

The Rate of Information Transfer as a Measure of Rapid Changes in Arctic Sea Ice

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November 26, 2022

Abstract

Arctic sea ice has substantially changed over the past four decades, with a large decrease in sea-ice area and volume. The exact causes of these changes are not entirely known. In our study, we make use of the Swedish Meteorological and Hydrological Institute Large Ensemble (SMHI-LENS). This ensemble consists of 50 members realized with the EC-Earth3 global climate model and covers the period 1970-2100. We apply the Liang-Kleeman information flow method to analyze the cause-effect relationships between Arctic sea ice and its potential drivers. We show that recent and future changes in Arctic sea ice are mainly driven by air and sea-surface temperatures and ocean heat transport. Conversely, changes in Arctic sea ice also considerably impact temperature and ocean heat transport. Finally, we find a progressive decrease in the influence of sea-ice area and volume on air temperature and ocean heat transport through the twenty-first century.

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SMHI The Rate of Information Transfer as a Measure of Rapid Changes in Arctic Sea Ice
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Summary

- The Long-Khramov rate of information transfer allows to quantify the directional dependence between Arctic sea ice and its drivers.
- Recent and future changes in Arctic sea ice are mostly driven by air and sea surface temperatures, and ocean heat transport.
- The influence of Arctic sea ice on air temperature and ocean

Methodology

Rate of information transfer

- Quantified by the rate of information flowing from variable X to variable Y [Equation 1]

$$T_{Y \leftarrow X} = \frac{1}{\det \mathbf{C}} \cdot \sum_{k=1}^d \Delta_{jk} C_{k,d+1} \cdot \frac{C_{d+1}}{C_{ii}}$$

where $\mathbf{C}$ is the covariance matrix of all the variables of variables, Δ_{jk} is the volume of $\mathbf{C}_{\Delta_{jk}}$ the sample covariance between X_j and the rate between X and Y is the sample covariance of X_j and Y .

Results in figures

Results explained

Member analysis

- Rate of information transfer computed for each member separately
- Member analysis shows that the rate of information transfer is mostly driven by SST and SST (Fig. 1a)
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Why is it important?

- Arctic sea ice area and volume have decreased by ~2 million km² and ~15,000 km³ since 1979, respectively.
- Model projections show a continuation of this process, which depends on the emission scenario.
- Even drivers of these changes are increasingly known.
- Influence of sea ice on climate is usually quantified via radiative and transport feedbacks.
- Global reference frameworks advance quantify sea-ice loss between models.
- In our study, we use the Long-Khramov information flow method to quantify the influence of potential climate drivers on Arctic sea ice area and volume, as well as the oceanic impact of sea ice area and volume on these climate drivers.
- Our understanding of climate processes in polar regions (and in other latitudes) could greatly benefit from using the information flow method.

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Accept

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SUMMARY

- The Liang-Kleeman rate of information transfer (see Methodology) allows to quantify the directional dependence between Arctic sea ice and its drivers
- Recent and future changes in Arctic sea ice are mainly driven by air and sea-surface temperatures and ocean heat transport (see Results)
- The influence of Arctic sea ice on air temperature and ocean heat transport progressively decreases through the twenty-first century (see Results)

This research is supported by the JPI-Climate / JPI-Oceans ROADMAP (<http://www.jpi-climate.eu/joint-activities/joint-calls/CPILoud/ROADMAP>) project and has received funding from the Belgian Federal Science Policy Office (BELSPO, contract B2/20E/P1/ROADMAP).

METHODOLOGY

1. Rate of information transfer

- Causality measured by rate of information flowing from variable X_j to variable X_i (Liang, 2021 (<https://www.mdpi.com/1099-4300/23/6/679>)):

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

where $\mathbf{C}$ is the covariance matrix, d is the number of variables, Δ_{jk} are the cofactors of $\mathbf{C}$, $C_{k,di}$ is the sample covariance between all X_k and dX_i / dt (Euler forward difference approximation), C_{ij} is the sample covariance between X_i and X_j , C_{ii} is the sample variance of X_i .

