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Atmospheric circulation compounds anthropogenic warming and impacts of climate extremes in Europe

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1 **Diagnosing dynamical changes in the climate system, such as those**
2 **in atmospheric circulation patterns, remains challenging. Here, we**
3 **study 1950–2021 trends in the frequency of occurrence of atmospheric**
4 **circulation patterns over the North Atlantic. Roughly 7% of atmo-**
5 **spheric circulation patterns display significant occurrence trends, yet**
6 **they have major impacts on surface climate. Increasingly frequent**
7 **patterns drive heatwaves across Europe, and enhanced wintertime**
8 **storminess in the northern part of the continent. Over 91% of re-**
9 **cent heatwave-related deaths and 33% of high-impact windstorms**
10 **in Europe were concurrent with increasingly frequent atmospheric**
11 **circulation patterns. While the trends identified are statistically sig-**
12 **nificant at the p-0.05 level, they are not necessarily anthropogenic.**
13 **Atmospheric patterns which are becoming rarer correspond instead**
14 **to wet, cool summer conditions over northern Europe and wet winter**
15 **conditions over continental Europe. The combined effect of these**
16 **circulation changes is that of a strong, dynamically-driven year-round**
17 **warming over most of the continent and large regional and seasonal**
18 **changes in precipitation and surface wind.**

Climate Change | Atmospheric Dynamics | European Windstorms | European Heatwaves

1 **E**xtreme weather events exact a heavy and steadily in-
2 creasing socio-economic toll on Europe(1), eliciting both
3 scientific and media interest in the atmospheric circulation pat-
4 terns favouring the occurrence of heatwaves(2), cold spells(3),
5 heavy precipitation (4) and windstorms (5, 6). Such circula-
6 tion patterns may be understood as spatial patterns in a
7 given atmospheric variable which repeatedly occur in conjunc-
8 tion with specific classes of extreme events. The question of
9 whether, how and why these circulation patterns have been
10 changing under anthropogenic forcing, including the role of
11 Arctic Amplification, has been the source of a lively discussion
12 (7–25).

13 Of particular interest is whether atmospheric circulation
14 patterns favouring specific extreme events have become more
15 persistent. There are arguments supporting an increasingly
16 persistent summer atmospheric circulation (10, 16, 18), which
17 is highly relevant for the occurrence of heatwaves (19, 20).
18 Nonetheless, (23) did not find a systematic slowdown of plan-
19 etary waves over the Northern Hemisphere summers in the
20 historical period, and (26) found no significant changes in
21 their amplitude. There are also studies arguing against any
22 projected changes in summertime atmospheric persistence,
23 albeit on a regional scale (27). No consensus has therefore
24 been reached on the topic. The picture for the winter season
25 is equally debated, with some studies finding recent increases

in the sinuosity of the midlatitude flow (11) and others not
finding any trends in planetary wave phase speeds (23) or
amplitude (26), or arguing for a decrease in high-amplitude
waves and blocking — both regarded as persistent atmospheric
patterns associated with surface extremes (8). A complemen-
tary line of work has analysed the frequency of occurrence
of specific atmospheric circulation patterns associated with
extreme events (9, 13–15, 24). Heatwaves have again been a
focus, on account of their intimate link with the large-scale
atmospheric circulation and rapidly increasing frequency and
duration (25). For example, (9) found that well over half of
the total trend in hot extremes over Europe may be linked to
the increased occurrence of anticyclonic circulation patterns
over the eastern part of the continent. Similarly, (15) argues
that circulation patterns like the ones associated with the 2003
European heatwave may become increasingly frequent. Along
a similar line, (24) identified observed changes in hemispheric-
scale, wavy atmospheric circulation patterns that increase the
risk of concurrent heatwaves across eastern North America,
eastern and northern Europe and other downstream regions in
Eurasia. Specific atmospheric patterns associated with other
extremes, including drought (7) and heavy precipitation (21)
have also been studied. The two perspectives of persistence
and frequency of occurrence are intimately related, since re-

Significance Statement

We address the key question of whether and how the dynamics of the atmosphere may enhance the impacts of anthropogenic climate change. We specifically focus on temperature and surface wind extremes in Europe, two classes of events which have caused a high death toll and large insured losses over the continent in the recent decades. We find that large scale atmospheric patterns which favor summertime heatwaves and wintertime windstorms over large parts of the continent are becoming increasingly frequent. This effect sums to that of the average global climate change. A key implication of our work is that circulation changes modulate extreme events already in the present climate.

D.F. conceived the study, performed the bulk of the analysis and compiled the European windstorm database. G.M. performed the analysis on the EM-DAT database and contributed to compiling the European windstorm database. D.F. and G.M. wrote and revised the manuscript. D.F., G.M., A.J., M.V. and P.Y. read and integrated the manuscript. .

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50 currence of a specific pattern likely translates to persistent
51 weather and vice-versa (20).

52 Previous efforts in the detection of atmospheric circulation
53 shifts have often focused on the average behaviour (11, 16, 23),
54 for example presenting hemispheric or seasonal-mean results,
55 or on a specific set of extreme events or reference circulation
56 patterns (9, 15, 19, 24, 28, 29), for example the canonical
57 North Atlantic weather regimes. Here, we consider in turn
58 each daily atmospheric circulation pattern in 72 years of sea-
59 level pressure reanalysis data (30), over the period 1950–2021
60 and the Euro-Atlantic sector. This results in 26280 daily
61 sea-level pressure latitude-longitude maps. The same analy-
62 sis is repeated using 500 hPa geopotential height data (See
63 Supplementary Information). Our approach is thus distinct
64 from the above-cited studies since we look at individual at-
65 mospheric circulation patterns without needing to select a
66 priori a set of reference patterns. We then select those days in
67 the winter (December–February) and summer (June–August)
68 seasons displaying significant trends in the occurrence of their
69 analogues (hereafter referred to simply as "occurrence trends"
70 see Methods and Supplementary Information Figs. S1, S2).
71 The vast majority (92.7%) of circulation patterns show no
72 significant occurrence trend in the historical period; 5.1 %
73 show increasing trends and 2.2 % show decreasing trends.
74 Notwithstanding their rarity, the circulation patterns with sig-
75 nificant occurrence trends have major implications for surface
76 climate. To isolate the effect of circulation changes we also
77 consider daily reanalysis data for 10 meter horizontal winds,
78 2 meter temperatures, and precipitation rates, over the same
79 time-period and region. All the data shown in the composites
80 are detrended and deseasonalized (see Methods).

