On the linkage between future Arctic sea ice retreat, Euro-Atlantic circulation regimes and temperature extremes over Europe

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Abstract. The question to what extent Arctic sea ice loss is able to affect atmospheric dynamics and climate extremes over mid-latitudes still remains a highly debated topic. In this study we investigate model experiments from the Polar Amplification Intercomparison Project (PAMIP) and compare experiments with future sea ice loss prescribed over the entire Arctic, as well as only locally over the Barents/Kara Sea with a present day reference experiment. The first step is to perform a regime analysis

5 and analyze the change of occurrence frequencies of five computed Euro-Atlantic winter circulation regimes. Forced by future Arctic sea ice conditions, most models show more frequent occurrences of a Scandinavian blocking pattern ~~and of a regime that resembles the negative~~ in at least one winter month, whereas there is an overall disagreement between individual models on the sign of frequency changes of two regimes that respectively resemble the negative and positive phase of the North Atlantic Oscillation. Focusing on the ECHAM6 PAMIP experiments, we subsequently employ a framework of conditional

10 extreme event attribution. It demonstrates how detected regime frequency changes can be used to decompose sea ice induced frequency changes of European temperature extremes into two different contributions: one "~~Changed-Regime~~changed-regime" term that is related to dynamical changes in regime occurrence frequencies, and another more thermodynamically motivated "~~Fixed-Regime~~fixed-regime" contribution that is related to increased surface temperatures during a specific circulation regime. We show how the overall fixed-regime warming effect, but also an increased Scandinavian blocking pattern frequency under

15 future sea ice reductions can equally contribute and shape the overall response signal of European cold extremes in midwinter. We also demonstrate how a decreased occurrence frequency of an anticyclonic regime over the eastern Atlantic dynamically counteracts the fixed-regime warming response and results in no significant changes in overall January warm extreme occurrences. However, when compared to other characteristics of future climate change, such as the thermodynamical impact of globally increased sea surface temperatures, the effects of Arctic sea ice loss on European temperature extremes are of

20 secondary relevance.

1 Introduction

Recent global warming includes a phenomenon called Arctic Amplification that comes along with an up to four times faster warming of Arctic regions compared to global average over recent decades (Rantanen et al., 2022). This amplified Arctic

warming is predominantly observed in winter time and is accompanied by an unprecedented shrinkage of Arctic sea ice concentration and thickness (Stroeve and Notz, 2018). Model projections forced under different greenhouse gas scenarios show clear evidence of a continuation of sea ice decline, with some models suggesting a seasonally ice-free Arctic till the mid of the century (Notz and Coauthors, 2020) . Aside from local ecological and economical impacts (Meredith et al., 2019) the question to what extent Arctic climate change and related sea ice loss may impact mid-latitude weather and general atmospheric dynamics has received a lot of attention over the last years and decades (e.g. Cohen et al., 2020; Screen, 2017b; Handorf et al., 2015; Cohen et al., 2014). A large variety of potential hemispheric-wide atmospheric responses have been detected and hypothesized in connection to Arctic sea ice loss. Such responses include for instance a commonly observed negative winter NAO response (e.g. Screen, 2017b; Nakamura et al., 2015; Jaiser et al., 2012), a highly debated weakening and stronger meandering of the jet stream that may result in more stationary and slower propagating large-scale Rossby waves (Francis and Vavrus, 2012; Barnes and Screen, 2015; Riboldi et al., 2020), as well as an intensification of the Scandinavian and Ural highs leading to continental winter cooling over Eurasia (Cohen et al., 2018). In this respect, dynamical pathways have been proposed relating for instance sea ice and snow cover anomalies in autumn to enhanced vertical wave activity fluxes and a weakened stratospheric polar vortex (Smith et al., 2022). These stratospheric disturbances could subsequently propagate downward and finally result in a late winter negative NAO response (Cohen et al., 2014; Nakamura et al., 2016; Jaiser et al., 2016; Sun et al., 2015). Especially the Barents/Kara Sea region, being a hotspot of recent Arctic sea ice retreat, has been argued to play an essential role for triggering such dynamical pathways (Screen, 2017a; Jaiser et al., 2016; Kretschmer et al., 2016). Nevertheless, no overall consensus about linkages and the underlying dynamical pathways has been reached until now (Cohen et al., 2020; Blackport and Screen, 2020), mostly due to discrepancies between observational and modeling studies. A recent study by Siew et al. (2020) highlighted for instance that the intermittent and state-dependent character of the aforementioned stratospheric pathway might be a potential reason for the typical low signal-to-noise ratios of atmospheric responses to sea ice changes. Furthermore, Petoukhov and Semenov (2010) showed how the modeled atmospheric response can depend on the magnitude of prescribed sea ice loss in the Barents/Kara Sea in a highly nonlinear way. Although most studies on Arctic-midlatitude linkages focus on the role of sea ice changes, several recent studies (He et al., 2020; Labe et al., 2020) also highlighted the importance of the vertical extent of Arctic warming into the upper troposphere compared to sea ice loss alone.

From a more large-scale and regime-oriented perspective, atmospheric dynamics can be viewed in a variety of conceptual frameworks (Hoskins and Woollings, 2015) including for instance jet stream states, blockings or atmospheric circulation regimes. Especially the framework of circulation regimes has been employed in a large variety of studies (e.g. Crasemann et al., 2017; Horton et al., 2015) in order to characterize the atmospheric circulation. Circulation regimes provide physically meaningful categorizations (Hochman et al., 2021) of atmospheric low-frequency variability into different regime states and have also been considered as preferred or quasi-stationary states of the underlying nonlinear atmospheric system (Hannachi et al., 2017). It has been hypothesized that weak external forcings imposed to the atmospheric system are able to modify the occurrence probability of such regime states (Corti et al., 1999; Gervais et al., 2016), while not affecting the overall regime structure (Palmer, 1999). Indeed, Crasemann et al. (2017) compared atmosphere-only model experiments forced under low and high sea ice conditions relative to the recent past and showed how the occurrence probability of certain Euro-Atlantic

circulation regimes can be significantly affected by such Arctic sea ice changes. In this case the induced sea ice changes were considered as weak forcings applied to the atmospheric system.

A major interest for human society nowadays is given by the question to what extent the recently observed increasing number of climate extremes (Coumou and Rahmstorf, 2012) can be attributed and affected by anthropogenic global warming (Otto, 2016). Basically, there is an overall agreement that from a thermodynamical perspective global warming will lead to less (more) frequent and intense cold (warm) extremes. Nevertheless, the occurrence of cold spells like over Europe in 2010 (Cattiaux et al., 2010) or the more recent cold air outbreak over North America in 2021 (Bolinger et al., 2022) might be considered as contradictions to this simplified thermodynamical perspective. In this respect, Cattiaux et al. (2010) illustrated for instance how the European winter cold spell in 2010 was, from a thermodynamical point of view, perfectly in line with recent global warming when accounting for the anomalous negative NAO situation during this winter. Shepherd (2016) framed a storyline approach aiming to unfold specific classes of extreme events into the different contributing factors by including both, dynamical and non-dynamical contributions. However, circulation changes found in climate model simulations typically suffer low signal-to-noise ratios (Scaife and Smith, 2018; Smith et al., 2022). Therefore, changes regarding the dynamical situation leading to a certain extreme should only be included into an analysis when there is solid evidence that changes in atmospheric circulation can be expected or reliably detected (Trenberth et al., 2015; Shepherd, 2016).

Since, as mentioned above, Arctic sea ice retreat has been proven to be potentially able to modify atmospheric large-scale dynamics, the question appeared how changes in mid-latitude weather can be dynamically and thermodynamically attributed to Arctic sea ice changes. Screen (2017b) compared large ensembles of atmosphere-only experiments forced under low and high sea ice conditions relative to the recent past. They observed that despite an intensification of negative winter NAO events under low Arctic sea ice conditions an expected dynamically induced European cooling response was absent, mostly due to compensation effects related to an overall thermodynamical warming. Another study by Deser et al. (2016a) investigated how different complexities of an ocean model can affect the large-scale hemispheric circulation response to Arctic sea ice loss. They compared model simulations with Arctic sea ice conditions constrained to the late 21st and to the 20th century. On the one hand they argued that under reduced sea ice conditions elevated sea level pressures over northern Siberia and Arctic regions are associated with anomalous northeasterly advection of cold Arctic air masses towards central Eurasia. This dynamically induces a cooling response over the respective central Eurasian regions. On the other hand, this dynamical cooling effect may be thermodynamically counteracted by elevated SSTs, which was especially the case for coupled ocean–atmosphere model setups. Recent studies however argue that such coupled model setups artificially overestimate the impact of sea ice loss (England et al., 2022). Recently, Chripko et al. (2021) studied fully coupled model experiments where the sea ice albedo parameter was reduced to ocean value yielding mostly ice-free conditions from July to October and moderate sea ice reductions in winter. When compared to a control simulation they detected winter warming signals over Europe and North America in the sensitivity experiment. By applying a dynamical adjustment method (Deser et al., 2016b) they showed that these overall responses could be explained by a combination of a dynamical response and a residual contribution.

Based on such previous studies that decomposed and linked changes in mid-latitude weather and dynamics to Arctic sea ice loss, as well as due to the high societal relevance of extreme events nowadays the question arises to what extent future sea ice retreat is able to impact the occurrence of extreme weather events.

95 Therefore, in this study we investigate the impact of future Arctic sea ice loss on the mid-latitude circulation over the Euro-Atlantic domain and related European temperature extremes. Here, we will focus on winter temperature extremes over the European region that can have significant impacts on society (Díaz et al., 2005) and economy (Savić et al., 2014; Añel et al., 2017) over such densely populated regions. In order to assess and isolate the impact of Arctic sea ice changes we will investigate model experiments from the Polar Amplification Intercomparison Project (PAMIP). The experiments that are considered

100 here are forced under present day and reduced future sea ice conditions over the entire Arctic, as well as under sea ice conditions only locally reduced over the Barents/Kara Sea. The latter allows for assessing the role of sea ice loss specifically in the Barents/Kara Sea region. Focusing on circulation changes detected in the ECHAM6 model, as well as by employing a framework of conditional extreme event attribution (Yiou et al., 2017) we will demonstrate how overall sea ice induced changes in extreme occurrences can be decomposed into two different contributions: one ~~Fixed-Regime~~ fixed-regime term that compares

105 extreme occurrence frequencies for a given and fixed circulation regime, as well as a ~~Changed-Regime~~ changed-regime contribution term that is related to changes in occurrence frequencies of the respective regime. More specifically, the analysis steps can be divided into different research questions that are partially linked to each other:

1. Within the methodological framework of atmospheric circulation regimes, what changes in the wintertime atmospheric large-scale circulation over the Euro-Atlantic sector can be expected under future Arctic sea ice retreat?

110 2. Which regimes can be associated with preferred occurrences of winter temperature extremes over Europe?

3. What overall frequency changes of extreme occurrences over the continental Northern Hemisphere can be detected in response to future sea ice changes in ECHAM6?

4. Based on the sea ice induced changes in circulation regimes detected in ECHAM6, to what extent can frequency changes of European extremes be related to ~~Fixed-Regime and Changed-Regime~~ fixed-regime and changed-regime contributions?

115 When studying the impact of Arctic sea ice changes on mid-latitude circulation and weather, the question may arise how such impacts compare to atmospheric responses induced by other facets of future climate change. Therefore, in order to assess the relative importance of sea ice loss on future changes in European extremes, the analysis will be complemented by investigating the impact of a globally increased future sea surface temperature (SST) background state prescribed in one of the experimental setups.