- Normalization:

$$\tau_{j \rightarrow i} = \frac{T_{j \rightarrow i}}{Z},$$

where Z is the normalizer, which takes into account influences from all variables on variable X_i as well as the effect of noise.

- If $|\tau_{j \rightarrow i}| = 100\%$, X_j has the maximum influence on X_i
- If $|\tau_{j \rightarrow i}| = 0\%$, X_j has no influence on X_i
- Statistical significance computed via bootstrap resampling (95% confidence interval)

2. Climate model data

- SMHI-LENS (Swedish Meteorological and Hydrological Institute Large Ensemble): 50 members run with the global climate model EC-Earth3 (Wyser et al., 2021 (<https://gmd.copernicus.org/articles/14/4781/2021/>))
- Model components: IFS cy36r4 (atmosphere; ~ 80 km), NEMO3.6/LIM3 (ocean/sea ice; $\sim 1^\circ$)
- 1970-2014: CMIP6 forcing
- 2015-2100: SSP1-1.9 and SSP5-8.5 (**results shown here are for SSP5-8.5**)
- Variables: March / September Arctic sea-ice area (SIA), March / September Arctic sea-ice volume (SIV), Arctic near-surface air temperature (T_{2m} ; annual mean), Arctic sea-surface temperature (SST; annual mean), total Arctic Ocean heat transport (OHT_A ; annual mean), ocean and atmospheric heat transports at $70^\circ N$ (OHT_{70N} , AHT_{70N} ; annual mean), winter Arctic Oscillation Index (AOI)

More details: Docquier et al. (in review), preprint available on ESSOAr (<https://doi.org/10.1002/essoar.10507846.1>)