81 During both the boreal summer and winter seasons, the at-
82 mospheric circulation patterns with positive occurrence trends
83 show highly coherent surface climate anomalies over Europe.
84 Specifically, patterns occurring more frequently in winter (Fig.
85 1a) show a north-south cyclonic-anticyclonic dipole. This is
86 associated with anomalously windy, warm and wet conditions
87 in Northern and Eastern Europe and moderately warm, dry
88 conditions in Southern Europe (Fig. 1c, e, g). Wintertime
89 patterns with a negative occurrence trend (Fig. 1b) have a
90 cyclonic structure in the central North Atlantic and an anti-
91 cyclone to the Northeast, leading to anomalously windy, cool
92 and dry conditions over Northern Europe and windy, wet and
93 moderately warm conditions over Central and Southern Euro-
94 pe (Fig. 1d, f, h). We next analyse the average wintertime
95 temperature and precipitation associated with patterns dis-
96 playing either decreasing or increasing occurrence trends over
97 Europe (Fig. 1i, j; see mask in Supplementary Information Fig.
98 S3). We do not detect additional significant circulation-related
99 trends, besides the thermodynamics ones that have been al-
100 ready removed from the datasets (see methods). However, for
101 2 meter temperatures associated to patterns with increasing
102 occurrence trends, the average value of 1.8–1.9 °C (black solid
103 line, Fig 1i) is coherent with the large areas displaying positive
104 2 meter temperature anomalies in Fig. 1e).

105 In summer, the atmospheric circulation patterns with pos-
106 itive occurrence trends are associated with an anticyclonic
107 anomaly over the Labrador Sea and Greenland and a cyclonic
108 anomaly to the East of the British Isles (Fig. 2a). Such cir-
109 culation patterns drive anomalously calm and dry conditions
110 over much of the continent and anomalously warm conditions

over Western Europe (Fig. 2c, e, g). Circulation patterns with
decreasing occurrence trends show a ridge of high pressure
anomalies over the Atlantic and a large cyclonic structure
centred over Northern Europe. These are associated with
anomalously calm conditions over most of Europe and cool
and wet conditions over Northern Europe (Fig. 2d, f, h).
Again, no significant trends appear in summertime European
precipitation or temperature associated with any of the above
patterns (Figs. 2i, j).

We underline that these results are obtained for detrended
and deseasonalised datasets, meaning that the observed signals
are chiefly related to the atmospheric circulation. Analyses
conducted on NCEP/NCAR reanalysis data (Supplementary
Information Figs. S4, S5) and E-OBS gridded observational
data (Supplementary Information Figs. S6, S7) provide similar
conclusions. Supplementary Information Figs. S8–S11 further
show that the atmospheric circulation patterns with positive
or negative occurrence trends are generally similar at the
beginning and end of the analysis period. In other words, the
atmospheric circulation patterns identified as having significant
occurrence trends stay roughly constant over time. We further
find that the identified trends are robust to changes in the
details of our trend calculation (Supplementary Information
Figs. S12–S17).

We repeat the same analysis for the detrended 500 hPa
geopotential height data (Supplementary Information Figs.
S18, S19), to verify whether our results depend on the choice
of observable for the atmospheric circulation. There is a
strong resemblance of the spatial anomaly composites for
atmospheric circulation patterns with increasing frequency
of occurrence in sea-level pressure and 500 hPa geopotential
height, as well as the associated impacts on temperature,
winds and precipitation (cf. Figs 1a, c, e, g, 2a, c, e, g and
Supplementary Information Figs. S18a, c, e, g, and S19a, c,
e, g). Patterns with decreasing frequency of occurrence show
larger differences in both seasons (cf. Figs 1b, d, f, h, 2b,
d, f, h and Supplementary Information Figs. S18b, d, f, h
and S19b, d, f, h). This may be partly attributable to the
smaller sample size for patterns with decreasing occurrence
trends compared to patterns with increasing occurrence trends
(approximately 43% of the sample size, computed jointly over
winter and summer). It further suggests that our results
for these sets of days should be interpreted with care. We
conclude that the qualitative large-scale circulation patterns for
days showing increasing occurrence trends and the associated
surface anomalies are robust to the choice of the observable,
while some notable differences emerge for days with decreasing
occurrence trends. We estimate the magnitude of the trends for
days showing significant positive or negative trends as number
of analogues/decade for the different datasets and variables we
consider here (see Table S1). Overall, we find that increasing
(resp. decreasing) trends induce between 2 and 4 analogue
occurrences more (resp. less) every decade.

The surface climate anomalies associated with increasingly
or decreasingly frequent circulation patterns can be directly
related to the occurrence of high-impact summertime heat-
waves and wintertime stormy weather in Europe. We draw
high-impact heatwaves and the associated excess deaths from
the EM-DAT disaster database (31) (see Methods). We exam-
ine 228 heatwave days over the analysis period. 9.7% of these
days correspond to atmospheric circulation patterns increasing

172 in occurrence (versus a climatological summertime occurrence
173 of these patterns of 4.5% and a 97.5th percentile from ran-
174 dom sampling of 7.0%), while none correspond to decreasing
175 occurrence trends. As a term of comparison, only 3.1% of
176 480 cold spell days match atmospheric circulation patterns
177 increasing in occurrence, in line with climatology. The 228
178 heatwave days occur during 10 major heatwave episodes, as-
179 sociated with 83,462 deaths. Four of these heatwaves include
180 an above-average fraction of days with positive occurrence
181 trends (see Methods). The latter heatwaves are responsible for
182 91.4% of the total heatwave-related excess deaths (Fig. 3a).
183 Excluding from the analysis the summer 2003 heatwave, which
184 is associated with days with positive occurrence trends and
185 alone accounts for the bulk of the total heat-related deaths in
186 Europe in the period considered, we still find that heatwaves
187 associated with circulation patterns with positive occurrence
188 trends are responsible for 43.2% of total excess deaths. Al-
189 though this heuristic argument is based on a limited sample of
190 events, it nonetheless serves to illustrate the potential societal
191 impact that circulation patterns with increasing occurrence
192 trends may have, and motivates a future, more systematic
193 impact-based analysis.