120 ## 2 Data

In this study we initially analyze different sea ice sensitivity simulation data from the Polar Amplification Intercomparison Project (PAMIP, Smith et al., 2019). Table S1 provides an overview ~~over~~ of the considered models and ensemble sizes. The

PAMIP protocol aims at a better understanding of the impact and relative roles of Arctic sea ice and SST changes on the global climate system. Therefore, each sensitivity experiment includes ~~a certain number of~~ at least 100 ensemble members of one-year-long atmosphere-only time slice simulations that are forced under different annual cycles of sea ice, but also SST boundary conditions. As recommended by Smith et al. (2019), initial conditions of each ensemble member are based on AMIP simulations for 1st April 2000 and each ensemble member was run for 14 months, but the first two months were finally excluded for model spin up reasons. In order to study the impact of future sea ice changes on circulation regimes we analyze sensitivity simulations forced under:

- present day SST and present day sea ice conditions (pdSIC/pdSST, PAMIP setup 1.1)

- present day SST and future/reduced Arctic-wide sea ice conditions (futArcSIC, PAMIP setup 1.6)

- present day SST and future/reduced sea ice in the Barents/Kara Sea region 65-85°N, 10-110°E (futBKSIC, PAMIP setup 3.2).

For the analyses in Secs. 4.2–4.4 we focus on the 100 ensemble members from the ECHAM6 experiments. ECHAM6 is the latest release of the atmospheric general circulation model ECHAM that was developed at the Max-Planck-Institute for Meteorology (MPI) in Hamburg (Stevens et al., 2013). The ECHAM6 setup used for the PAMIP experiments operates on 95 vertical layers up to 0.01 hPa, and with a spectral T127 horizontal resolution (resulting in a zonal resolution of 100 km in the tropics and for instance 25 km at 75°N). In order to contrast the importance of future SST with Arctic sea ice changes in the very end of this study, we also consider 100 ensemble members from an ECHAM6 sensitivity simulation forced under

- present day sea ice and globally raised future SST conditions (futSST, PAMIP setup 1.4).

The pdSIC/pdSST simulations serve in a first place as reference simulations to which the sensitivity simulations futArcSIC and futBKSIC are compared with. Comparisons of sea ice and SST forcing fields of the respective present day and sea ice sensitivity simulations are shown in Smith et al. (2019) in Figs. 5 and 6. In winter, future sea ice conditions are predominantly reduced over the Barents/Kara Sea, the Sea of Okhotsk, the Bering Sea and parts west and east of Greenland. Summer conditions are characterized by strong reductions and ice-free areas over central Arctic regions.

Present day forcing fields are obtained from the climatologies of observations from the Hadley Centre sea ice and Sea Surface Temperature dataset over the period 1979–2008 (Rayner et al., 2003). Future conditions are derived from RCP8.5 multimodel simulations for a 1.43 (2)°C warming scenario over present day (preindustrial) conditions (for more details see Smith et al. (2019) Appendix A). At grid points where sea ice has been removed under future conditions the present day SSTs are replaced by future SSTs if the difference in sea ice concentration between future and present day is greater than 10 %. Sea ice thickness at each grid point is set to 2 m for all simulations and greenhouse gas forcings are constantly set to present day conditions of the year 2000.

For the analysis presented in this study we use daily sea level pressure (slp) data from all PAMIP models, as well as daily maximum/minimum near-surface air ~~temperature~~ temperatures (Tmax/Tmin) from ECHAM6. The daily temperature and slp

data in ECHAM6 are provided on a regular lon-lat grid with 0.9375° resolution. The slp data from ECHAM6 (and from all other PAMIP models as well) are however finally regridded to a 100×100 km equal-area grid (see also next ~~Section~~section).

In order to complement and backup certain parts of our analysis with real world data, we additionally used slp, 2 meter temperature and sea ice area data from the ERA5 reanalysis over the period 1979–2018 (Hersbach et al., 2020).

3 Methods

3.1 Circulation regimes

In this study, we compute centroids C_i of atmospheric circulation regimes for the extended winter season with the k-means clustering algorithm (Michelangeli et al., 1995; Crasemann et al., 2017; Straus et al., 2007) applied to daily slp anomaly data merged together from two different experiments (typically the pdSIC/pdSST reference simulation and one of the sensitivity simulations) over the Euro-Atlantic domain (90°W-90°E,20°N-88°N). Before applying the clustering algorithm, slp data were regridded to a 100×100 km equal-area grid in order to avoid grid point convergence at higher latitudes. Generally speaking k-means clustering aims to minimize the intra-cluster to inter-cluster variance ratio by an iterative allocation and exchange procedure of cluster members (MacQueen, 1967). In order to reduce computational demands, we applied a dimensionality reduction via principal component analysis prior to the clustering algorithm. Here we used the first 20 principal components that roughly explain around 90% of variance of the winter slp anomaly fields. Further increasing the number of principal components did not affect the final outcome of the clustering algorithm. The k-means algorithm has been initialized 1000 times and the best result in terms of minimizing the aforementioned variance ratio has been finally chosen. Based on the Euclidean distance, the respective slp anomaly field or atmospheric flow F at each day is finally assigned to the best-matching cluster centroid C_i.

Slp anomalies are generally calculated as deviations from an annual cycle, which is obtained by averaging each day of a year over all years. For the merged pdSIC/pdSST+futBKSIC and pdSIC/pdSST+futArcSIC datasets we computed a joint annual cycle of both simulations. It shows that the resulting cluster allocations are not considerably affected by whether the slp anomalies have been calculated as deviations from the joint annual cycle as described above, or by removing the annual cycles for each experiment individually (as done by e.g. Crasemann et al., 2017). This is also related to the fact that when contrasting the reference with both sea ice sensitivity experiments the respective winter slp background states showed mostly negligible differences, neither did they project on any mode of variability. In contrast to the sea ice sensitivity simulations, the relatively strong forcing in the ECHAM6 futSST experiment leads to an evident change of the slp background state (with respect to the reference simulation) that strongly projects on a positive NAO pattern. This background difference pattern significantly affects the final cluster allocations when subtracting a joint annual cycle. Therefore, we computed the annual cycle for both simulations individually when merging data from the futSST and the pdSIC/pdSST experiments to take into account the different background states.

A subtle part when applying cluster algorithms as k-means is to prescribe the number of clusters and therefore make an assumption about the number of existing atmospheric circulation regimes beforehand. Several attempts have been made in

order to determine such an optimal number of winter regimes with most studies indicating a number between four and six regimes (Falkena et al., 2020). Here we stick to a cluster number of five which is supported by recent studies (Crasemann et al., 2017; Dorrington and Strommen, 2020).

3.2 Conditional extreme event attribution framework

In this study we also aim to identify thermodynamically and dynamically induced contributions to overall European temperature extreme frequency changes in the ECHAM6 sea ice sensitivity experiments. Dynamically induced changes in the occurrence frequencies of certain local extreme events are related to changes in the relevant dynamical conditions, e.g. in terms of more frequent occurrences of the respective atmospheric flow patterns that promote a certain extreme. In contrast, thermodynamical contributions are typically associated with changes of extreme probabilities that would also occur in the absence of any relevant dynamical changes (e.g. due to overall global warming). From a methodological point of view it is however challenging to clearly separate dynamical and thermodynamical components. This issue is related to the fact that there is generally no unique way to define and detect changes in all contributing dynamical and non-dynamical factors that impact a certain class of extreme event.

Nevertheless, a variety of approaches have been outlined over the years (e.g. Yiou et al., 2017; Deser et al., 2016b; Vautard et al., 2016; Cassano et al., 2007) that aim to decompose atmospheric responses into thermodynamical and dynamical contributions. In this study a framework for conditional extreme event attribution (Yiou et al., 2017) is utilized. This method provides a suitable approach for decomposing changes in extreme event occurrence frequencies while employing the framework of circulation regimes.

In this study winter extreme events are defined as exceedances (or falls below) of a threshold temperature that is based on the 100 simulated winters in the ECHAM6 reference pdSIC/pdSST /pdSICsimulation. The threshold temperature $T_{\text{ref,w}}$ ($T_{\text{ref,c}}$) of warm (cold) extreme events at a given grid point is computed for each winter month separately as the 0.95 (0.05) quantile of the respective underlying daily Tmax (Tmin) distribution in pdSIC/pdSST/pdSIC.

Based on this definition we define the probabilities p_0 (p_1) in a counterfactual (factual) world of a warm extreme occurrence at a certain grid point as the probability Pr that some daily $Tmax_0$ ($Tmax_1$) in the counterfactual (factual) world exceeds the defined threshold temperature:

$$p_{0/1} = Pr(Tmax_{0/1} > T_{\text{ref,w}}). \tag{1}$$

In this study, we define the factual world (the world as it is) as the pdSIC/pdSST /pdSICreference simulation. The counterfactual world (a world that might occur) is given by the different ECHAM6 PAMIP sea ice sensitivity simulations mentioned before.

Similar to warm extremes, probabilities p_0 and p_1 of cold extreme occurrences are defined as the probability that some daily $Tmin_0$ ($Tmin_1$) in the counterfactual (factual) world falls below the defined threshold temperature:

$$p_{0/1} = Pr(Tmin_{0/1} < T_{\text{ref,c}}). \tag{2}$$

By employing Bayes' formula the occurrence probabilities of cold extremes (and similar for warm extremes) can be expressed with conditional probabilities as

$$p_{0/1} = Pr(Tmin_{0/1} < T_{\text{ref,c}} | F_{0/1} \in \mathcal{C}_{\text{ref}}) \cdot \frac{Pr(F_{0/1} \in \mathcal{C}_{\text{ref}})}{Pr(F_{0/1} \in \mathcal{C}_{\text{ref}} | Tmin_{0/1} < T_{\text{ref,c}})} \tag{3}$$

Here, $\mathcal{C}_{\text{ref}}$ describes the set of all slp anomaly fields or atmospheric flows $F_{0/1}$ in the respective world that are allocated to a certain reference regime centroid C_{ref}. When applying this decomposition we assume that the storyline of an extreme at a specific grid point can be explained by the presence of a specific reference regime C_{ref}.

The probability or risk ratio ρ compares the extreme occurrence probabilities of cold and warm extremes in the counterfactual (p_0) and in the factual world (p_1). When using Eq. 3 this ratio can be multiplicatively decomposed into

$$\rho = \frac{p_0}{p_1} = \rho_{FR} \cdot \rho_{CR} \tag{4}$$

that is a term ρ_{CR} ("~~Changed-Regime~~changed-regime") relating changes in extremes to changes in regime occurrences, and another term ρ_{FR} ("~~Fixed-Regime~~fixed-regime") that considers such changes in extremes by fixing atmospheric dynamics to a certain circulation regime.

For cold extremes, the ~~Fixed-Regime~~ fixed-regime contribution term is given by

$$\rho_{\text{FR}} = \frac{Pr(Tmin_0 < T_{\text{ref,c}} | F_0 \in \mathcal{C}_{\text{ref}})}{Pr(Tmin_1 < T_{\text{ref,c}} | F_1 \in \mathcal{C}_{\text{ref}})} \tag{5}$$

This contribution term describes the extreme occurrence probability ratio between both worlds given a regime allocation $F_{0/1} \in \mathcal{C}_{\text{ref}}$ to a certain reference regime set $\mathcal{C}_{\text{ref}}$. This term has previously been named thermodynamical contribution (Yiou et al., 2017), as the atmospheric circulation is fixed in terms of circulation regimes. Nevertheless, caution is needed when using such names as this term to a certain extent assumes that the regime pattern structures do not change between simulation scenarios. For weak forcings this has however been shown to be a valid assumption (Palmer, 1999) (see also ~~Figure ?? for~~ Fig. S3 for a comparison of different pattern structures computed for different combinations of simulations). In addition to this, the individual flows allocated to a respective set $\mathcal{C}_{\text{ref}}$ may also differ between different simulations.

