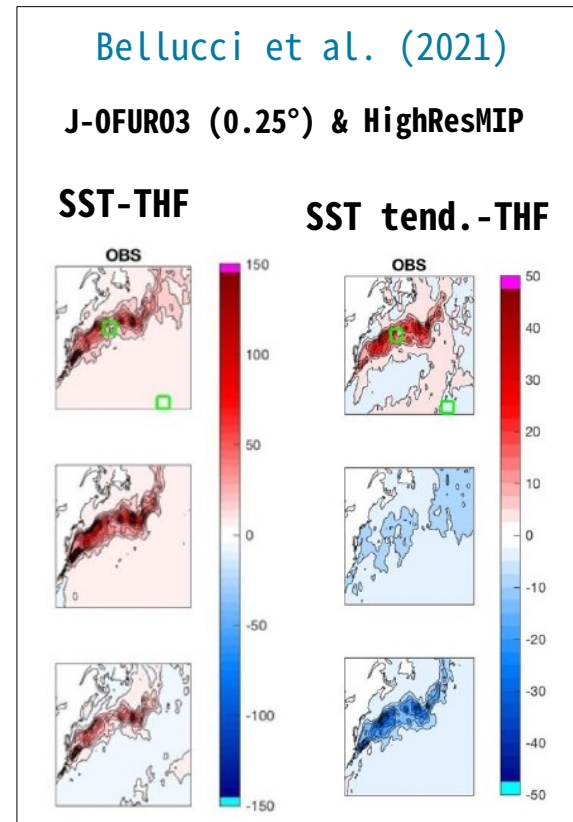
1st case study: SST – Turbulent heat flux (THF)



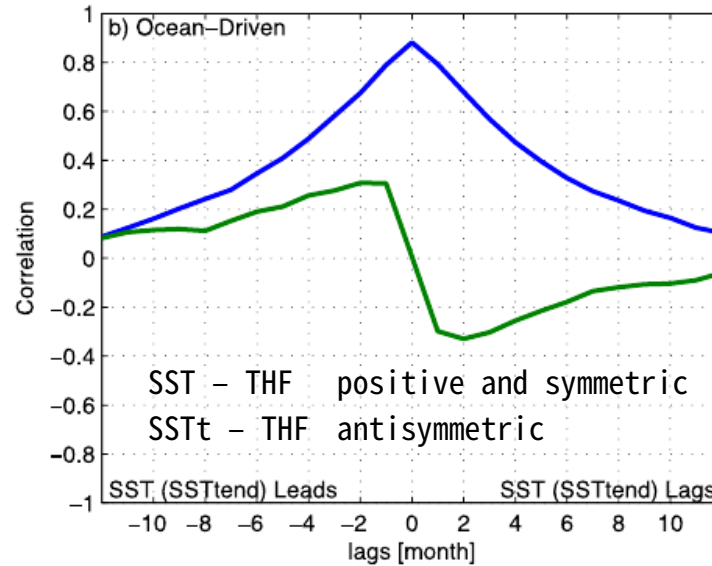
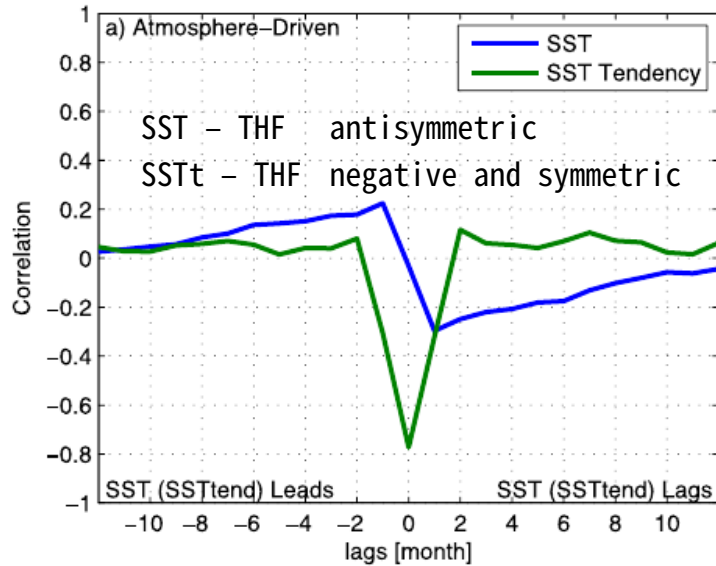
Bishop et al. (2017)
OI SST (0.25°) & OAFlux (1°)