194 We conduct a similar analysis for 90 European windstorms
195 which resulted in a high number of casualties and/or large
196 insured losses, extending the storm database from (32) (see
197 Methods and Supplementary Information Fig. S20). Over a
198 total of 438 windstorm days, we find that 66 (15%) are as-
199 sociated with circulation patterns with increasing occurrence
200 trends and only 4 (0.9%) with patterns with decreasing trends.
201 We find that 30 windstorms (33%) show an above-average
202 fraction of days with positive occurrence trends while 4 events
203 (4.4%) are linked to negative trends (Fig. 3b and Supple-
204 mentary Information Fig. S20b). These statistics aggregate
205 values on a continental level. A geographically-resolved pic-
206 ture highlights that the proportion of windstorms associated
207 with patterns showing positive occurrence trends is larger in
208 Continental/Northern Europe than in Southern Europe (Fig.
209 3b, e.g 40% in United Kingdom versus 0% in Italy). This
210 is consistent with the pattern shown in Fig. 1c,g), which
211 is reminiscent of the anomalies associated with destructive
212 windstorms over Continental Europe found by (5).

213 In this paper, we identified the atmospheric circulation
214 patterns over the North Atlantic and Europe that have become
215 less or more frequent in the historical period. We based our
216 analysis on all daily patterns in the data, rather than on
217 aggregated patterns, average behaviour or a specific set of
218 extreme events or reference circulation patterns (9, 11, 15,
219 16, 19, 23, 24, 28, 29). In other words, we do not constraint
220 our analysis to a fixed number of reference spatial patterns
221 of some atmospheric field, but rather view each atmospheric
222 pattern as unique and characterised by rare recurrences (a
223 few % of all data in our analysis). This approach does not
224 require introducing any specific decomposition bases and the
225 associated hyperparameters. We are thus able to isolate robust
226 linear trends in the frequency of occurrence of specific, observed
227 circulation patterns. Our results show minimal differences if a
228 quadratic or cubic fit is applied instead of a linear one. The
229 difference from previous studies is not only methodological, but
230 also interpretative. For example, we find increasing occurrence
231 of zonal flow patterns and decreasing occurrence of antizonal
232 patterns over the North Atlantic during winter, which do not

emerge from a conventional weather regime-based perspective
(33, 34).

Our method is flexible, and could for example be used
as a complement to extreme event attribution studies con-
ditioned on the circulation (35, 36), akin to the storyline
approach (37). The latter approach has been criticized for
not taking into account the role of climate change-induced
circulation changes (38), and we provide a readily applicable
toolkit to address this problem. Our approach is not data or
location-specific, and may be applied to different regions or
datasets. This makes it well-suited as a tool to evaluate the
consistency of dynamical trends in numerical simulations with
those observed in reanalysis data. In the future, our approach
could be complemented by the use of machine learning tech-
niques for identifying atmospheric analogues, as opposed to
using the Euclidean distance (e.g. 39).

Our analysis does not enable to make a robust state-
ment as to the potential role of natural variability versus
anthropogenically-driven climate change in modulating the
trends in atmospheric circulation pattern occurrence over the
recent decades. An analysis of climate modes of low-frequency
variability highlights a possible modulation of the El Niño-
Southern Oscillation and the North Atlantic Oscillation vari-
ability on trends in circulation pattern occurrence (see Sup-
plementary Information and Supplementary Information Figs.
S21-S23), which should be taken into consideration when in-
terpreting our results. Indeed, recent work has suggested that
the variability of the North Atlantic Oscillation can modu-
late decadal temperature trends (33). In future work, one
may also leverage initial condition large ensembles to separate
forced signals from internal variability in the context of our
analogue analysis (e.g. 40). A further limitation of our study
is that it assumes that the circulation patterns of interest
have good analogues in the dataset being used (see Methods).
This assumption is problematic in the presence of strong non-
stationarities, which may lead to unprecedented atmospheric
states. A theoretical advance would be to complement the
analysis with a systematic investigation of analogue quality,
based on the distance between a day and its analogues. Pre-
liminary analyses to this effect show no long-term trends in
analogue quality, yet highlight some interannual variability
which may relate to the above-discussed El Niño-Southern
Oscillation modulation. A second caveat is that the analysis
of surface impacts through the EM-DAT and the windstorms
database likely suffers from temporal and spatial inconsisten-
cies in reporting — a common limitation of impacts data (41).
In our specific case, the partial reliance of the windstorms
database on catalogues compiled for the scientific literature
(see Methods), could alleviate the issue, but by no means
completely resolves it.

We underscore that the temperature anomalies associated
with increasingly frequent circulation patterns (Figs. 1e,
2e) are regionally comparable to or larger than the average
global climate change (+1.07°C, best estimates reported the
IPCC(42)) with winter (summer) temperatures anomalies up
to +5.7°C (+1.1°C). These results are obtained after detrend-
ing and deseasonalising the data, and are consistent with
those obtained for the 500 hPa geopotential height in the
Supplementary Information.

The circulation types we study, while rare (approximately
7% of all winter and summer days in our dataset), are in-

creasing and decreasing at a pace of about 2-4 analogues per decade, depending on the variable and dataset examined (see Table S1). If these trends are due to anthropogenic forcing, we expect the circulation patterns with positive occurrence trends to dominate the respective seasons in 2-4 decades. If, instead, they are due to internal variability, we expect to continue seeing their footprint for the next few decades. Regardless of their origin, we stress that these circulation patterns are associated with impactful extreme events, both deadly heatwaves and destructive windstorms. While these events are not exclusively caused by circulations changes, our results show that circulation changes cannot be neglected when evaluating the consequences of anthropogenic climate change or other sources of climatic variability on extreme weather events and their impacts.