The second contribution related to regime changes is defined as

$$\rho_{\text{CR}} = \rho_{\text{reci}} \cdot \rho_{\text{circ}} = \frac{Pr(F_1 \in \mathcal{C}_{\text{ref}} | Tmin_1 < T_{\text{ref,c}})}{Pr(F_0 \in \mathcal{C}_{\text{ref}} | Tmin_0 < T_{\text{ref,c}})} \cdot \frac{Pr(F_0 \in \mathcal{C}_{\text{ref}})}{Pr(F_1 \in \mathcal{C}_{\text{ref}})}. \tag{6}$$

The latter term ρ_{circ} is related to changes in the occurrence probability of the reference regime C_{ref} between both simulations. Therefore, this term has previously also been termed dynamical contribution (Yiou et al., 2017), as ρ_{circ} can be directly associated to dynamical changes within the framework of circulation regimes. The term ρ_{reci} evaluates changes in the probability of a circulation such as C_{ref} when given an extreme. ρ_{reci} allows for connecting the more meaningful and interpretable terms $\rho, \rho_{FR}, \rho_{\text{circ}}$ and it has also been suggested by Yiou et al. (2017) that it helps to reconcile the risk-based approach (estimation of ρ only) with the storyline approach.

3.3 Uncertainty estimates

Uncertainty and significance estimates are reported with confidence intervals based on the 0.05 and 0.95 quantile of boot-strapped distributions of the relevant statistic. If the computed confidence intervals do not include unity (for ratios) or a zero value (for differences) the signal is termed significant. Daily temperature time series, as well as daily nominal time series of cluster allocations typically exhibit significant temporal dependencies over several days. In order to preserve the temporal structure of the original daily data during the resampling procedure a moving block bootstrap is used here (Kunsch, 1989).

The original time series x_n of length n is therefore divided into overlapping blocks of size k, where the first Block contains $x_1, ..., x_k$ and the second block $x_2, ..., x_{k+1}$ etc. . Afterwards, a bootstrap sample is created by concatenating randomly picked blocks to a new time series of original length n and the statistic of interest (cluster frequency, ρ etc.) is computed for this generated bootstrap sample time series. When employing this procedure for statistics where multiple variables are involved (e.g. ρ_{reci} and ρ_{FR}), the time series of temperatures and regime allocations are blocked and resampled pairwise. This procedure is repeated 1000 times yielding a bootstrapped probability distribution of the respective statistic of interest. The block length k is set to 5 days corresponding to a typical persistence time of the circulation regimes.

4 Results and Discussion

In the upcoming section we present results of the analysis steps already outlined in the introduction. Initially, the impact of Arctic sea ice changes on the large-scale atmospheric winter circulation is assessed within the context of atmospheric circulation regimes (Sect. 4.1). Therefore, we initially present and discuss the regime structures of ECHAM6 and several other PAMIP models, and compare these pattern structures to ERA5 regimes (Sect. 4.1.1). Afterwards, we discuss how future Arctic sea ice changes in different PAMIP models impact the occurrence frequencies of such circulation regimes (Sect. 4.1.2). For the signals detected in ECHAM6 we identify those frequency changes that are in agreement with recently observed ERA5 tendencies. Focusing on ECHAM6, we subsequently demonstrate how winter temperature extremes over Europe can be associated with certain circulation regimes (Sect. 4.2). Based on the previous analysis steps and after discussing to what extent overall changes in winter temperature extremes can be detected in the ECHAM6 sea ice sensitivity simulations (Sect. 4.3), we finally assess how these changes over the European domain can be decomposed into ~~Fixed- and Changed-Regime~~ fixed- and changed-regime contributions (Sect. 4.4).

4.1 Regime structures and frequency changes induced by future Arctic sea ice retreat

To start with we discuss how the occurrence frequency of computed atmospheric circulation regimes is affected by future Arctic sea ice changes.
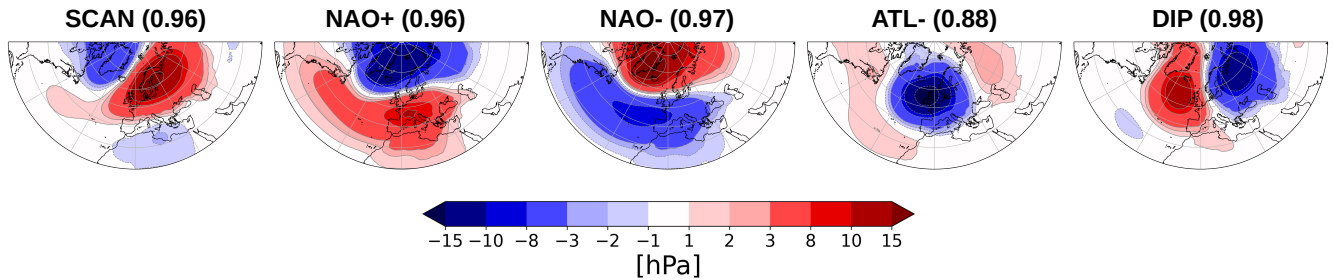
Figure 1. Five circulation regimes over the Euro-Atlantic domain computed from daily ECHAM6 PAMIP slp anomaly data for the extended winter season (December, January, February, March). The computed regimes include a Scandinavian Blocking pattern (SCAN), a positive and negative NAO-like pattern (NAO+/NAO-), an Atlantic trough pattern (ATL-) and a dipole pattern (DIP). The numbers in brackets show the pattern correlation coefficients of each pattern with the respective ERA5 pattern.

4.1.1 Regime structures

Figure 1 shows five circulation regimes computed for the extended winter season over the Euro-Atlantic domain (-90°W–90°W, 20°N–88°N). Daily slp anomaly data merged together from the ECHAM6 pdSIC/pdSST and the futArcSIC simulation data have been used here. The computed regimes closely resemble regimes found in previous studies (e.g. Crasemann et al.,
2017) and include a frequently detected Scandinavian Blocking regime (Dorrington and Strommen, 2020; Falkena et al., 2020; Yiou et al., 2017), termed SCAN, with an anticyclonic blocking structure over Scandinavia and parts of the Ural mountains. As shown in Fig. ??aS6, up to 40% of SCAN regime days are indeed accompanied by blocking activity over northern and northeastern Europe. Studies by for instance Jung et al. (2017) and Sato et al. (2014) showed that such an anticyclonic anomaly over northeastern Europe might be part of a wave train structure that originates from the east coast of North America and
is forced by warming anomalies over this remote region. Another regime is characterized by a cyclonic structure over the Atlantic and parts of western Europe (ATL-) and has previously been named negative Atlantic ridge (Falkena et al., 2020) or Scandinavian trough (Dorrington and Strommen, 2020). A dipole pattern (DIP) is found with positive pressure anomalies over the North Atlantic and negative pressure anomalies over northeastern Europe that has also been frequently termed Atlantic Ridge (Dorrington and Strommen, 2020; Falkena et al., 2020; Yiou et al., 2017). Finally, two of the computed regimes resemble
the positive (NAO+) and negative (NAO-) phase of the North Atlantic Oscillation, respectively. The structure of the individual regimes is relatively unaffected by the exact definition of winter season (e.g. by excluding March) and by modifications of the spatial domain (using e.g. -80°W-80°W, 30°N-88°N).

Compared to circulation regimes computed from ERA5 data (see Fig. ??S1), it appears that ECHAM6 is able to realistically reproduce the spatial structure of these five regimes. Indeed, Fig. ?? S3 indicates high spatial correlations (generally greater than 0.9), and similar (but e.g. for NAO+ slightly higher) pattern amplitudes when comparing regimes computed from different combinations of ECHAM6 model simulations with ERA5 regimes.

In addition to ECHAM6 we also computed five circulation regimes for other PAMIP models. The supplementary Fig. ~~S1~~ S4 displays Taylor diagrams that compare regime patterns computed from the eleven different PAMIP models with the ERA5 regime structures ~~(see Fig. ??)~~. It shows that nine PAMIP models are able to realistically reproduce the ERA5 regimes pattern

300 structures (pattern correlation averaged over all regimes greater than 0.8), whereas two models clearly stand out and show deficiencies in this respect (IPSL-CM6A-LR and FGOALS-f3-L). For the upcoming part of the analysis we only consider those nine PAMIP models that are able to realistically reproduce the ERA5 regime structures.

4.1.2 Regime frequency changes induced by future Arctic sea ice retreat

In order to assess the impact of future Arctic sea ice changes on the occurrence probability of certain regimes, Figs. ~~S2 and~~
305 ~~S3~~ 2 and S5 show monthly-splitted histograms for different PAMIP models that compare the relative occurrence frequencies of computed circulation regimes between the reference simulation and the futArcSIC (Fig. ~~S2~~2), as well as with the futBKSIC sea ice sensitivity experiment (Fig. ~~S3~~S5). Overall, it can be reported that regime frequency changes detected in the futArcSIC and futBKSIC sensitivity experiments of a specific model share similar features. Consistent with previous studies this again emphasizes the potential key role of sea ice loss in the Barents/Kara Sea region when trying to identify and understand linkages
310 between the Arctic and mid-latitudes. We decided to analyze the regime occurrence for each winter month separately as proposed pathways underlying Arctic-midlatitude teleconnections are often characterized by their evolution over the autumn-winter season (e.g. Kretschmer et al., 2016; Siew et al., 2020).

~~Seven out of~~ All nine models indicate ~~an~~ a significant increase of SCAN occurrences in futArcSIC in at least one winter month, while in contrast only the NorESM2-LM and CNRM-CM6-1 models show significantly decreased SCAN occurrences
315 (Fig. ~~S2), while this SCAN response is only significantly~~ 2). A significant increase of SCAN occurrences in at least one winter month is detected in two out of five models for futBKSIC (Fig. ~~S3~~S5). The respective months for which the SCAN response is detected generally differ between models. This may suggest that the underlying physical processes and pathways that lead to the occurrence of this SCAN increase are overall reasonably represented, but that the onset and timing of such processes may differ between models. In general, this winter SCAN response is consistent with previous studies, such as by Luo et al.
320 (2016) who related a strengthening of the Scandinavian or Ural Blocking in winter season to instantaneous sea ice loss in the Barents/Kara Sea region. Petoukhov and Semenov (2010) analyzed model simulations and showed that for moderate winter sea ice reductions over the Barents/Kara Sea an anticyclonic anomaly centered over the same region can be observed in February; however, they emphasized that such an anticyclonic circulation response depends on the actual prescribed magnitude of sea ice loss in the Barents/Kara Sea in a highly nonlinear way. Within the framework of circulation regimes Crasemann et al. (2017)
325 detected an increased December SCAN occurrence frequency—however only in response to recent Arctic sea ice loss. It should be mentioned that a variety of recent modeling studies (Kim et al., 2022; Peings, 2019) did not find any intensifications of Ural blockings in response to sea ice loss over the Barents/Kara Sea region.

Consistent with the recently reported weakening of mid-latitude westerly winds due to future sea-ice loss in the PAMIP-ensemble by Smith et al. (2022), five out of nine models in futArcSIC indicate preferred occurrences of the NAO- regime
330 in ~~mid-to late winter in futArcSIC~~ at least two winter months (see Fig. 2). In contrast, ~~several models also show decreased~~

also decreased NAO- occurrences in at least one winter month can be detected in five models as well—including ECHAM6. Occurrence frequency changes of the NAO+ regime in futArcSIC also show pronounced discrepancies among different models, with four (five) models indicating an increased (decreased) NAO+ occurrence frequency in at least one winter month. Hence, we are overall not able to detect a robust NAO response under future sea ice loss in futArcSIC when comparing the different

335 PAMIP models. For the futBKSIC experiment three out of five models indicate decreased NAO+ and NAO- occurrences ~~.~~ ~~Frequency changes for other regimes typically exhibit large variations between different models~~ (see Fig. S5), which may suggest a weakened dominance of NAO variability under future sea ice loss in the Barents/Kara Sea.