RESULTS IN FIGURES

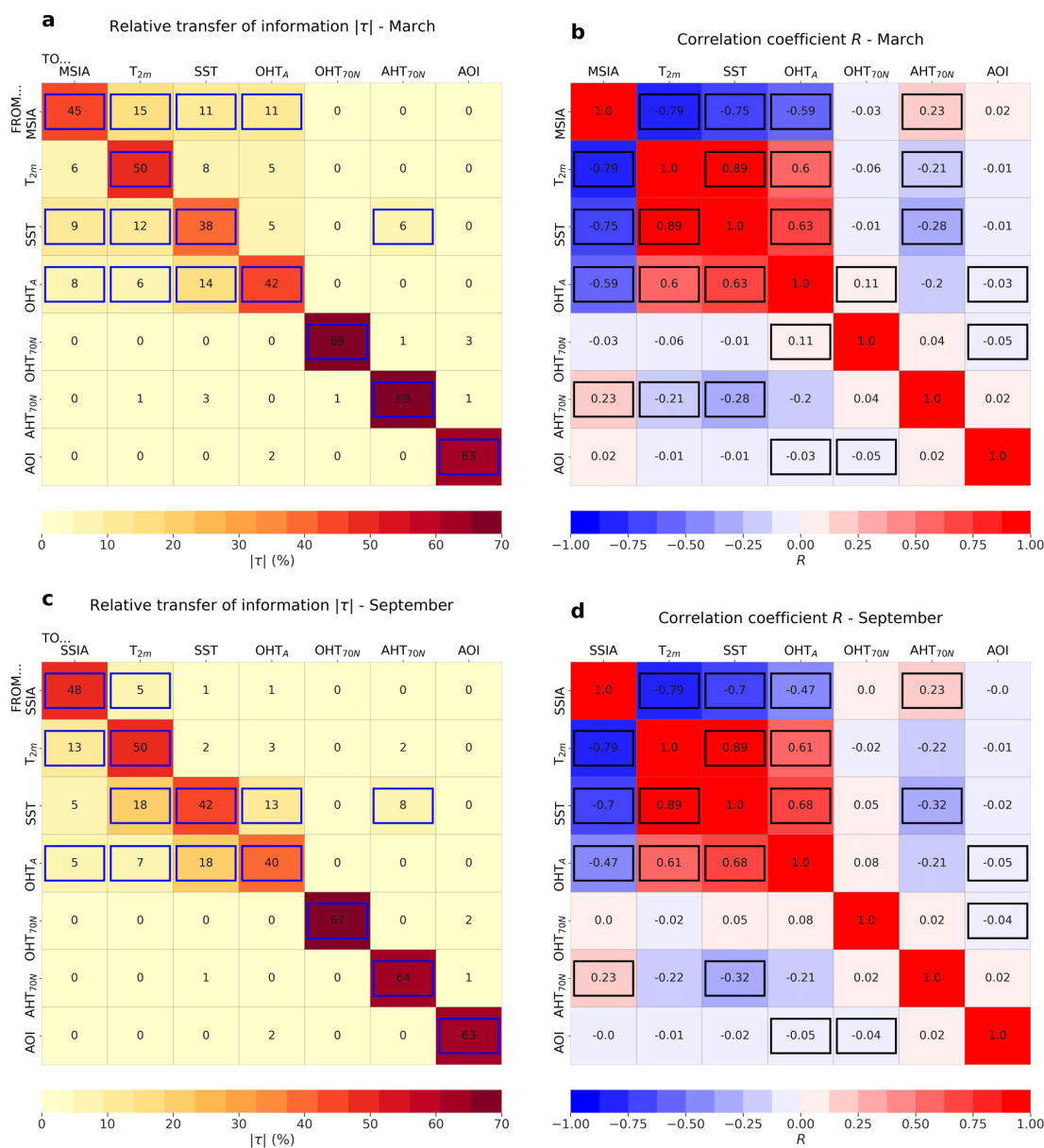


Fig. 1: Matrices of ensemble mean relative rate of information transfer (a,c) and correlation coefficient (b,d) between March (a,b) / September (c,d) sea-ice area (SIA) and its potential drivers (T_{2m} : Arctic near-surface temperature; SST: Arctic sea-surface temperature; OHT_A : total Arctic Ocean heat transport; OHT_{70N} : ocean heat transport at $70^\circ N$; AHT_{70N} : atmospheric heat transport at $70^\circ N$; AOI: Arctic Oscillation Index).