Materials and Methods

Computing trends in the occurrence of atmospheric circulation patterns. For the computation of the trends in the occurrence of atmospheric circulation pattern analogues (which we hereafter refer to as "occurrence trends"), we use daily sea level pressure and 500hPa geopotential height data from the ERA5 reanalysis (30) over the period 01/01/1950 – 31/12/2021 (26,280 days). This is the sample size for all analyses presented in the manuscript. The data have a horizontal resolution of $0.25^\circ \times 0.25^\circ$, and we restrict our analysis to 22.5N – 70N and 80W – 50E. This corresponds to the North Atlantic and Europe, with a size of 200×530 grid cells. Part of the analysis was repeated on additional datasets and subperiods of the full dataset (see Supplementary Information). For both the sea-level pressure and 500 hPa geopotential height data, we remove a grid point by grid point linear trend for the whole analysis period. This ensures that the selection of circulation analogues is not affected by long-term thermodynamic trends. We further deseasonalise the data using a mean seasonal cycle computed by averaging over the same calendar days. The robust trends in the occurrence of atmospheric circulation patterns are computed as follows:

1. We select daily sea-level pressure or 500 hPa geopotential height latitude-longitude maps, which we interpret as atmospheric circulation patterns.
2. We compute the Euclidean distance between daily maps, taking each daily map in turn as reference state and computing its distance from all other maps in the dataset. We then define a high quantile q to select the analogues. We chose $q = 0.98$, meaning that we take as analogues the 2% closest fields to the target. We describe below how the sensitivity to the choice of q is tested.
3. We divide the time interval of 73 years into 9 periods of roughly 8 years. We then count how many analogues N fall in each period t , obtaining $N(t)$ with $0 < t \leq 9$. Shortening these periods has no qualitative impact on our results (see Supplementary Figs. S12, S13).
4. We perform a linear fit of $N(t)$ of the type $at + b$. Using a cubic fit does not qualitatively affect the results (Figs. S14-S15).
5. We estimate the upper and lower 95% confidence intervals (CI) of the a parameter of the fit using the Wald method (43). If the lower and the upper bounds of the CI for a are positive (negative), we interpret this as a significant positive (negative) trend for the selected daily sea-level pressure or 500 hPa geopotential height map and quantile q . If the confidence interval contains zero, the trend is non-significant.
6. We repeat the above steps for $q = 0.99$ and $q = 0.995$. We retain as daily maps with significant increasing (decreasing) occurrence trends only those having consistent (same sign AND significant) occurrence trends for all three quantiles. These are the robust circulation patterns that are analysed in this paper. We additionally verify that the quality (i.e. distance) of analogues for these patterns is comparable to that for all other days in our dataset.

The 95% CIs that we compute indicate that there is a 5% chance that the real trend is out of the CI interval. Since the CIs are defined as symmetrical intervals, this implies that the chance that the "real" trend is zero or of opposite sign to that of our sample is $\leq 2.5\%$. In practice, since we use three different quantiles q simultaneously, this chance is very small, even taking into account that these are not independent samples. While analogues are a relatively common analysis approach in the atmospheric sciences, our methodology and its focus on trends is novel. It is further fundamentally different from conventional decompositions of the atmospheric variability, such as self-organising maps (44), k-means clustering (45), Empirical Orthogonal Functions or others (46). Indeed, we do not constraint our analysis to a fixed number of circulation patterns. The occurrence trends are thus identified for daily patterns rather than for a fixed pattern associated with, e.g. a self-organising map or weather regime. Indeed, the computation of the analogues does not require introducing any specific decomposition bases, which in turn introduce arbitrary hyperparameters such as the number of components into which the atmospheric variability is partitioned. Moreover, we do not "create" reference fields that may have never been observed in the data by orthogonal decomposition, averaging or centroid computations. Our approach thus relies directly on the atmospheric variability present in the data, without adding an intermediate projection step.

Computation of Significant Circulation and Surface Anomalies. After identifying the circulation patterns with robust increasing or decreasing occurrence trends for the whole year, we produce composite anomalies for the patterns on days within the winter (DJF) and summer (JJA) months (without including their analogues) for several daily variables: sea-level pressure, 500 hPa geopotential height, horizontal 10m wind speed and precipitation rate. All these data are detrended and deseasonalised using the same procedure as for sea-level pressure and 500 hPa geopotential height. Significance for the geographical anomalies is estimated by using a bootstrap procedure (sample size $m = 500$) consisting in randomly drawing from the whole dataset a number of days equal to the number of days in each composite, regardless of their trend. Significant anomalies are those below the 5th or above the 95th percentiles of the bootstrap distribution at each grid point. We have also tested gridpoint by gridpoint sign agreement of the days within each composite, to ensure that the composite anomalies result from a spatially coherent set of daily anomalies (S16-S17). Long term trends shown in the composites of Figs. 1-2 and Supplementary Information Figs. S4-S19 are computed by averaging, for each season, the spatial averages over Europe (see mask in Supplementary Information Fig. S3) for the patterns displaying occurrence trends. Years without patterns displaying occurrence trends are treated as NaNs. Significance for trends is estimated by using the Wald method and looking at the sign of both the upper and lower bounds of the CI. CIs of trends are displayed in the legends of the relevant figure panels.

Computing impacts of atmospheric circulation patterns with positive occurrence trends. We take heat waves and cold spells from the EM-DAT disaster database (31). We focus on events over Western-Central Europe (excluding e.g. Western Russia, outlying islands such as the Canary Islands etc.), excluding those heat waves and cold spells where no start and/or end day was provided, or which lasted only for one day. Dates included in several events were only counted once. Following these criteria, from a total of 34 heatwave and 66 cold spell episodes included in EM-DAT, we identified 228 days over 10 heatwaves and 480 days over 16 cold spells. Of the original 34 heatwave events, 3 were excluded because they were entirely outside Europe, 11 because they were missing start and/or end dates, 9 because a single day was assigned to the heatwave. Finally 2 heatwaves were merged into 1 because, even though they had different "disaster numbers" (i.e. unique identifiers of each disaster in EM-DAT), they had overlapping dates and occurred in geographically contiguous regions. Of the 66 cold spells, a relatively large number was excluded due to being outside of Europe (13 events were only registered in Russia). An additional 35 were excluded due to missing or single-day dates. Finally, four events were merged into 2 because, even though they had different "disaster numbers", they had overlapping dates and occurred in geographically contiguous

434 regions.

435 European Windstorms data is taken from an updated version
436 of the storm database of (32), which originally covered the period
437 1948–2015. This database is largely based on the catalogues by
438 (47) (ending 1972) and (48) (ending 2014). Additional windstorms
439 (2015–2020) have been integrated from the Wikipedia web-page
440 https://en.wikipedia.org/wiki/List_of_European_windstorms because of
441 their relevance in terms of human losses, damages or their profile
442 in the media. We have only selected windstorms that have been
443 analyzed by a meteorological office or research institute: a link
444 to the documentation is provided together with each entry in
445 the database. The database itself is available as Supplementary
446 Data. Using publicly contributed databases as sources for scientific
447 information is becoming a common practice in citizen science
448 projects, which are gaining momentum in the geosciences (49, 50).
449 Our database includes a total of 438 windstorm days and 90 distinct
450 windstorms or windstorm clusters. Supplementary Information Fig.
451 S20 presents the database entries per country (a) and per time of
452 occurrence (b).