Although most models reveal significant frequency changes of the ATL- regime for futArcSIC and futBKSIC in different months, a consistent response among the different models can hardly be detected in any winter month. Occurrence frequencies

340 of the DIP regime exhibit a significant decrease in five (two) out of nine (five) models in the futArcSIC (futBKSIC) experiment.

In order to later on demonstrate in Sect. 4.4 how such detected regime frequency changes can be employed to decompose sea ice induced changes in European temperature extremes, we finally ~~focus on~~ briefly highlight the regime frequency changes detected in the ECHAM6 experiments (see Fig. ~~??).~~ S2). We will focus on only one model as especially a comprehensive interpretation of the upcoming decompositions for all nine models is very challenging and beyond the scope of this study.

345 Indeed, ECHAM6 is one of the best models that are able to realistically reproduce the ERA5 regime structures (see Fig. ~~S1~~S4). This additionally allows us to reasonably contrast the modeled ECHAM6 regime frequency changes to regime frequency changes between recent ERA5 low and high detrended Arctic sea ice conditions (see triangles in Fig. ~~??), and therefore, to only consider the significant ECHAM6 regime changes in Sec. 4.4 that are supported by such ERA5 tendencies~~S2). Low (high) detrended Arctic sea ice conditions in ERA5 are defined as the lower (upper) 50% of linearly detrended monthly averaged

350 Arctic sea ice area anomalies over the period 1979–2018. Such an ERA5 analysis does not prove any causal link between recent sea ice loss and circulation regimes, and does not isolate the effect of recent sea ice retreat. Nevertheless, we consider such ERA5 tendencies as additional statistical evidence, especially when deciding which of the significant ECHAM6 regime frequency changes are considered for the decompositions in Sect. 4.4.

In agreement with other PAMIP models~~(see e.g. Fig. S2)~~, an overall midwinter increase of SCAN occurrence is detected in

355 both ECHAM6 sea ice sensitivity simulation as well as in the reanalysis (see Figs. ~~??a~~ S2a and f). Another significant signal found in both, reanalysis and ECHAM6, is a more frequent occurrence of the ATL- pattern in January under higher sea ice conditions in the reference simulation (Figs. ~~??d~~ S2d and i). Lastly, especially the ECHAM6 futBKSIC sensitivity simulation reveals a decreased occurrence frequency of the NAO+ and NAO- pattern in February (Figs. ~~??g~~ S2g and h). ~~However, only the~~ The diminished occurrence frequency of the NAO+ pattern can be detected in the reanalysis as well.

360 ~~Relative regime occurrence frequencies in ECHAM6 for different winter months compared between the pdSST/pdSIC reference simulation (blueish bars) and the futArcSIC (upper row, redish bars), as well as the futBKSIC sensitivity simulation (lower row, redish bars). Light redish and blueish bars indicate non-significant frequency differences between reference and sensitivity simulations, whereas the paired dark blueish/redish bars indicate significant differences in occurrence frequencies. Note that by definition the sum over all clusters for a specific month in a given simulation is one. The triangles indicate the~~

365 ~~respective ERA5 regime occurrence frequencies for recent low (upright bright-blueish triangles) and high (inverted yellow~~
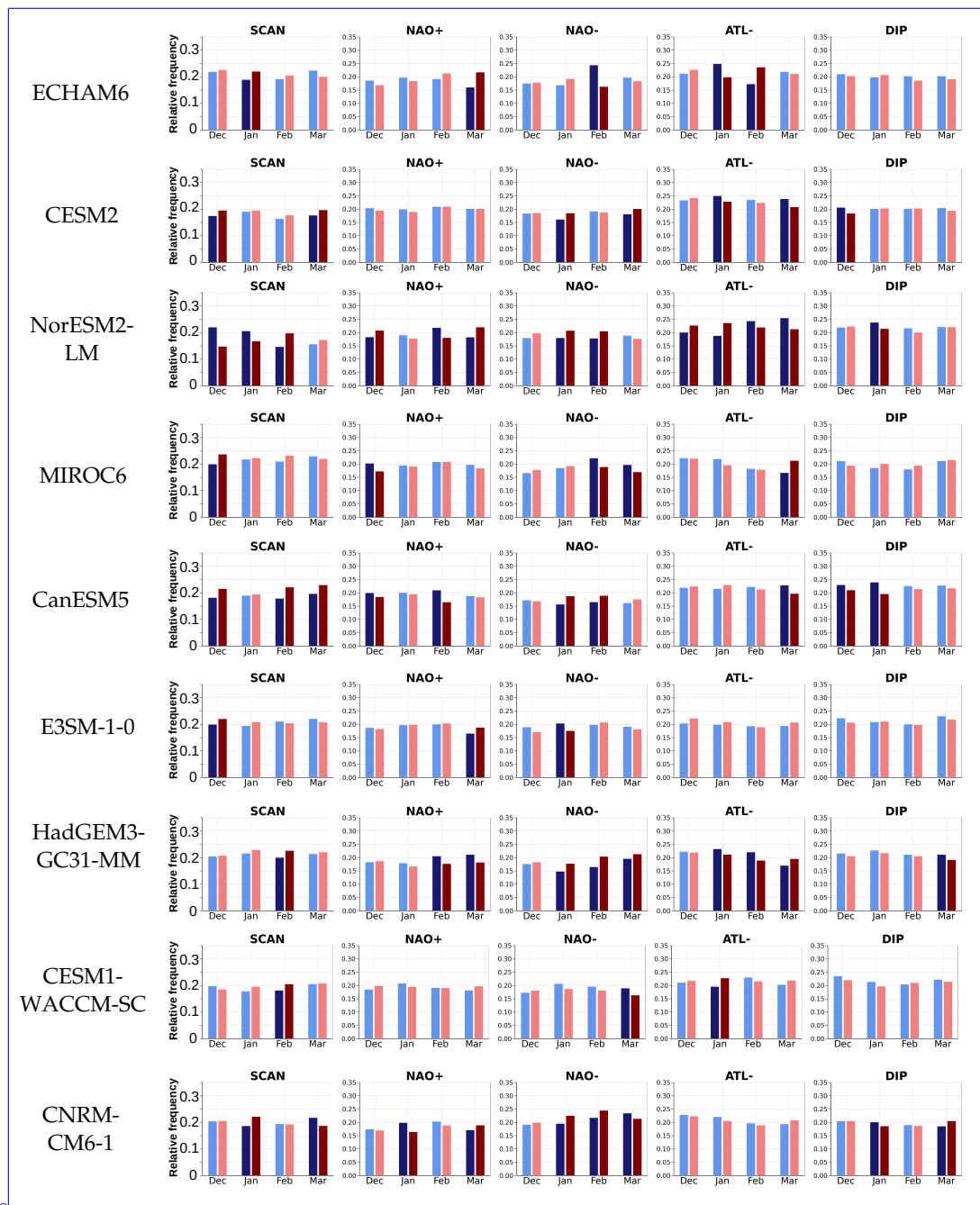
Figure 2. Relative regime occurrence frequencies in different PAMIP models for different winter months compared between the pdSIC/pdSST reference simulation (blueish bars) and the futArcSIC sensitivity simulation (redish bars). Light redish and blueish bars indicate non-significant frequency differences between reference and sensitivity simulations, whereas the paired dark blueish/redish bars indicate significant differences in occurrence frequencies. Note that by definition the sum over all clusters for a specific month in a given simulation is one. Only those nine PAMIP models were considered here that are according to Fig. S4 able to realistically reproduce the ERA5 regime structures.
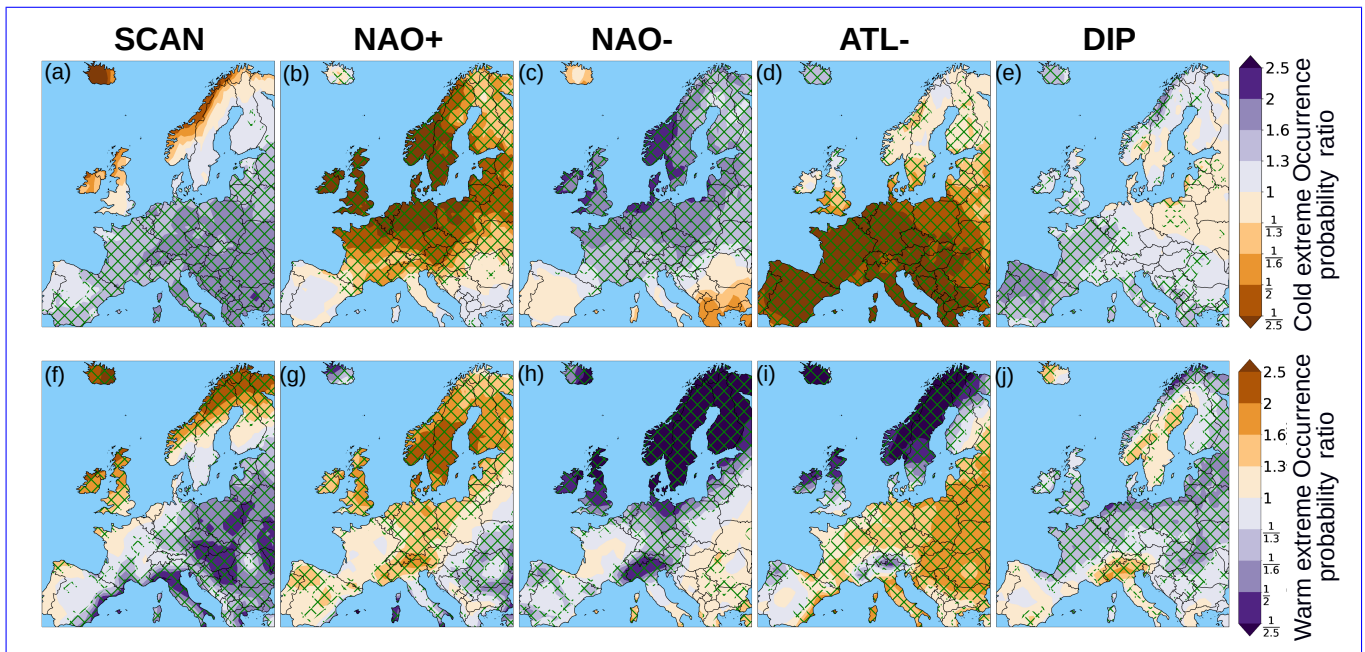
Figure 3. Temperature extreme occurrence probability ratios ~~(DJFM)~~ averaged over the extended winter season for different circulation regimes plotted over the European domain using the ECHAM6 PAMIP ~~pdSST/~~pdSIC/pdSST simulation. Upper row a-e: cold days, bottom row f-j: warm days. The plotted ratio compares the occurrence probability of an extreme day given a certain circulation regime to the unconditioned probability of an extreme occurrence. Thus, for instance violet values greater than one in a–e indicate a preferred cold extreme occurrence at a specific grid point ~~indicate preferred extreme occurrence~~ during the presence of a certain regime compared to the overall extreme occurrence. ~~Dotted~~ Note, that the color bar is reversed for warm extremes in f–j. Hatched areas indicate ratios that are significantly different from unity based on a moving bootstrap.