Bellucci et al. (2021)

J-OFURO3 (0.25°) & HighResMIP



1st case study: Background & Methods



Bishop et al. (2017);
based on energy
balance model from
Wu et al. (2006)

- 3 variables: SST, SST tendency, turbulent heat flux (THF = latent + sensible)
- J-OFUR03 satellite observations (0.25° resolution)
- Monthly data over 1988-2017
- Confirmed with SeaFlux satellite observations

Docquier et al. (accepted)

SST – THF

(taking SST tendency into account)

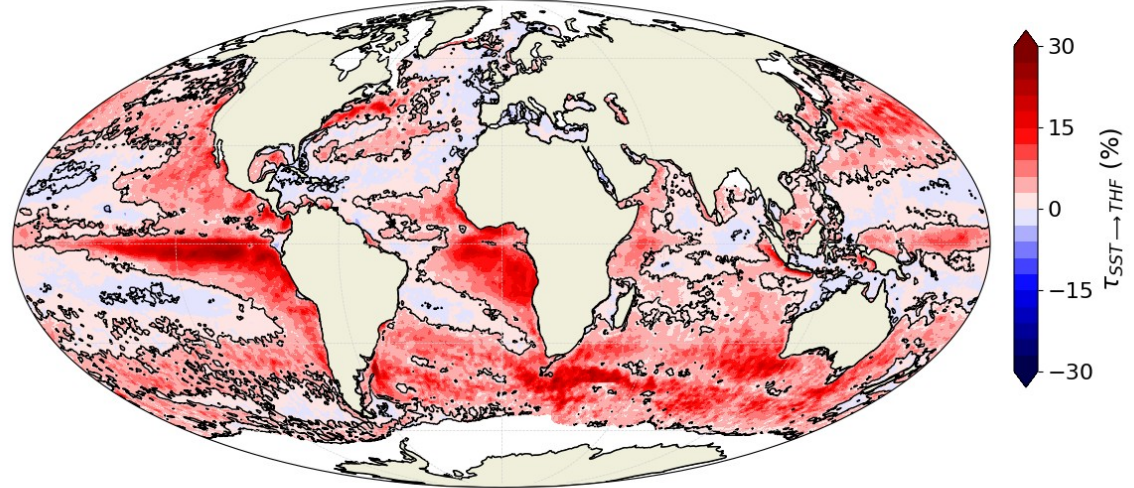
- **SST → THF:**

- Significant in a large number of regions
- > 0 : SST variability increases THF variability