453
454 The database is organized in four columns:

- 455 1. The day of occurrence in the format `yyyymmdd`;
- 456 2. The name(s) of the windstorm;
- 457 3. The country/countries or the region(s) affected;
- 458 4. A reference to a peer-reviewed article, a report or a press article
459 describing the importance of the windstorm.

460 As a caveat to our methodology, we note that the increasing
461 coverage of both meteorological instruments and technological means
462 of information result in an increasing number of windstorms with
463 time.

464 To determine whether a given extreme event is associated to
465 circulation patterns with positive or negative occurrence trends, we
466 first compute the fraction of days with positive occurrence trends
467 within each extreme event. We then compare these fractions to
468 the average frequency of occurrence of circulation patterns with
469 positive or negative occurrence trends within the 20 years centered
470 on the extreme. This is because the frequency of occurrence of
471 atmospheric circulation patterns with positive occurrence trends is
472 by definition higher in the later part of the data than in the earlier
473 years, such that each extreme should be compared to the period
474 within which it occurs. Based on this comparison, we then separate
475 extreme events into groups which have an above or below-average
476 fraction of daily analogues with either trend.

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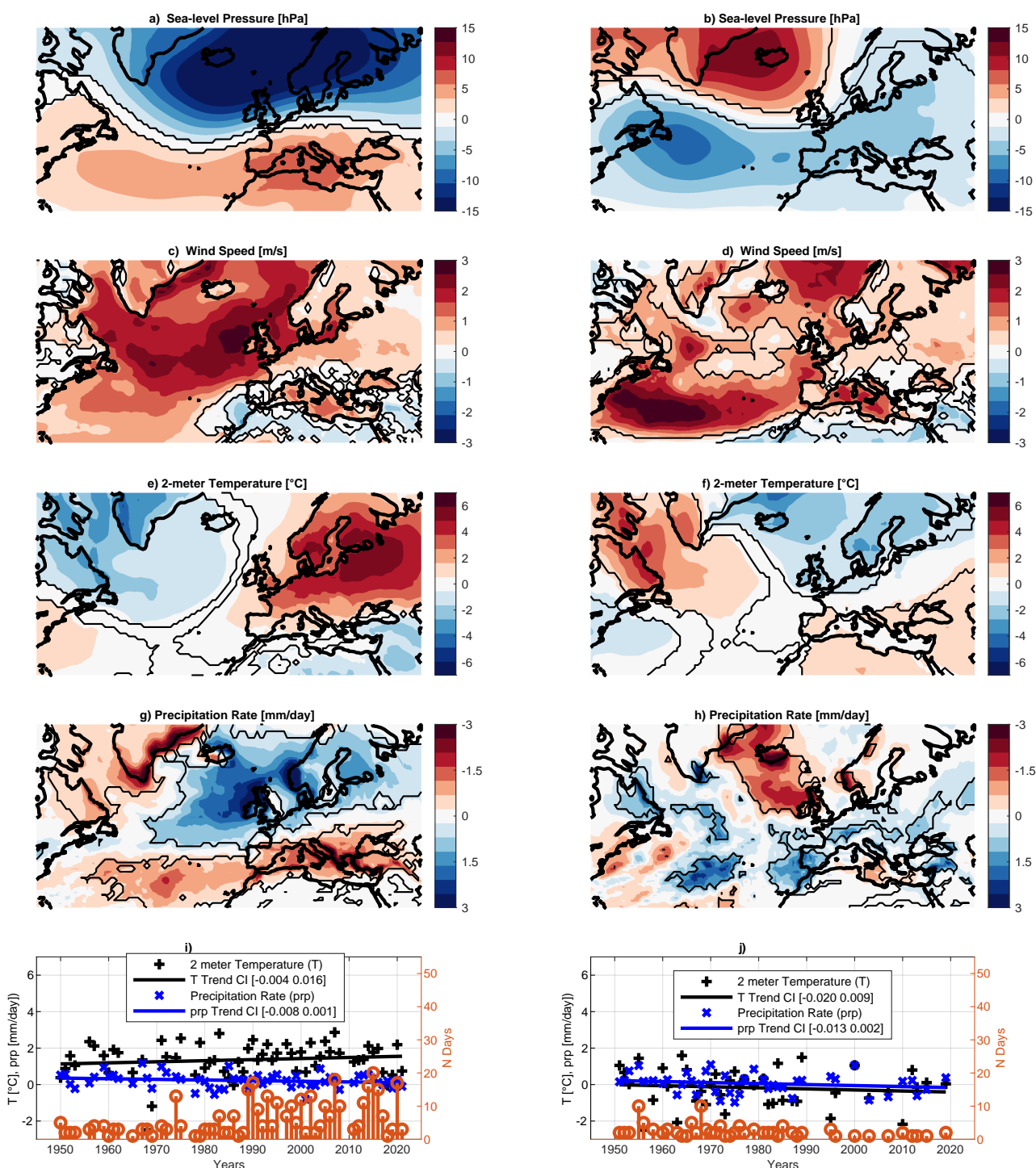


Fig. 1. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: Composite anomalies of DJF sea-level pressure (a,b), 10m horizontal wind speed (c,d), 2-meter temperatures (e,f) and precipitation rates (g,h) for days with increasing (a,c,e,g) or decreasing (b,d,f,h) occurrence trends. In the composites (a–h), contours indicate regions with changes significant at the one-sided 5% level, computed with a bootstrap sample size of 500. Spatial averages of seasonal temperature anomalies (black) and precipitation rates (blue) during the days with increasing (i) or decreasing (j) occurrence trends and count of days displaying the corresponding occurrence trend (orange stems) during DJF. Solid lines represent linear trends of the spatial averages with the 95% confidence intervals of the two linear fits in each panel shown in the legends. The averages in (i), (j) are computed on all European land points (see Supplementary Information Fig. S3).