~~triangles) detrended Arctic sea ice conditions. Only ERA5 occurrence frequencies for months where significant differences between low and high ice conditions were found are shown here. Significant differences are derived from a moving block bootstrap. The absolute numbers of regime day counts in ECHAM6 and ERA5 that were used for the computation are shown in Tabs. S2–S4.~~

4.2 Links between certain circulation regimes and European temperature extremes in ECHAM6

After examining how the occurrence probability of certain circulation regimes can be affected by sea ice changes we now discuss which of the computed ECHAM6 circulation regimes can be associated with temperature extremes over Europe. For this reason, Fig. 3 compares the occurrence probability of temperature extremes given a specific circulation regime to the unconditioned probability of an overall extreme occurrence. Although this is only shown here for the ECHAM6 pdSIC/pdSST ~~/pdSIC~~reference simulation, results when using data from the sensitivity model experiments are qualitatively extremely similar. The general consistency with ERA5 (Fig. ~~S4~~S9) suggests that for most European regions ECHAM6 is able to realistically

represent the relevant physical processes that lead to the occurrence of temperature extremes during the respective regime days. Figure 3c indicates that the presence of a NAO- regime is associated with an up to more than doubled probability than usual of cold extreme days over large parts of mid- to northern Europe. This reported link between NAO- events and winter cold

380 spells or negative temperature anomalies over northern Europe is well-established and frequently observed in studies (Cattiaux et al., 2010; Andrade et al., 2012; Rust et al., 2015; Screen, 2017b). Figure ~~??c~~ S8c shows how NAO- event are related to easterly zonal wind anomalies which consequently lead to favored cold air advection of continental air masses towards northern Europe. These easterly anomalies can generally also be related to a suppressed advection of warmer maritime air masses, favoring colder conditions over Europe. As shown in Fig. ~~??b~~S6b, up to 40% of NAO- regime days are associated with

385 atmospheric blocking activity over Greenland and the North Atlantic. Blocking conditions over these region have previously been related to European winter cold spells as well (Sillmann et al., 2011).

In addition to the NAO- regime preferred occurrences of cold extremes over central and eastern Europe can be observed during SCAN days in Fig. 3a. Links between anticyclonic systems over Scandinavian/Ural regions and cold days over large parts of Europe have been reported previously (Petoukhov and Semenov, 2010; Andrade et al., 2012), since Scandinavian high

390 pressure system are typically associated with cold air advection towards central Europe from northeastern European regions (see Fig. ~~??a~~S8a). Indeed, Lagrangian backward trajectory analyses (Bieli et al., 2015) showed that cold events over mid- and eastern Europe are induced by horizontal advection of air masses from Russia and far northeastern regions. These advective processes are furthermore characterized by an adiabatic and steady descent of the air masses. Additionally, Fig. 3e indicates preferred cold extreme occurrences over most parts of western Europe during the presence of the Dipole regime. This link is

395 related to southward advection (see Fig. ~~??e~~S8e) of Arctic air masses especially from regions east of Greenland (Bieli et al., 2015).

Warm days in winter over large parts of central, eastern and southern Europe occur preferably during the presence of the ATL- regime (see ~~Figure~~ Fig. 3i). As shown in Fig. ~~??a~~S7a, around and westwards of the British Isles the ATL- regime is associated with enhanced baroclinic activity and consequently an intensification of the North Atlantic storm track. Therefore,

400 more storm systems than usual may form and advect warm and moist Atlantic air masses towards mid- and southern Europe. Complementary, warm days over northern Europe are linked to the presence of the NAO+ regime (see Fig. 3g). Such warm extremes over northern Europe are linked to strengthened westerly transport of moist Atlantic air masses during positive NAO events resulting in enhanced latent energy transport towards Scandinavia (Vihma et al., 2020). As shown in Fig. ~~??b~~ S7b this can also be related to a poleward shift of the North Atlantic stormtrack towards northern Europe and the Arctic.

405 **4.3 Sea ice induced changes in winter temperature extremes in ECHAM6**

The upcoming ~~Section~~ section investigates the overall changes in occurrence frequencies of continental northern hemispheric winter temperature extremes, which can be expected under future Arctic sea ice loss in ECHAM6. Therefore, Figs. 4 and 5 depict the overall occurrence ratio ρ of cold and warm extremes, comparing the extreme occurrence probability in the futArc-SIC and futBKSIC experiments with the reference simulation. Figure 4 indicates a general tendency towards less frequent

410 cold extreme occurrences in the future sea ice scenario simulations over the mid- to high northern latitudes. From a thermody-

namical perspective this observation is consistent with the fact that more open water areas and the associated elevated surface temperatures in the sensitivity runs provide an additional energy source to the atmosphere. However, the spatial pattern and the signals' magnitude strongly depend on the specific month and whether sea ice is reduced over the entire Arctic (see Fig. 4e-h) or just over the Barents/Kara Sea (see Fig. 4a-d). Although spatial tendencies show to some extent relatively similar patterns in both sensitivity simulations, futArcSIC exhibits much more pronounced reductions in cold extremes by a factor of more than 2.5 over high northern latitudes. Contrary, some parts over mid- and northern Eurasia show more frequent cold extreme occurrences in futBKSIC from January to March. This observation is consistent with the frequently reported Eurasian cooling response to sea ice loss in the Barents/Kara Sea (Cohen et al., 2018) that has been associated with a strengthening of the Siberian high. Over Europe significant reductions of cold extreme occurrences can be observed in futBKSIC in February (Fig. 4c), as well as in the futArcSIC simulation in February and March (Figs. 4g and h). Interestingly, January tends to exhibit slightly more cold extremes over central and eastern Europe in both sensitivity simulations (Figs. 4b and f).

As illustrated in Fig. 5 significant changes in the occurrence of warm extremes are generally less pronounced compared to cold extremes. Over Europe an overall tendency towards more frequent occurrences of warm extremes can be detected especially under diminished Arctic sea ice conditions in the futArcSIC simulation (Figs. 5e-h). In many regions and months, reductions in cold extreme occurrences are accompanied by increased probabilities of warm extremes. This might be associated with an overall thermodynamical shift of the underlying temperature distribution due to reduced sea ice concentrations and warmer surface temperatures in the sensitivity experiments. For futArcSIC this is e.g. the case over northern Siberia in December (Figs. 4e and 5e) or over Europe in March (Figs. 4h and 5h). However, several regions such as central Europe in February show for instance in futBKSIC reductions in cold extreme occurrences but no significant complementary changes in warm extremes (see Figs. 4c and 5c). Such asymmetric responses in the tails of the temperature distributions could be thermodynamically explained by a stronger warming of northerly polar winds compared to southerly winds as argued by Screen et al. (2014). Nevertheless, such responses could also be a result of other contributing factors, such as changes in the occurrence frequencies of atmospheric flows leading to certain extremes. In rare cases such as over central and eastern Europe in January, the futArcSIC experiment even shows an increased occurrence probability of both, cold and warm extremes (see Figs. 4f and 5f). This might be also related to an overall increase of temperature variability.