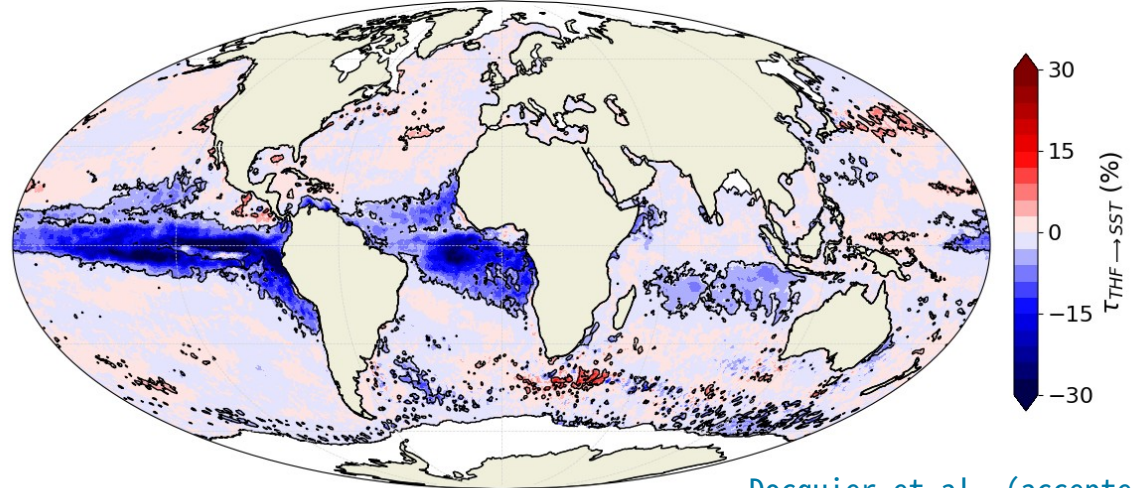
- **THF → SST:**

- Lower number of regions, including equatorial regions
- < 0 : THF variability decreases SST variability

(a) $\tau_{SST \rightarrow THF}$ 1988-2017 - J-OFURO3



(b) $\tau_{THF \rightarrow SST}$ 1988-2017 - J-OFURO3



SST tendency – THF (taking SST into account)