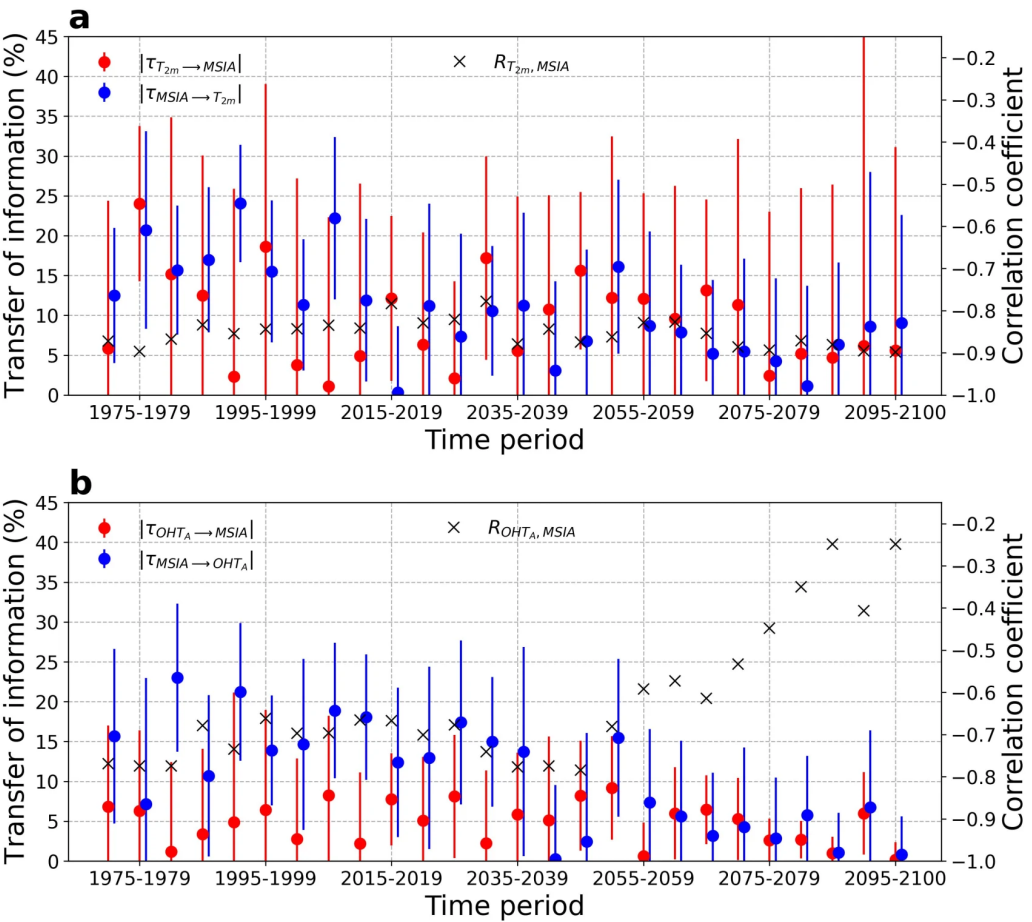


Fig. 2: Time evolution of relative rate of information transfer (dots and error bars) and correlation coefficient (crosses), for each period of 5 years, between March sea-ice area (MSIA) and near-surface air temperature (T_{2m}) (a), and between MSIA and total Arctic Ocean heat transport (OHT_A) (b).