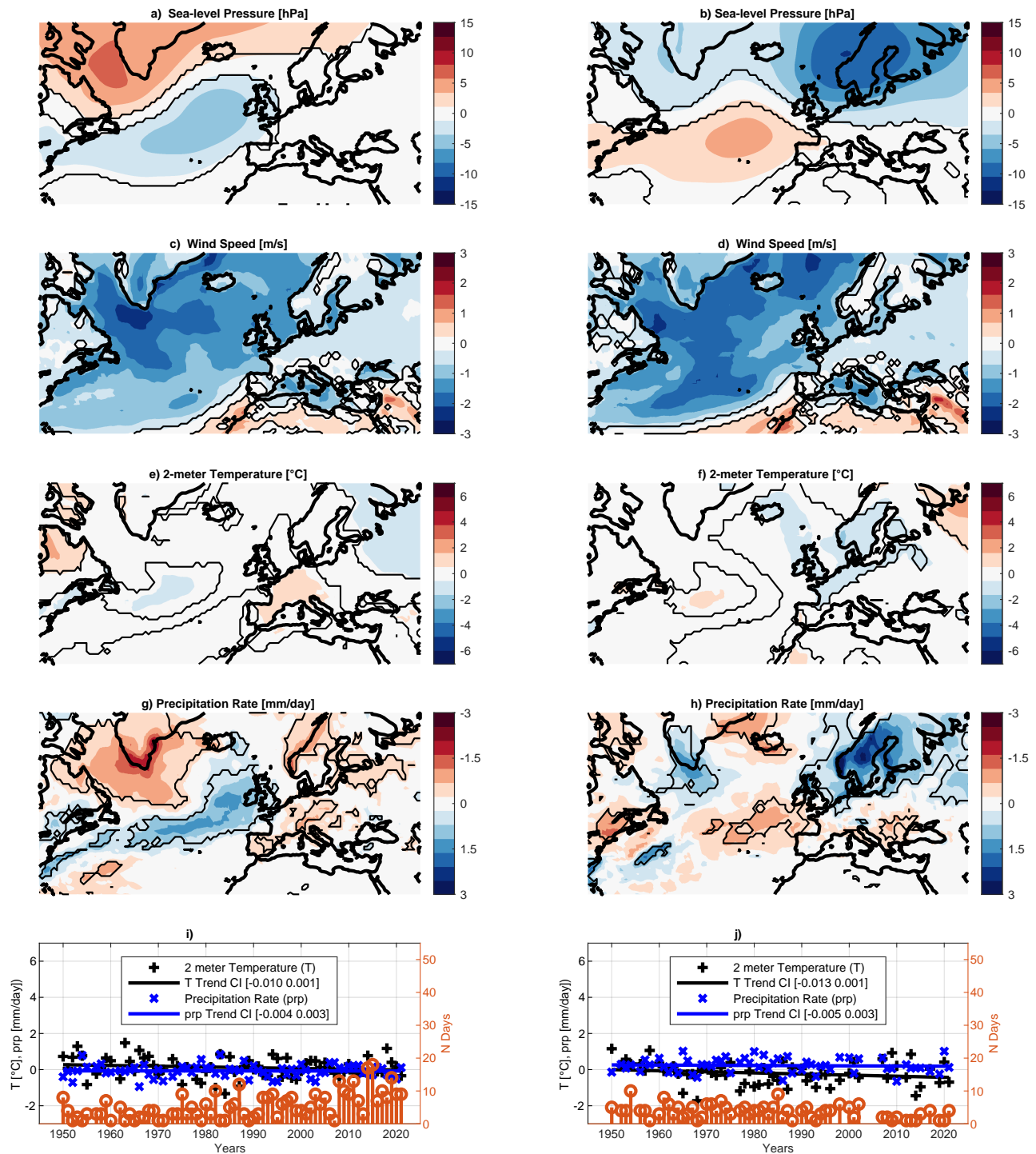


Fig. 2. Sea-level pressure summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for JJA.

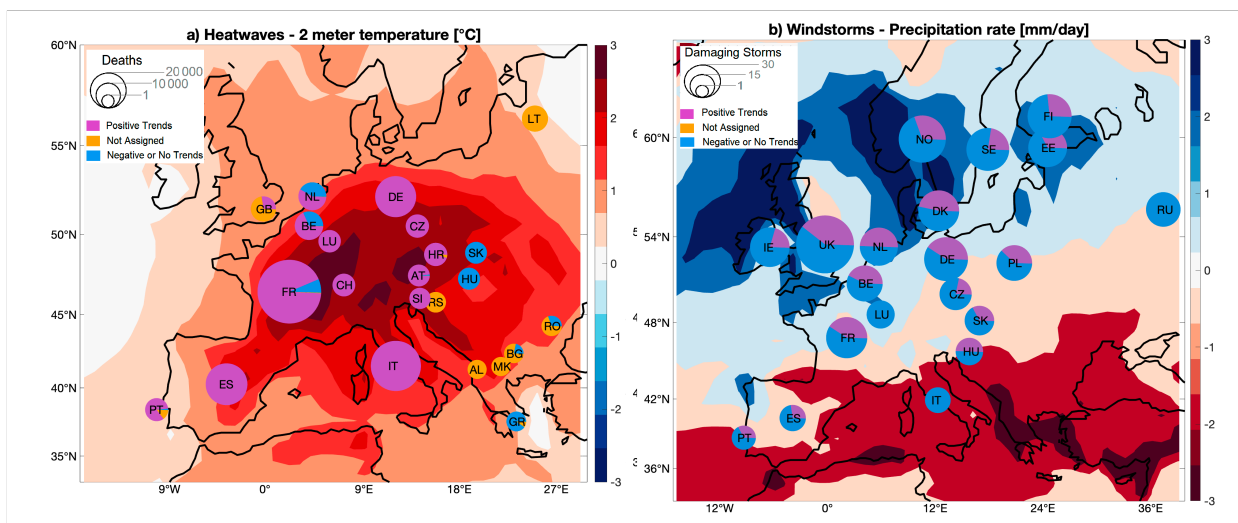


Fig. 3. Impacts of changing atmospheric circulation patterns in terms of heatwave casualties and European Windstorms. (a) The size of the pie chart for each country shows the total number of heatwave excess deaths as recorded in the EM-DAT database for the whole year. The purple slices show the fraction of excess deaths associated with heatwaves showing an above-average frequency of circulation patterns with a positive occurrence trend. The yellow slices show the excess deaths associated with heatwaves that we excluded from our analysis (see Methods). Finally, the blue slices show the corresponding fractions for heatwaves showing circulation patterns with no or negative occurrence trends. The shading on the geographical map shows the temperature anomalies ($^{\circ}\text{C}$) during the 228 heatwave days retained for analysis. (b) The size of the pie chart for each country shows the total number of destructive windstorms in the storm database for the whole year (see Methods). The purple slices show the fraction of windstorms showing an above-average frequency of circulation patterns with a positive occurrence trend. The blue slices show the corresponding fraction for windstorms showing analogues with no or negative occurrence trends. The shading on the geographical map shows the precipitation anomalies (mm day^{-1}) during the 438 windstorm days retained for analysis.