4.4 Decomposition of extreme frequency changes in ECHAM6

Now focus finally shifts back to temperature extremes over Europe. We try to understand to what extent sea ice induced changes in ECHAM6 extreme occurrences over Europe can be decomposed into ~~Fixed-Regime and Changed-Regime~~ fixed-regime and changed-regime contributions. Therefore, we now employ the conditional extreme event attribution framework described in Sect. 3.2. On the one hand, in Sect. 4.1.2 we identified ~~those~~ and compared significant regime frequency changes ~~in~~ found ECHAM6 ~~that are in agreement with~~ with other PAMIP models and recently observed ERA5 tendencies~~:~~. Based on these results and in order to discuss decompositions for different regime storyline we focus on the SCAN, NAO+ and ATL- regime frequency changes in January and/or February. On the other hand, in Sect. 4.2 we discussed how these regimes can be statistically and dynamically related to preferred occurrences of European temperature extremes.
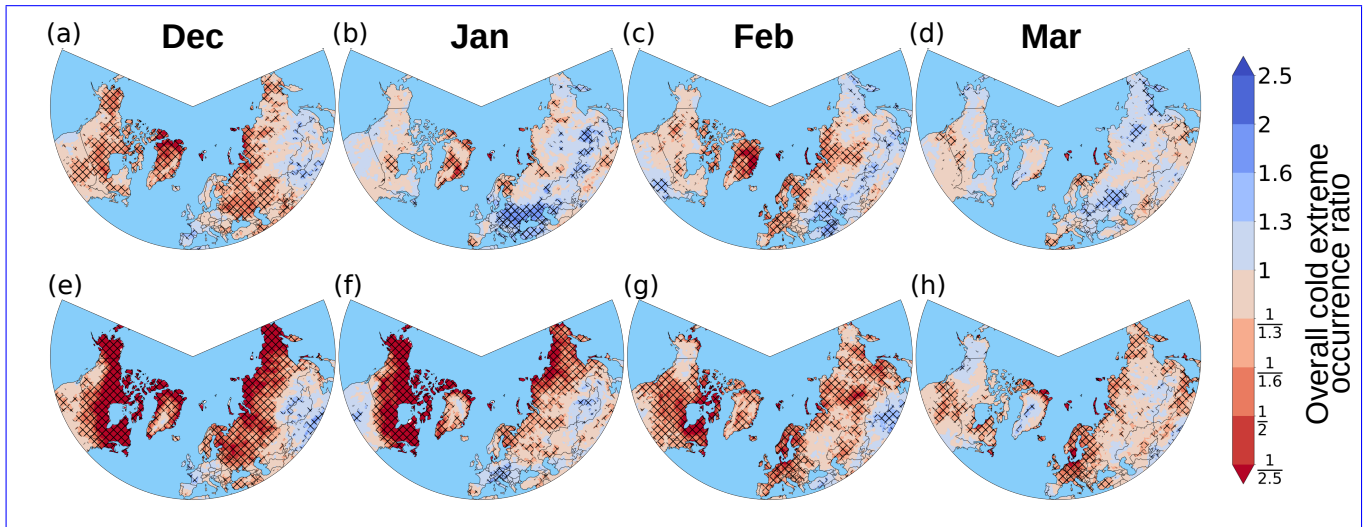
Figure 4. Cold extreme occurrence ratio for December, January, February and March in ECHAM6. The occurrence probability of northern hemispheric continental cold extremes are compared between the sensitivity experiments ~~(more frequent occurrences red)~~ vs. the ~~pdSST/~~pdSIC/pdSST reference simulation~~(more frequent occurrences blue)~~. Upper row (a)–(d): futBKSIC sensitivity run. Bottom row (e)–(h): futArcSIC sensitivity run. Blue indicates more frequent cold extreme occurrences, and red indicates less frequent cold extreme occurrences in the sensitivity experiments. Hatching indicates regions where the ratio differs significantly from unity based on a moving block bootstrap.
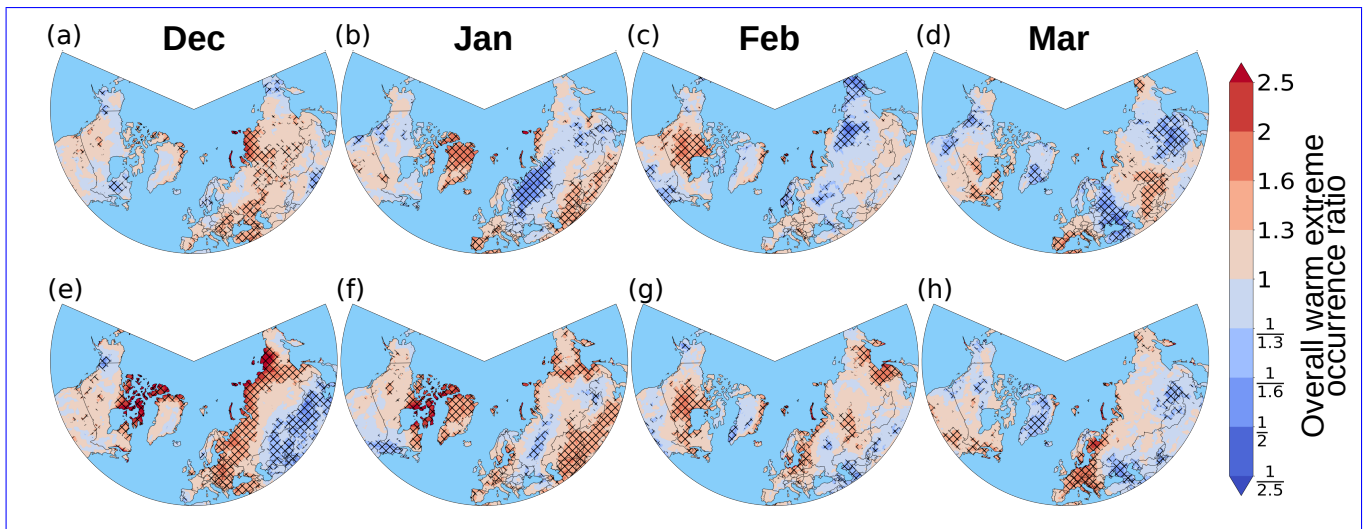
Figure 5. Same as in Fig. 4 but for warm extremes. Note that the colorbar is reversed compared to Fig. 4, such that red indicates more frequent warm extreme occurrences, and blue indicates less frequent warm extreme occurrences in the sensitivity experiments.
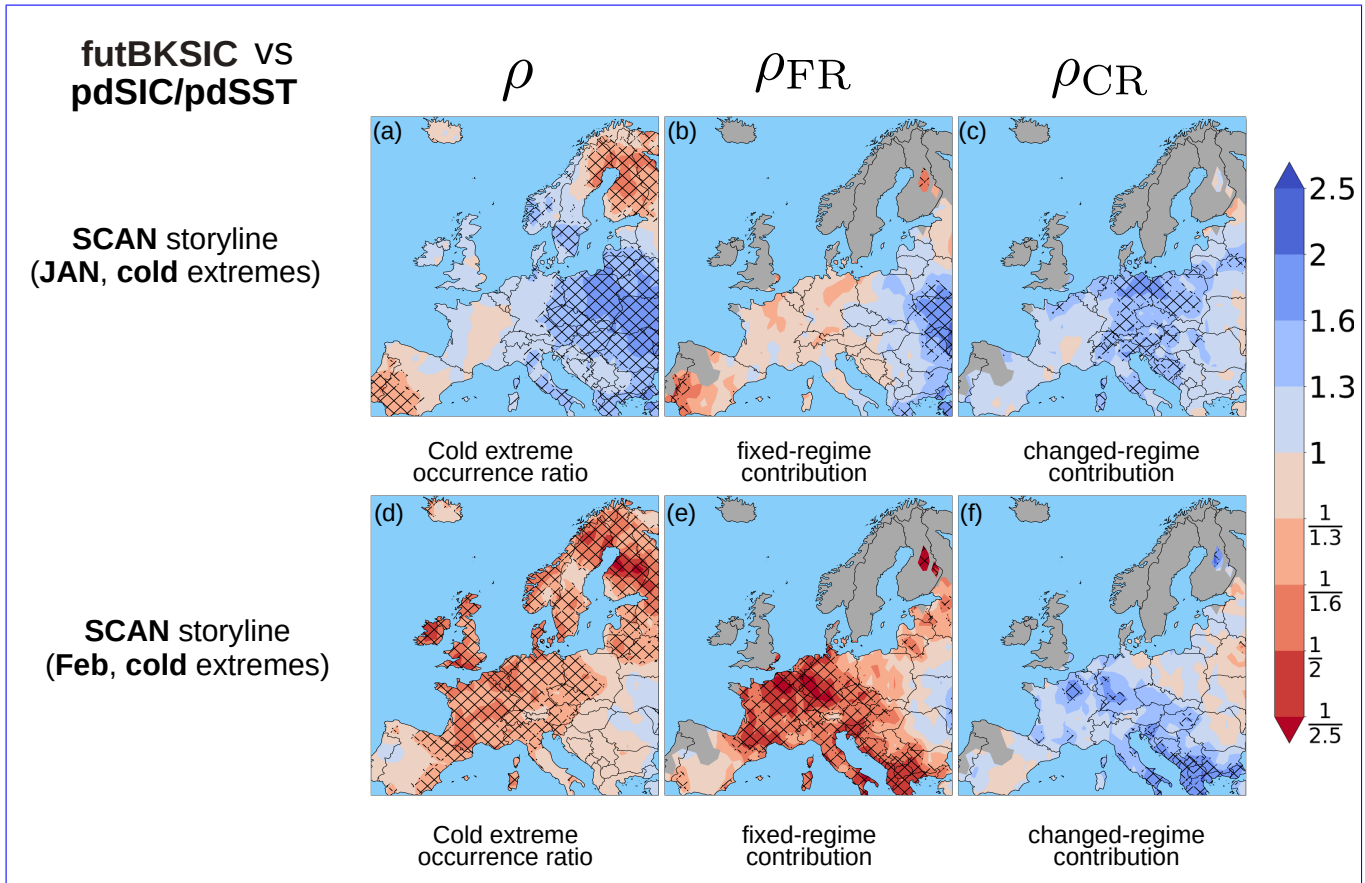
Figure 6. Conditional extreme event attribution framework for European cold extremes assuming a SCAN-storyline. Compared are the ~~pdSST/~~the futBKSIC sensitivity and the pdSIC/pdSST reference ~~simulation (blue~~ simulations. Blue indicates ~~favored occurrence)~~ more frequent cold extreme occurrences, and ~~the futBKSIC sensitivity simulation (~~red indicates ~~favored occurrence)~~less frequent cold extreme occurrences in futBKSIC. Upper row: January with $\rho_{\mathrm{circ}} = 1.26$. Bottom row: February with $\rho_{\mathrm{circ}} = 1.23$. ρ_{circ} greater than unity means that the SCAN regime occurs more frequent in the futBKSIC simulation for both months (see also Fig. ~~??f~~S2f). The first column shows the overall cold extreme occurrence ratio $\rho = \rho_{\mathrm{FR}} \cdot \rho_{\mathrm{CR}}$ between both simulations, the second column shows the ~~Fixed-Regime~~ fixed-regime contribution ρ_{FR}, the third one shows the ~~Changed-Regime~~ changed-regime contribution ρ_{CR}. Hatching indicates regions where the ratios significantly differ from unity based on a moving block bootstrap. ρ_{FR} and ρ_{CR} are only plotted for regions where statistically significant preferred winter cold extreme occurrences during SCAN days were identified in Fig. 3a.

445 The following decompositions of overall responses in extreme occurrences are considered here for the futBKSIC simulation: European cold extremes along a SCAN storyline in January and February, warm extremes along a ATL- storyline in January, as well warm extremes along a NAO+ storyline in February. Results for the futArcSIC simulation are shown in the Supplementary and are also discussed below. Only months for which significant changes in regime occurrence frequencies have been detected in ECHAM6 (see Sect. 4.1) are considered here, since the

450 physical interpretation of the changed-regime term ρ_{CR} strongly relies on significant changes in ρ_{circ}. The employed decomposition method assumes that the presence of the respective reference regime C_{ref} is necessary for an extreme to occur; hence, ρ_{CR} and ρ_{FR} are only plotted over regions where Fig. 3 indicates statistically significant more frequent extreme occurrences during C_{ref}.

Figure 6 shows the decomposition of the overall cold extreme occurrence ratio ρ between the futBKSIC sensitivity simulation and the pdSIC/pdSST reference experiment for January (Figs. 6a-c) and

455 February (Figs. 6d-f). The SCAN regime was chosen as the reference pattern C_{ref} since it could be associated with cold extremes over central, western and eastern Europe (Fig. 3a) and revealed significant frequency changes in the midwinter months as well (Fig. S2f). In January it shows that eastern and parts over central Europe are associated with significantly more frequent cold extremes in the futBKSIC simulation (Fig. 6a). The decomposition reveals that these signals can especially

460 over central Europe be associated with a significant contribution of the changed-regime ρ_{CR} term (Fig. 6c). This contribution is related to a 26% increase of SCAN regime occurrences in the futBKSIC simulation in January (see also Fig. S2f). Such a changed-regime contribution is however absent in more eastern parts of Europe, where the fixed-regime term ρ_{FR} significantly contributes to more frequent cold extreme occurrences (Fig. 6b). ρ_{FR} compares the extreme occurrence probability during SCAN days. Hence, it can not be ruled out that the individual daily flow

465 patterns allocated to the SCAN regime change in a way that they more frequently promote the occurrence of southwestward cold air advection towards Eastern Europe, and thus, also the occurrence of cold extremes over this region.

In February, strong frequency decreases of cold extremes over large parts of western, central and northern Europe can be observed in the futBKSIC simulation (Fig. 6d). In contrast to January, the predominant part of these overall changes is explained by the fixed-regime term ρ_{FR} (Fig. 6e). This might be interpreted as an overall thermodynamical

470 warming effect since more ice-free areas in the model simulations are typically associated with warmer surface temperatures and with overall stronger ocean–to–atmosphere heat fluxes. Such additional heat and energy sources provided to the atmosphere are finally distributed via the climatological mean circulation. As air masses from northeastern Europe and the Barents/Kara Sea frequently serve as source regions for advective processes leading to cold spells over central Europe (Bieli et al., 2015), an average warming of these reservoir regions may suppress the occurrence of cold extremes over Europe in the futBKSIC

475 simulation. As it can be seen for the changed-regime term ρ_{CR} in Fig. 6f, February frequency changes in SCAN occurrences basically tend to favor cold extremes over most parts of Europe. However, compared to the fixed-regime term ρ_{FR} (Fig. 6e) these signals are relatively small and non-significant over most areas.

The same analysis for January is illustrated in Fig. S10 but considers the futArcSIC simulation instead of the futBKSIC simulation. The overall cold extreme response (Fig. S10a) shows a significantly increased (decreased) probability of cold
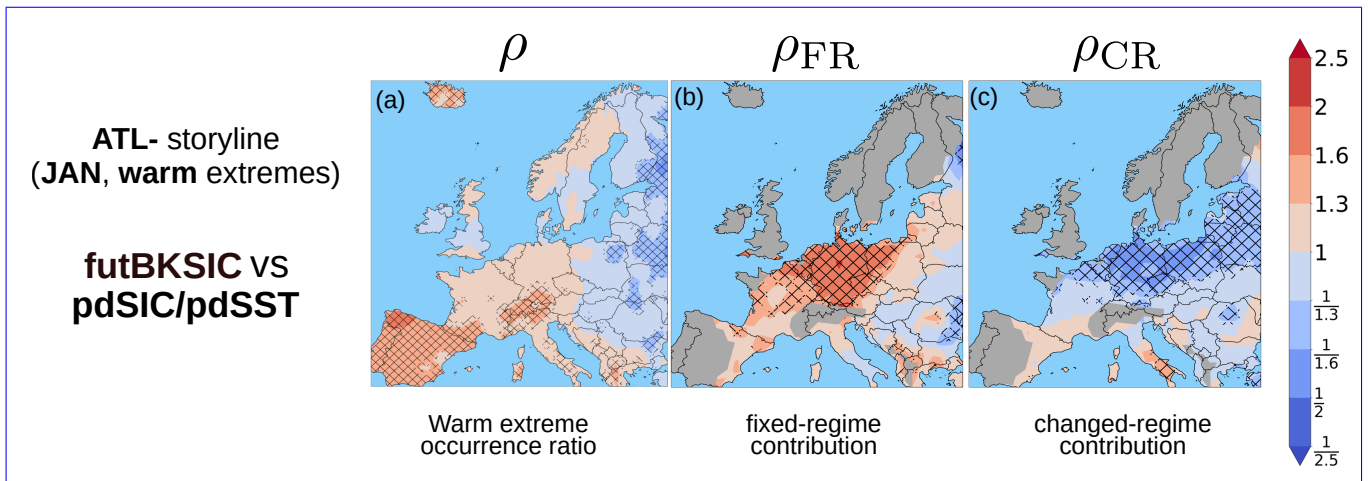
Figure 7. Same as in Fig. 6 (futBKSIC vs. ~~pdSST/~~pdSIC/pdSST simulation) but for January warm extremes and along a ATL- regime storyline. Compared to Fig. 6 the color bar is reversed, such that red indicates more frequent warm extreme occurrences, and blue indicates less frequent warm extreme occurrences in futBKSIC. Occurrence ratio of ATL- regime occurrence in January is given as $\rho_{\text{circ}} = 0.8$. Thus, the ATL- occurs less frequent in the futBKSIC simulation (see also Fig. ~~??i~~S2i). ρ_{FR} and ρ_{CR} are only plotted for regions where statistically significant preferred winter warm extreme occurrences during ATL- days were identified in Fig. 3i.