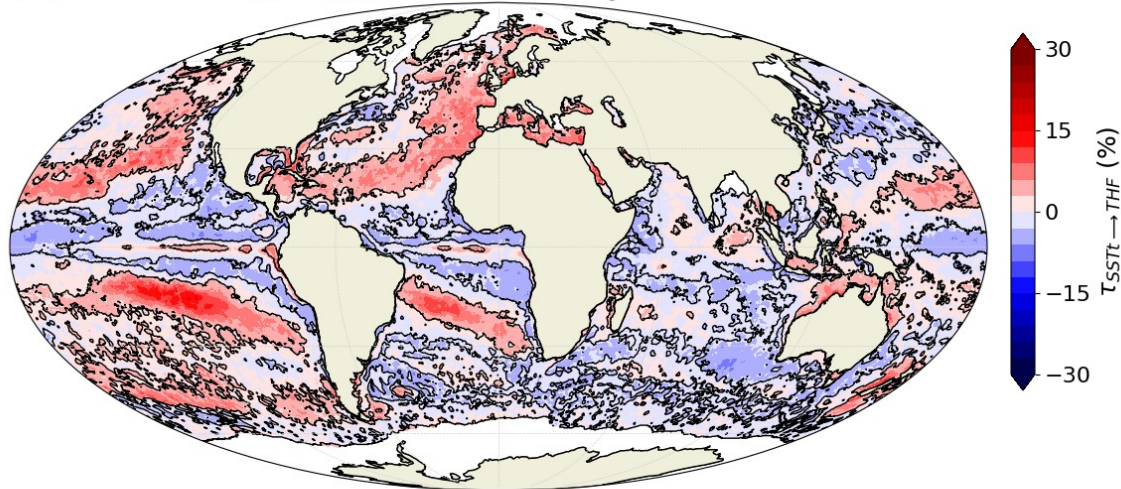
- **SSTt → THF:** →

- Significant in a large number of regions
- > 0 and < 0

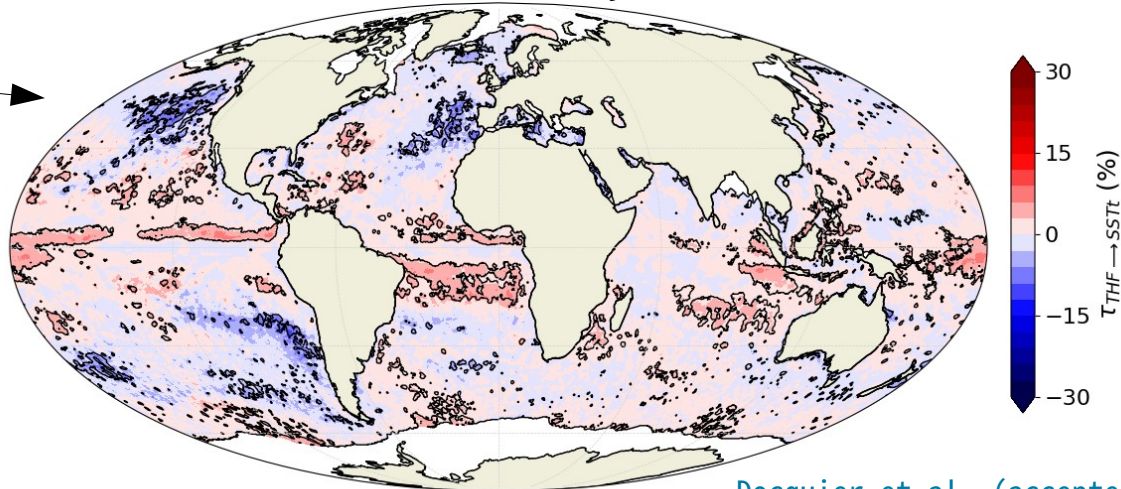
- **THF → SSTt:** →

- Lower number of regions, including North Atlantic, northeastern Pacific, and tropical regions
- > 0 and < 0

(a) $\tau_{SSTt \rightarrow THF}$ 1988-2017 - J-OFURO3



(b) $\tau_{THF \rightarrow SSTt}$ 1988-2017 - J-OFURO3

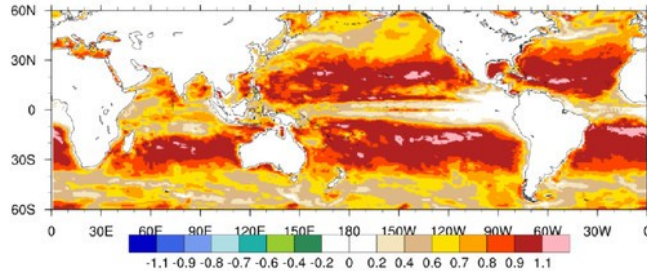


2nd case study: Ocean heat budget upper 50m

Thermodynamical contribution

VDIFF: vertical diffusion including surface heat flux

a) LOW-RES: Tendency & VDIFF

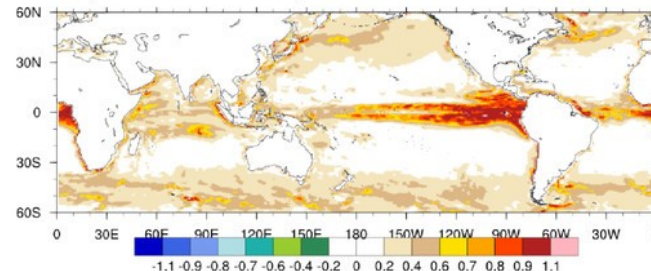


Low
resolution

Dynamical contribution

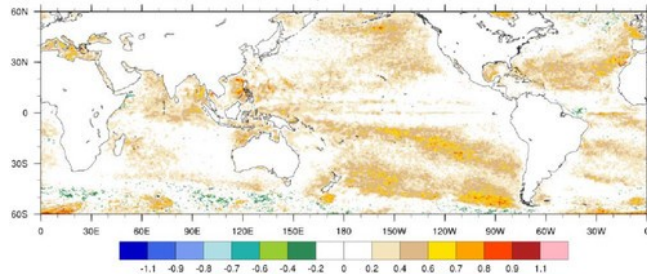
OHFC: ocean heat flux convergence

b) LOW-RES : Tendency & OHFC



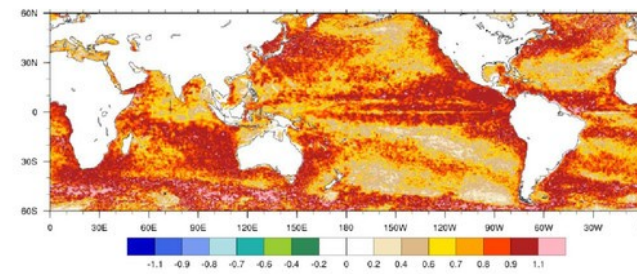
Small et al. (2020)