1

2 **Supporting Information for**

3 **Atmospheric circulation compounds anthropogenic warming and impacts of climate extremes** 4 **in Europe**

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8 **This PDF file includes:**

9 Figs. S1 to S23

10 Table S1

11 SI References

12 Use of additional datasets

13 In addition to the ERA5 data used in the main text, we have repeated the computation of occurrence trends of atmospheric
14 circulation patterns using sea-level pressure from the NCEP/NCAR reanalysis (1) (Figs. S4, S5) and 2m-temperature and
15 precipitation from E-OBS v25.0e over land (Figs. S6, S7). We use both datasets over the same period as in the main analysis,
16 namely 01/01/1950 – 31/12/2021. The NCEP/NCAR data has a horizontal resolution of $2.5^\circ \times 2.5^\circ$, while the E-OBS v25.0e
17 data has a horizontal resolution of $0.25^\circ \times 0.25^\circ$. We restrict our analysis to the same 80W – 50E and 22.5N – 70N domain
18 used in the main text. Figures S4, S6 and S5, S7 correspond respectively to Figs. 1 and 2 computed for ERA5. Allowing
19 for the different spatial resolutions of the datasets, which result in some modulation of the intensity of the anomalies, the
20 corresponding patterns in all variables and seasons closely resemble each other. The fact that very different datasets provide
21 the same conclusions supports the robustness of our analysis.

22 Analysis on sea-level pressure and 500 hPa geopotential height

23 We identify days with significant occurrence trends using analogues computed on either sea-level pressure or 500 hPa geopotential
24 height. In winter, we note a very close resemblance of the 500 hPa geopotential height and sea-level pressure large-scale patterns
25 for days with increasing occurrence trends. In both cases, the anomaly structure is an NAO-like meridional dipole favouring an
26 intensified zonal flow. This leads to very similar surface anomalies (cf. Figs. 1a, c, e, g and S18a, c, e, g). For the wintertime
27 decreasing occurrence trends, the 500 hPa geopotential height pattern has a marked meridional and quadripolar structure,
28 while the sea-level pressure pattern is chiefly dipolar and anti-zonal. The surface anomalies over Europe are regionally different,
29 although especially for precipitation some of the large-scale features are comparable (cf. Figs. 1b, d, f, h and S18b, d, f, h). A
30 similar picture emerges for summer (Figs. 2 and S19). The large-scale patterns for days with increasing occurrence trends
31 are broadly similar: both variables display a mid-Atlantic trough and a zonally extended band of positive anomalies in the
32 northernmost part of the domain. We nonetheless note a meridional shift in the anomaly poles and the fact that 500 hPa
33 geopotential height shows comparatively larger positive anomalies over Scandinavia. The patterns for decreasing occurrence
34 trends again show larger differences. In the case of sea-level pressure, there is a dipolar pattern with the positive pole to the
35 North of the Azores and the negative pole over Scandinavia. The 500 hPa geopotential height anomaly pattern is wave-like.
36 This is to some extent also mirrored in the surface anomalies (cf. Figs. 2b, d, f, h and S19b, d, f, h). The scale would appear
37 comparable to the wave-5 pattern of (2), although the surface temperature anomalies do not match those of the latter study.
38 The pattern also resembles the Node 7 500 hPa geopotential height pattern shown in (3), although the latter study finds that
39 this pattern has a significantly increasing occurrence trend over the historical period, contrary to what we find in our analysis.
40 The regional North Atlantic view we take may explain part of the discrepancy with the circumhemispheric wave view of (2, 3).

41 Modulation of circulation pattern occurrences by natural variability

42 The atmospheric circulation pattern occurrence trends we identified in our analysis may be modulated by natural variability.
43 To investigate this, we have analysed the correspondence between atmospheric circulation pattern occurrence trends and the El
44 Niño–Southern Oscillation (ENSO) as diagnosed through the Niño 3.4 index, the Atlantic Multidecadal Oscillation (AMO) and
45 the North Atlantic Oscillation indices. Specifically, we have tested whether days with positive or negative occurrence trends
46 correspond to anomalous ENSO, AMO or NAO values and whether high or low ENSO, AMO or NAO values favour positive or
47 negative occurrence trends. For ENSO and AMO, we have conducted the analysis using monthly indices computed from the
48 NOAA/ERSSTv5 dataset (4) and for NAO, we used the monthly NOAA Climate prediction center teleconnection index (5)
49 retrieved from KNMI’s climate explorer. The Niño 3.4 timeseries is normalised over the 1981–2010 period, while the AMO
50 timeseries is normalised over the 1971–2010 period. For the NAO index, the standardized anomalies are calculated based on
51 the 1950–2000 climatological daily mean and standard deviations. Thus none of these indices have a zero average over the
52 analysis period we consider. The Niño 3.4 index has shown an upward trend in recent decades (6, 7), something which we
53 also see in our data. The AMO and NAO indices also show a weak trend. We have therefore conducted our analysis on both
54 the raw and linearly detrended timeseries (the latter being centred on zero due to the detrending process). There is a clear
55 association between ENSO variability and atmospheric circulation pattern occurrence trends. Niño 3.4 values are significantly
56 positive (negative) for days with positive (negative) occurrence trends (Fig. S21a,b). Similarly, days with positive (negative)
57 occurrence trends show a clear preference for months in the upper (lower) tertiles of the Niño 3.4 distribution (Fig. S21c,d).
58 The above holds for both the raw and detrended Niño 3.4 timeseries, although the results in Fig. S21c,d are stronger for the
59 undetrended ENSO timeseries. We thus conclude that there appears to be a statistical modulation of ENSO variability on
60 trends in the occurrence of atmospheric circulation patterns in the North Atlantic sector, and that the association between
61 positive (negative) ENSO phases and days with positive (negative) occurrence trends is stronger if the long-term ENSO trend is
62 taken into account. Focussing on days with positive occurrence trends, during winter we hypothesise that this may be linked to
63 the NAO-like footprint of ENSO on North Atlantic sea-level pressure (8). During summer the picture is not as clear, and recent
64 work has found only weak relations between ENSO and the summer NAO (9). For the NAO index itself, we only find a clear
65 association with days with increasing occurrence trends (Fig. S22c), which tend to be associated with positive NAO phases.
66 This is expected on the basis of the pattern obtained in Figure 1a as it consists of a strong zonal flow typically associated with
67 positive phases of the North Atlantic Oscillation index. No significant association between atmospheric circulation pattern
68 occurrence trends and the AMO has been instead detected (Fig. S23).