480   extreme occurrences over some parts of central (northeastern) Europe. The increased cold extreme probability over central Europe in Fig. ~~??a~~S10a shows how two non-significant contributions (Figs. ~~??b~~S10b and c) may add up to a significant overall response, whereas the decreased cold extreme probability over northeastern Europe is mostly explained by the ~~Fixed-Regime~~ fixed-regime term ρ_{FR} (Fig. ~~??b~~S10b).

     Figure 7 shows the decomposition for European warm extremes in January~~by~~, considering the ATL- regime as the reference
485   pattern C_{ref}. Here, the non-presence of significant signals in the overall warm extreme occurrence ratio over most parts of Europe (Fig. 7a) is especially over mid- and parts of eastern Europe a result of opposing ρ_{FR} (Fig. 7b) and ρ_{CR} (Fig. 7c) contributions. On the one hand, the reduced ATl- occurrence in the futBKSIC simulation can be associated with less frequent advections of warm air masses by Atlantic storm systems. On the other hand, an overall thermodynamical warming effect as mentioned before due to more open water areas tends to favor the occurrence of warm extremes.

490   A similar line of reasoning for January warm extremes along a ATL- storyline can be used in Fig. ~~??~~S11 where the futArcSIC simulation is considered and both contributions also appear to counteract each other. Here, an overall tendency towards more warm extremes can be observed over several parts of Europe compared to the futBKSIC simulation. This stems from a stronger dominance of the ~~Fixed-Regime~~ fixed-regime term ρ_{FR} (Fig. ~~??b~~S11b) probably due to a more pronounced thermodynamical forcing for Arctic-wide sea ice loss compared to sea ice loss over the Barents/Kara Sea only.

495   Figure 8 shows the decomposition for European warm extremes in February. The NAO+ regime is considered here as the reference pattern C_{ref}, since, on the one hand it can be associated with warm extremes especially over more northern parts of Europe. On the other hand it showed significantly less frequent occurrences in the futBKSIC simulation in February. The overall
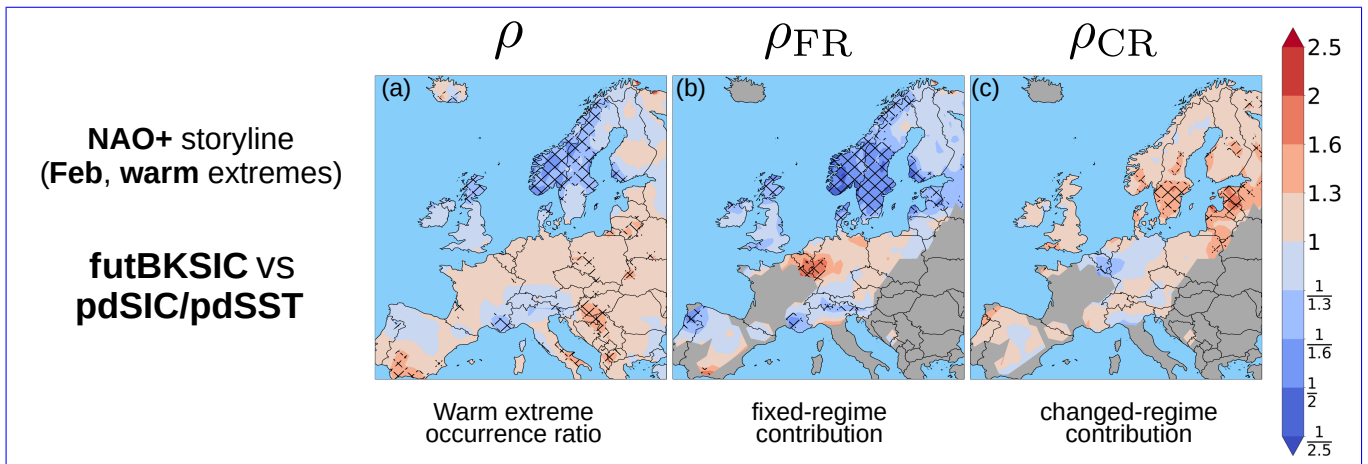
Figure 8. Same as in Fig. 6 (futBKSIC vs. ~~pdSST/~~pdSIC/pdSST simulation) but for February warm extremes and assuming a NAO+ regime storyline. Here, red indicates more frequent warm extreme occurrences, and blue indicates less frequent warm extreme occurrences in futBKSIC. Occurrence ratio of NAO+ regime occurrence in February is given as $\rho_{circ} = 0.8$. Thus, the NAO+ regime occurs less frequent in the futBKSIC simulation (see also Fig. ~~??g~~S2g). ρ_{FR} and ρ_{CR} are only plotted for regions where statistically significant preferred winter warm extreme occurrences during NAO+ days were identified in Fig. 3g.

warm extreme occurrence ratio ρ only shows some significantly less frequent extreme occurrences in the futBKSIC simulation over parts of Scandinavia (Fig. 8a). These signals are mostly explained by the ~~Fixed-Regime~~ fixed-regime contribution ρ_{FR} in
500   Fig. 8b. In contrast, ρ_{CR} only shows a relatively small significant contribution (Fig. 8c).

Finally, the previous results are contrasted to results for global SST changes in order to assess the relative importance of Arctic sea ice loss compared to a future increase of global SSTs. Therefore, Fig. 9 shows the overall response and the two contributions ρ_{FR} and ρ_{CR} for midwinter cold extremes comparing the ECHAM6 reference and the futSST simulation. The NAO- pattern was set as the reference pattern here. First, it shows that cold extremes occur massively and significantly less
505   frequent in the futSST simulation over all parts of Europe (Fig. 9a). Secondly, these overall changes are almost completely explained by the ~~Fixed-Regime~~ fixed-regime term ρ_{FR} (Fig. 9b). Although in this case the NAO- regime only shows non-significant changes between both simulations ($\rho_{circ} = 0.96$), even significant and more distinct changes in regime occurrences could not contribute in the same way as the ~~Fixed-Regime~~ fixed-regime contribution. Figure ~~9~~ 9 illustrates the decomposition of changes in extreme occurrences only for a NAO- storyline, but results for other storylines suggest the same qualitative
510   picture: the thermodynamical impact of globally increased SSTs dominates dynamical impacts related to regime frequency changes, regardless of the chosen reference regime. A similar picture is found for warm extremes. Therefore, we can conclude that although future sea ice loss is able to affect extreme occurrences over Europe, compared to future SST increases and for sure also to future global warming the effect is rather small.
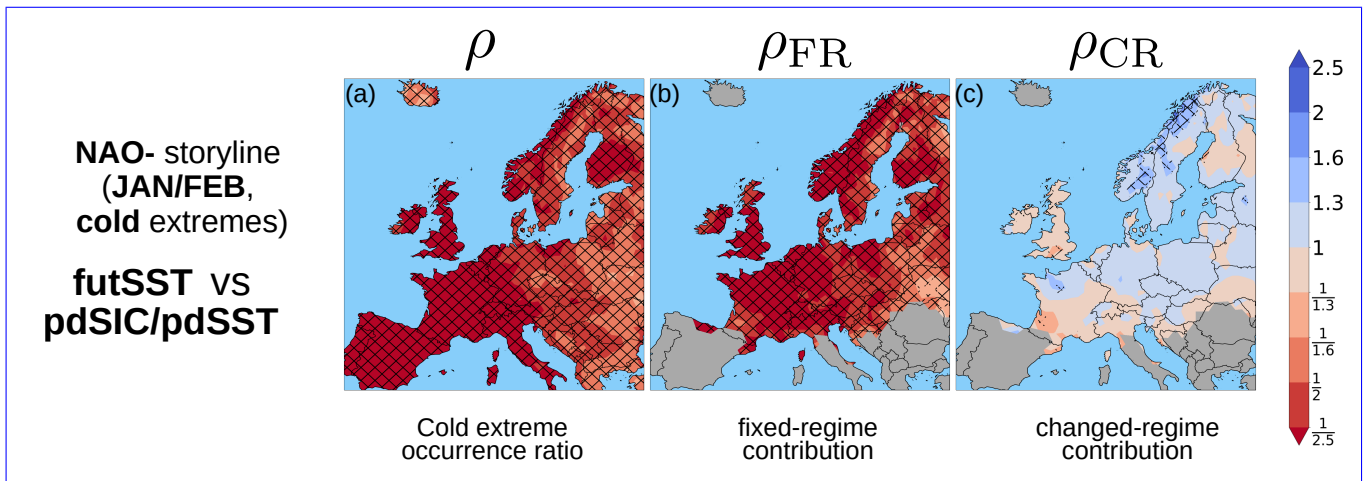
Figure 9. Similar to Fig. 6, but comparing the ~~pdSST/~~pdSIC/pdSST reference simulation and the futSST sensitivity simulation~~(blue~~. Blue indicates ~~favored~~ more frequent cold extreme occurrences~~in pdSST/pdSIC~~, and red indicates ~~favored~~ less frequent cold extreme occurrences in futSST~~)~~. Analyzed are cold extremes in January/February and a NAO- storyline is assumed here. ρ_{FR} and ρ_{CR} are only plotted for regions where statistically significant preferred winter cold extreme occurrences during NAO- days were identified in Fig. 3c.

5 Summary

515 The ~~general~~ aim of this paper was, first, to discuss how future Arctic sea ice retreat is able to impact large-scale atmospheric dynamics in terms of occurrence frequency changes of Euro-Atlantic circulation regimes, and secondly, to demonstrate how such regime frequency changes can be employed to decompose sea ice induced frequency changes in European temperature extremes into a dynamically motivated "~~Changed-Regime~~changed-regime" and a more thermodynamically motivated "~~Fixed-Regime~~fixed-regime" contribution. Therefore, for the most part we investigated data from ECHAM6 sea ice sensi-
520 tivity model experiments that are part of the PAMIP data pool. We considered simulations forced under future sea ice reduction over the entire Arctic, as well as only over the Barents/Kara Sea and compared them to a sensitivity simulation forced under present day conditions.

Analyzing ten additional PAMIP models, we initially studied how such future sea ice reductions affect the occurrence frequency of five Euro-Atlantic atmospheric circulation regimes that were computed with k-means clustering. Focusing on
525 ECHAM6, we afterwards discussed which circulation regimes can be associated with cold or warm extremes over Europe, and how the prescribed sea ice loss in the sensitivity simulations can impact the occurrence frequency of temperature extremes over the Northern Hemisphere. Based on the previous analysis steps, we employed a framework of conditional extreme event attribution and decomposed the overall sea ice induced ECHAM6 extreme frequency changes over Europe along suitable regime storylines. The decomposition of changes in extreme event frequencies finally yielded respective contributions, one
530 ~~Changed-Regime~~changed-regime contribution related to changes in the atmospheric circulation (changes in regime occurrence

frequencies), and another ~~Fixed-Regime~~ fixed-regime contribution that is related to increased surface temperatures during a specific atmospheric circulation regime.

The findings of the different analysis steps and research questions mentioned in the beginning can be summarized as follows:

- *Within the methodological framework of atmospheric circulation regimes, what changes in the wintertime atmospheric large-scale circulation over the Euro-Atlantic sector can be expected under future Arctic sea ice retreat?* As already motivated by Crasemann et al. (2017), we also detected significant changes in the occurrence frequency of winter circulation regimes when contrasting idealized atmosphere-only model simulations forced under present day and future Arctic sea ice conditions. ~~Seven~~ Most PAMIP models revealed an increase of Scandinavian blocking (SCAN) occurrences under future Arctic sea ice conditions in different winter months. ~~Furthermore,~~ Despite the finding that consistent with recent studies several models indicated more frequent occurrences of a NAO- pattern under future sea ice loss in mid-to late winter~~.~~

  , an overall disagreement between individual models on the sign of frequency changes of NAO+ and NAO- regimes predominates.