c) HI-RES: Tendency & VDIFF



High
resolution

d) HI-RES: Tendency & OHFC



- 3 variables: ocean heat content (OHC) tendency, ocean heat transport (OHT) convergence, net surface heat flux (Q_{net})
- HighResMIP climate models (1° vs. 0.25° in the ocean)
- Monthly data over 1988-2017

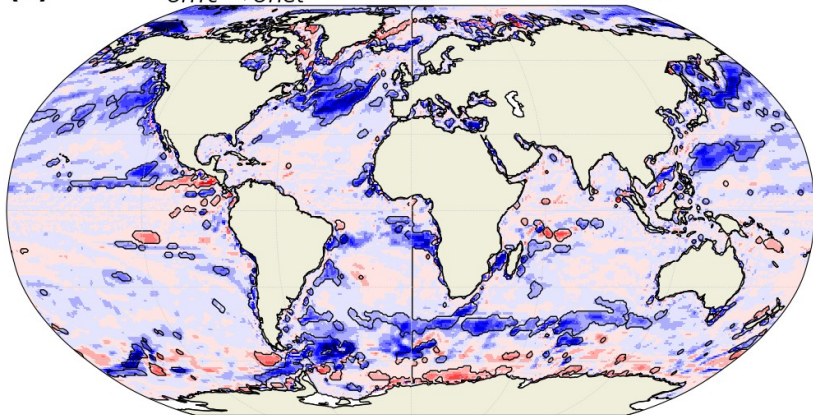
Docquier et al. (in review)