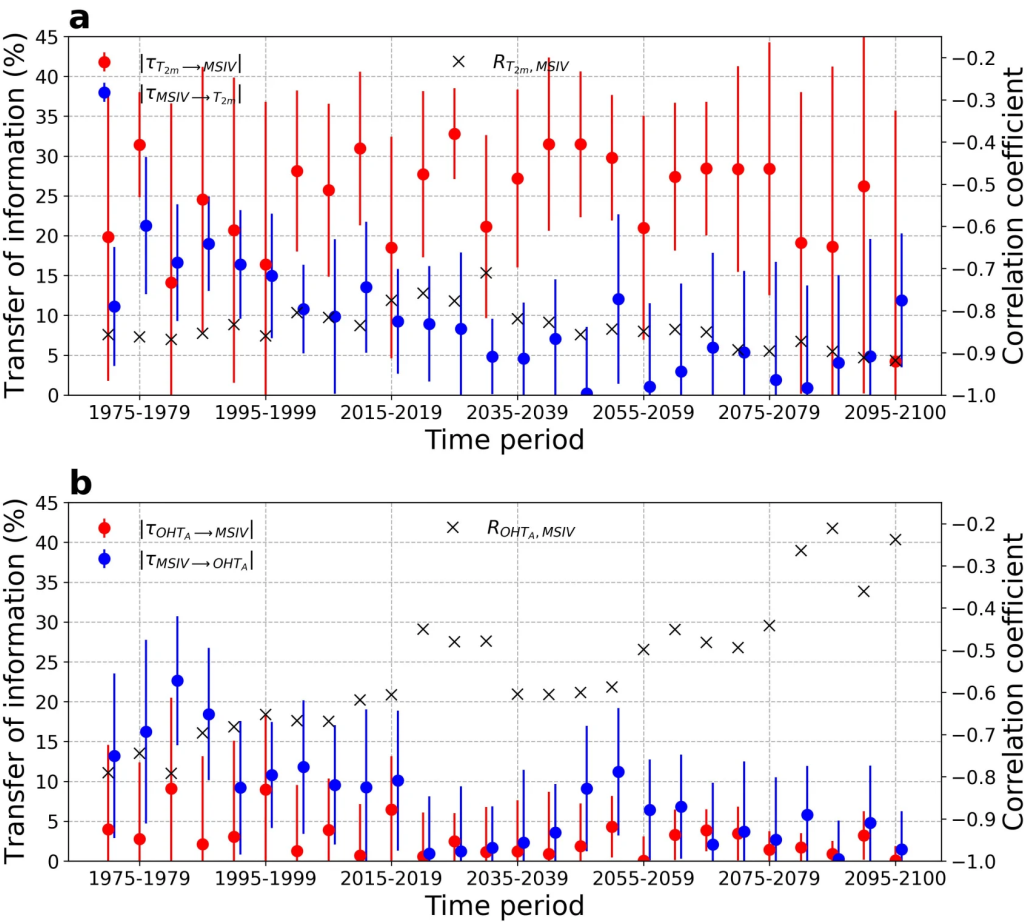


Fig. 3: Time evolution of relative rate of information transfer (dots and error bars) and correlation coefficient (crosses), for each period of 5 years, between March sea-ice volume (MSIV) and near-surface air temperature (T_{2m}) (a), and between MSIV and total Arctic Ocean heat transport (OHT_A) (b).

RESULTS EXPLAINED

1. Member analysis

- Rate of information transfer computed for each member separately across time and then merged together (ensemble mean) (**Fig. 1**)
- Winter ocean-driven influence: March Arctic sea-ice area mainly driven by Arctic sea-surface temperature (SST) and Arctic Ocean heat transport (OHT_A) (**Fig. 1a**)
- Summer atmospheric-led influence: September Arctic sea-ice area mainly driven by Arctic near-surface air temperature (T_{2m}) (**Fig. 1c**)
- Influence of sea-ice area on T_{2m} , SST and OHT_A (**Fig. 1a,c**)
- No influence of ocean and atmospheric heat transports at $70^\circ N$ (OHT_{70N} and AHT_{70N}) and winter Arctic Oscillation Index (AOI) on sea-ice area despite significant correlations (Fig. 1) --> external driver at play

2. Time analysis

- Rate of information transfer computed for each period of 5 years separately across the member space (**Figs. 2-3**)
- Progressive loss of influence of sea-ice area and volume on T_{2m} and OHT_A --> weaker interactions as sea-ice area and volume decrease (**Fig. 2**)
- Rate of information transfer from T_{2m} to sea-ice volume remains more constant across time than from OHT to sea-ice volume --> long-lasting effect of T_{2m} (**Fig. 3**)

More information: Docquier et al. (in review), preprint available on ESSOAr
(<https://doi.org/10.1002/essoar.10507846.1>)

WHY IS IT IMPORTANT?

- Arctic sea-ice area and volume have decreased by ~2 million km² and ~12,000 km³ since 1979, respectively (Onarheim et al., 2018 (<https://doi.org/10.1175/JCLI-D-17-0427.1>); Schweiger et al., 2019 (<https://doi.org/10.1175/JCLI-D-19-0008.1>))
- Model projections show a continuation of this process, which depends on the emission scenario (SIMIP, 2020 (<https://doi.org/10.1029/2019GL086749>); Docquier & Koenigk, 2021 (<https://doi.org/10.1038/s43247-021-00214-7>))
- Exact drivers of these changes are not entirely known
- Influence of one variable on another is usually quantified via correlation and regression analyses
- Causal inference frameworks allow to quantify causal links between variables, and thus go beyond classical correlation analyses
- In our study, we use the Liang-Kleeman information flow method to analyze the influence of potential climate drivers on Arctic sea-ice area and volume, as well as the reverse impact of sea-ice area and volume on these climate drivers (Liang, 2021 (<https://doi.org/10.3390/e23060679>))
- Our understanding of climate processes in polar regions (and at other latitudes) could greatly benefit from using the information flow method

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ABSTRACT

Arctic sea ice has substantially changed over the past four decades, with a large decrease in sea-ice area and volume. The exact causes of these changes are not entirely known. In our study, we make use of the Swedish Meteorological and Hydrological Institute Large Ensemble (SMHI-LENS). This ensemble consists of 50 members realized with the EC-Earth3 global climate model and covers the period 1970-2100. We apply the Liang-Kleeman information flow method to analyze the cause-effect relationships between Arctic sea ice and its potential drivers. We show that recent and future changes in Arctic sea ice are mainly driven by air and sea-surface temperatures and ocean heat transport. Conversely, changes in Arctic sea ice also considerably impact temperature and ocean heat transport. Finally, we find a progressive decrease in the influence of sea-ice area and volume on air temperature and ocean heat transport through the twenty-first century.

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