- *Which regimes can be associated with preferred occurrences of winter temperature extremes over Europe?* We showed and discussed that cold (warm) extremes over southern, central and eastern Europe occur significantly more frequent during SCAN (ATL-) days, whereas especially cold extremes over central- to northern Europe are on average significantly more frequently associated with negative (positive) NAO regime events.

- *What overall frequency changes of extreme occurrences over the continental Northern Hemisphere can be detected in response to future sea ice changes in ECHAM6?* We found that prescribed sea ice reductions in the ECHAM6 model simulations resulted in an overall tendency towards less cold extreme days, especially over high northern continental regions. A general tendency towards more warm extremes was less clear. However, the signal structures, their signs as well as their significances highly depend on the specific region and month. Finally we noticed that reductions in cold extreme occurrences are not necessarily accompanied by more frequent occurrences of warm extremes, and vice versa.

- *Based on the sea ice induced changes in circulation regimes detected in ECHAM6, to what extent can frequency changes of European extremes be related to ~~Fixed-Regime~~ fixed-regime and ~~Changed-Regime~~ changed-regime contributions?* The decomposition of overall responses of midwinter extreme occurrences in ECHAM6 revealed a rather complex picture. In several cases we could associate significant ~~Changed-Regime~~ changed-regime contributions related to occurrence frequency changes of certain regimes to preferred or unfavored occurrences of extremes. This was especially the case for increased January cold extremes related to increased Scandinavian Blocking occurrences, or decreased January warm extremes related to a reduced frequency of the ATL- pattern. Furthermore, we observed in several cases that the ~~Fixed-Regime~~ fixed-regime contribution yielded from a thermodynamical point of view intuitively expected decreased (increased) occurrence frequencies of cold (warm) extremes under future sea ice conditions. Finally, we noticed different scenarios for the resulting overall extreme occurrence frequency response. First, one contribution may dominate and re-

sults in a significant overall response. This was for instance the case for February cold extremes under a SCAN storyline where the overall reduced extreme occurrence frequency is explained by the ~~Fixed-Regime~~ fixed-regime contribution. Secondly, changes in regime occurrences may counteract the ~~Fixed-Regime~~ fixed-regime warming or cooling trend resulting in no detectable overall change in extreme occurrences. This was especially observed for January warm extremes under a ATL- storyline.

When analyzing changes in midwinter cold extremes induced by future raised global SSTs we detected a strong and significant decrease of cold extremes occurrences over entire Europe, especially when contrasted to results obtained for future sea ice reductions. Furthermore, this decrease was nearly completely explained by the ~~Fixed-Regime~~ fixed-regime contribution. This suggests a dominance of thermodynamical warming arguments over changes in atmospheric dynamics when trying to understand future changes in European temperature extremes. Overall these findings indicate that although future Arctic sea ice loss is for sure able to affect temperature extremes over Europe and the related atmospheric dynamics, the total effect size compared to globally raised temperatures that are expected in the future is relatively small.

6 Concluding remarks

Finally, we want to outline some potential prospects for future studies, as well as some limitations regarding the interpretation of the results that may arise from the specific model setup, methodologies and sample sizes used in this study.

First, the presented analysis for ECHAM6 was conducted based on 100 ensemble members of one-year-long time slice simulations for each respective experimental setup. In this respect, recent studies by Streffing et al. (2021); Peings et al. (2021); Sun et al. (2022) suggested that 100 ensemble members may not be enough in order to isolate the forced mean response from internal atmospheric variability in PAMIP sea ice sensitivity experiments. Furthermore, the results in Sects. 4.2–4.4 can differ for other PAMIP models, but ~~extending~~ conducting the decomposition method as applied in this study for each PAMIP model individually would be difficult: especially a comprehensive summary and interpretation of decomposition results for different models would be very challenging, in particular due to the fact that each model tends to simulate its distinct significant regime frequency changes in different months. Hence, the presented ECHAM6 analysis might be considered as a first step and adapting the employed decomposition methodology for a feasible implementation into a multimodel analysis might provide a prospect for future studies.

The question to what extent the detected winter changes in extremes or circulation regimes are a result of time-delayed stratospheric pathways triggered by sea ice loss in autumn cannot be answered with the presented methodology and experimental design. From the experimental side this would require more tailored model experiments as for instance done by Blackport and Screen (2019). They compared the delayed effect of autumn and year-round sea ice loss on the winter circulation by using coupled model experiments with modified albedo parameters. When only studying model experiments with prescribed year-round sea ice loss, more dynamical based analyses (e.g. Jaiser et al., 2016) have to be conducted in order to assess the role of stratospheric pathways and autumn sea ice loss. This was however not the focus of the present study.

Furthermore, we investigated atmosphere-only model experiments that do not allow for a representation of atmosphere–ocean feedbacks. In this respect, previous studies stressed the importance of an interactive ocean model (Screen et al., 2018). This may allow for representing additional oceanic pathways such as altered ocean currents that were shown to amplify circulation responses to Arctic sea ice loss. However, in contradiction to this hypothesis a recent study by England et al. (2022) shows that different approaches that impose sea ice perturbations in a coupled model setup add artificial heat to the Arctic region. This causes a spurious warming signal that is added to the warming expected from sea ice loss alone, and therefore finally results in an overestimation of the climate response to sea ice retreat in coupled model setups.

The atmospheric response to sea ice loss also depends on the exact prescribed patterns of sea ice and SST boundary forcing (Screen, 2017a; McKenna et al., 2018) and the model used. Crasemann et al. (2017) for instance studied sea ice sensitivity simulations conducted with the general circulation model for Earth Simulator (AFES, Nakamura et al. (2015)). Compared to the experiments used in our study their simulation data consisted of two perpetual runs over 60 years, however forced under sea ice conditions averaged over the early 80s and the early 2000s respectively. Additionally, their SST background states were set to the early 80s. With respect to circulation regime changes they detected an increase of the Scandinavian blocking pattern under low sea ice conditions in December, as well as a more frequent occurrence of the NAO- pattern in February and March.

The five circulation regimes that were used throughout the study only provide coarse categorizations of the atmospheric flow and contain a variety of more specific synoptic patterns. In the case of European winter temperature extremes we discussed that some of these few large-scale variability patterns might be suitable in order to describe the typical atmospheric circulation during such extremes, or at least contain most of the relevant synoptic patterns. The atmospheric situations during e.g. spatially confined precipitation extremes, as well as summer heatwaves that typically co-occur with an atmospheric ridge may be too unique and uncommon in order to be examined and allocated to a certain large-scale circulation regime. An analogue approach might be more suitable for such extremes.

The framework of conditional extreme event attribution employed in this study provides only one unique way to decompose atmospheric responses. The individual decompositions assume that the occurrence of a certain extreme can be completely associated with the presence and changes of a certain circulation regime. Studies by Vautard et al. (2016) or Cassano et al. (2007) proposed for instance an approach where the individual contribution terms related to specific regimes add up to the overall response. However, Vautard et al. (2016) also showed very limited suitability of this methods when working with a very small number of circulation regimes.

Furthermore, it should be noted that within this study we only considered changes in the occurrence probability of extremes defined by a fixed threshold temperature in a present day simulation. Similarly, changes in circulation regimes have also only been considered in terms of frequency changes. When aiming to draw conclusions about changes in the intensity and severity of extremes, other factors have to be taken account such as the actual strength of advection processes. Therefore, it might be helpful to distinguish between days that strongly (weakly) project onto a relevant pattern (e.g. NAO- for cold extremes) and are therefore connected to stronger (weaker) advective processes. This may provide an additional refinement possibility of the approach employed for upcoming studies.

630 In conclusion, the present study provides a complementary and useful perspective on the question how future Arctic sea ice retreat can impact large-scale atmospheric dynamics, as well as to what extent European temperature extremes are affected by future Arctic sea ice loss and how these changes can be separated into dynamically and thermodynamically contributing factors.

635 ~~Five circulation regimes over the Euro-Atlantic domain computed from daily ERA5 sea level pressure anomaly data (1979–2018) for extended winter season (December, January, February, March).~~

~~Taylor diagram (?) that summarizes different statistics in order to compare computed model patterns with regime patterns obtained from ERA5. Different symbols indicate different regimes and different colors stand for different combinations of model simulations for which regimes are computed in this study. The black star symbolically indicates the ERA5 reference~~
640 ~~pattern. The concentric quadrants centered around the origin show the pattern standard deviation of the different model patterns relative to the standard deviation of the ERA5 reference patterns. The blue polar axis depicts the pattern correlation coefficient between the respective model patterns and the reanalysis pattern. The green concentric semicircles centered around the black reference point indicate the centered root mean square error (CRMSE) when comparing model and reanalysis patterns. Thus, model symbols close the reference star mean high resemblance between model and reanalysis pattern.~~

645 ~~Mean DJFM relative blocking frequency (fraction of blocked days) at the same time a SCAN or negative NAO regime is present in ECHAM6 PAMIP pdSST/pdSIC simulation. Blocking frequency is calculated at a grid point level based on a hybrid, two-dimensional blocking index. Daily blocked grid points were identified based on the inversion of meridional gradients in the 500 hPa geopotential height (gph) field according to a modified version of the index from ?, and on areas of strong positive gph anomalies associated with the blocking detection. Finally, blocking events of a duration of at least 4 days and an area of~~
650 ~~$1.5 \times 10^6\,\mathrm{km}^2$ were selected by a subsequent tracking algorithm described in ?.~~

~~Synoptic-scale activity anomalies (DJFM) for the ATL- and NAO+ regimes computed from PAMIP pdSST/pdSIC model data. Synoptic-scale activity is computed here as the 2–6 day bandpass filtered standard deviation of slp data (?). It provides a measure for baroclinic activity and characterizes stormtrack locations. Only anomalies that significantly differ from zero are shown in colors.~~

655 ~~Wind anomalies at 700 hPa (DJFM) for the circulation regimes computed from PAMIP pdSST/pdSIC model data.~~

~~Same as in Fig. 6 but comparing the futArcSIC and pdSST/pdSIC simulations, and only for January cold extremes along a SCAN regime storyline. Occurrence ratio of SCAN regime occurrence in January is given as $\rho_{\mathrm{circ}} = 1.17$. Thus, the SCAN occurs more frequent in the futArcSIC simulation (see also Fig. ??a).~~

~~Same as in Fig. 6 but comparing the futArcSIC and pdSST/pdSIC simulations, and for January warm extremes along a~~
660 ~~ATL- regime storyline. Occurrence ratio of ATL- regime occurrence in January is given as $\rho_{\mathrm{circ}} = 0.79$. Thus, the ATL- regime occurs less frequent in the futArcSIC simulation (see also Fig. ??d). ρ_{FR} and ρ_{CR} are only plotted for regions where statistically significant preferred winter warm extreme occurrences during ATL- days were identified in Fig. 3i.~~

Data availability. ERA5 data have been stored and accessed via the Deutsches Klimarechenzentrum (DKRZ) in Hamburg. The ECHAM6 model simulations were conducted, originally stored and accessed at the DKRZ as well, but are also available for instance at the system of the Earth System Grid Federation (ESGF). Data of other PAMIP models are also available at the ESGF.

Author contributions. DH and JR developed the original idea for the paper. JR conducted the analysis and wrote the original draft. AR did the blocking frequency plots. AR, HR, UU and DH supervised and contributed to the interpretation of the results and provided feedback for the manuscript. TD did the ECHAM6 model simulation and also provided feedback for the manuscript.

Competing interests. The authors declare that they have no conflict of interest.

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