Low resolution

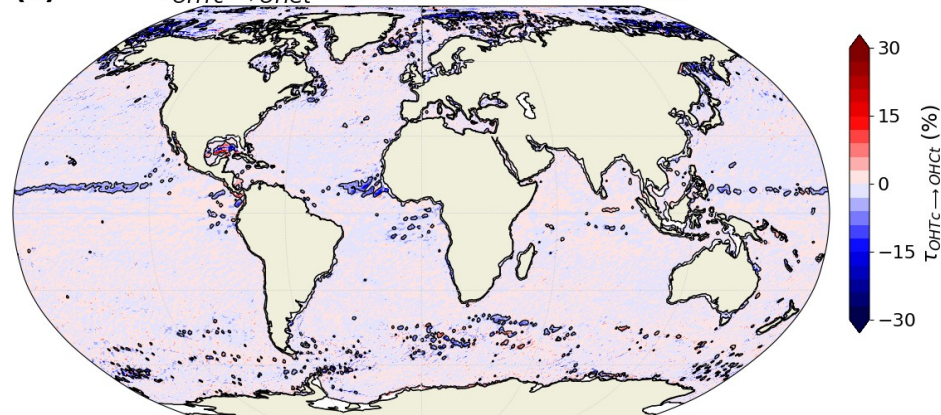
High resolution

Dynamical
influence

(a) $\tau_{OHTc \rightarrow OHct}$ 1988-2017 - EC-Earth3-LR

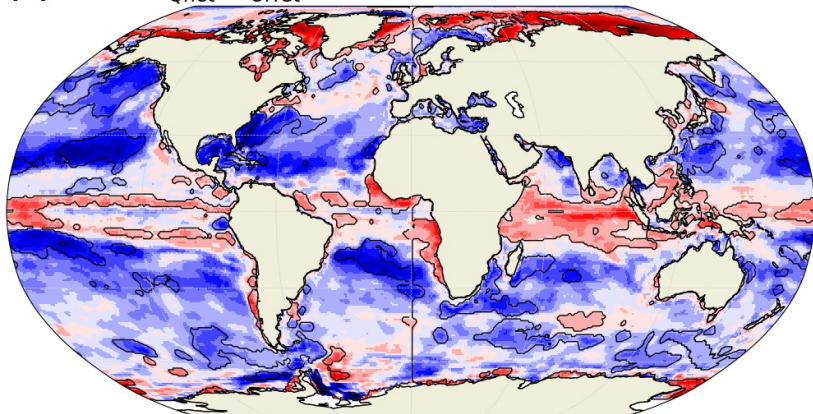


(a) $\tau_{OHTc \rightarrow OHct}$ 1988-2017 - EC-Earth3-HR

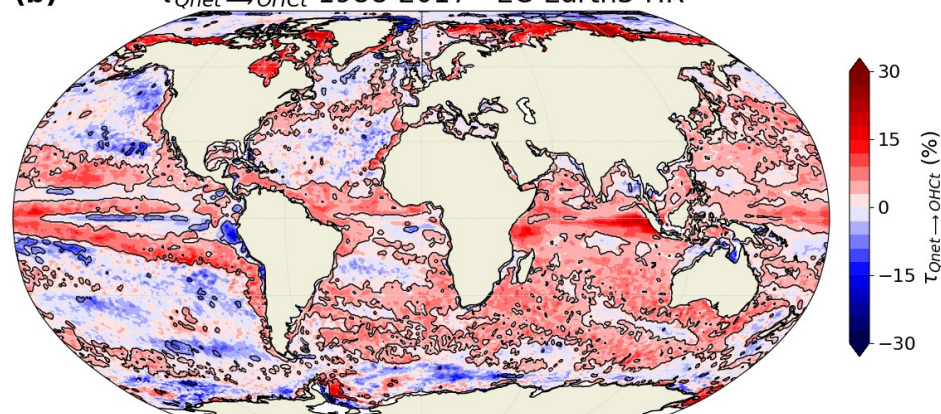


Thermodynamical
influence

(b) $\tau_{Qnet \rightarrow OHct}$ 1988-2017 - EC-Earth3-LR



(b) $\tau_{Qnet \rightarrow OHct}$ 1988-2017 - EC-Earth3-HR

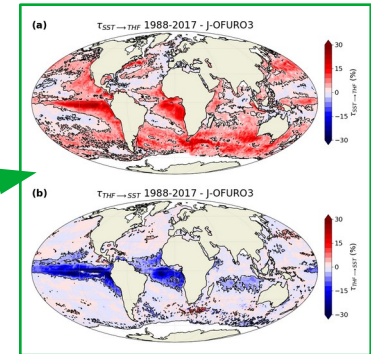


Conclusions

- The **Liang-Kleeman rate of information transfer** allows to quantify the directional dependence between variables and goes beyond correlation analyses

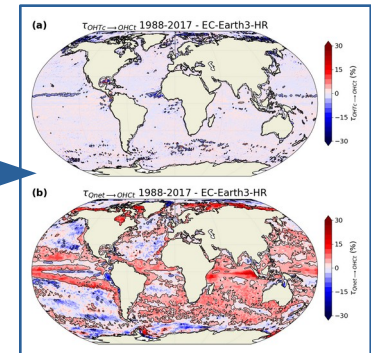
- **1st case study: Larger number of regions with a significant ocean → atmosphere influence vs. atmosphere → ocean influence based on satellite data**

Docquier et al. (accepted), The rate of information transfer as a measure of ocean-atmosphere interactions, Earth System Dynamics



- **2nd case study: Lower number of regions with a significant ocean dynamical influence at high resolution vs. low resolution based on global climate models**

Docquier et al. (in review), Interactions between ocean heat budget terms in HighResMIP climate models measured by the rate of information transfer, EGU sphere



- Comparison to other causal methods is ongoing