69 Notwithstanding the possible low-frequency modulation of the atmospheric circulation pattern occurrence trends by internal
70 climate variability, the patterns with increasing or decreasing occurrence trends that we identify appear stable throughout the
71 analysis period. Figs. S8–S11 show the occurrence trend analysis for the first or last 20% of circulation pattern occurrences
72 computed using the full reference period. With the exception of circulation patterns with decreasing occurrence trends in
73 winter, which have a very small sample size, there is very little difference between the two timeperiods considered.

74 **Magnitude of trends per decade**

75 Table S1 shows the magnitude of the trends for days with significant occurrence trends, estimated for each dataset or variable
76 whose analysis is reported in Figs. 1, 2, S4–7, S12–15 and S18–19. The trends are averaged over the different quantiles used for
77 the analysis and are always expressed in terms of number of analogues/decade. There is a close match between ERA5 and
78 NCEP trends for sea-level pressure and no difference when trends are estimated on ERA5 with a cubic rather than a linear fit.
79 When increasing the number of periods from 9 to 14 (that is from 8 years to 5 years bin lengths), the absolute value of trends
80 slightly reduces. Trends for 500 hPa geopotential height are larger in absolute values. Overall we have that increasing (resp.
81 decreasing) trends induce between 2 and 4 analogues more (resp. less) every decade.

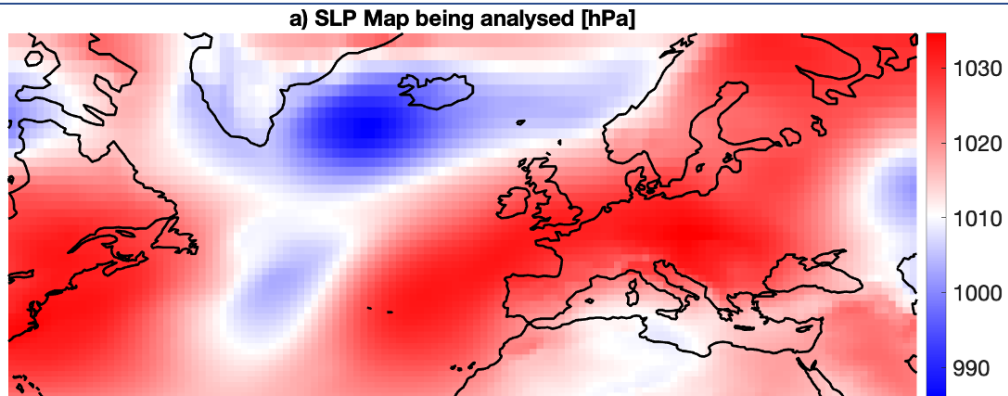
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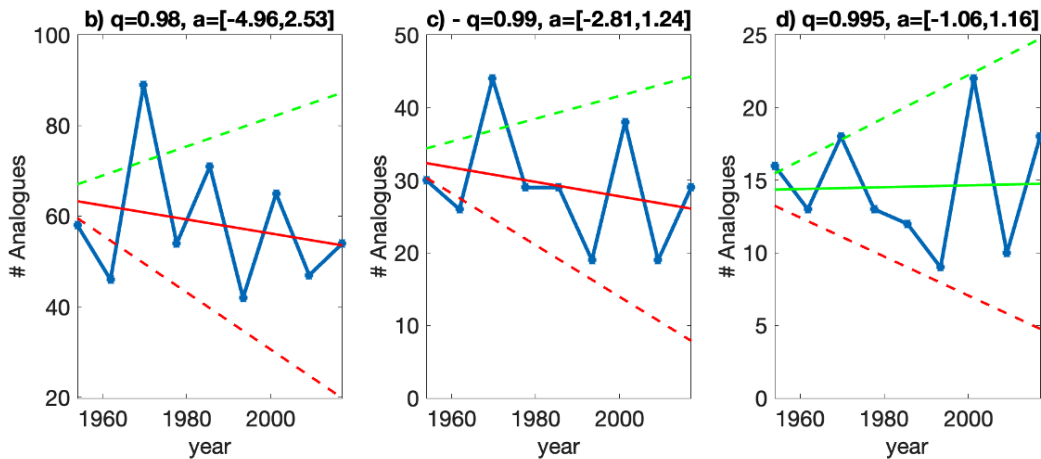
Dataset-Variable (N periods) [Type of fit]	Winter		Summer	
	Increasing	Decreasing	Increasing	Decreasing
ERA5-SLP (9) [linear]	3.7±1.0	-3.1±0.7	3.6±0.9	-2.9±0.5
ERA5-SLP (14) [linear]	2.3±0.5	-2.2±0.4	2.4±0.5	-2.0±0.3
ERA5-SLP (9) [cubic]	3.7±1.0	-3.1±0.7	3.6±0.9	-2.9±0.5
ERA5-Z500 (9) [linear]	4.1±1.1	-4.1±1.1	4.2±1.2	-3.2±0.8
NCEP-SLP (9) [linear]	3.9±1.0	-3.2±0.7	4.0±1.1	-4.3±1.2

Table S1. Magnitude of trends estimated for different datasets, variables and number of periods for which the trends are computed over the 73 years of data. 9 periods correspond to roughly 8 years bin lengths, 14 periods to roughly 5 years bin lengths. The type of fit indicates whether a linear or a cubic fit has been applied to the data. The magnitude of all trends is expressed in Number of analoges/decade, so that results are comparable. Trends are averaged over the three different quantiles used in the study: $q = 0.98, 0.99$ and 0.995 , and for all the days displaying increasing or decreasing occurrence trends in Winter or in Summer. The uncertainty is given in terms of one standard deviation of the mean.

1) Select a sea-level pressure map to analyse.



2) Count the number of analogues per period for different quantiles.

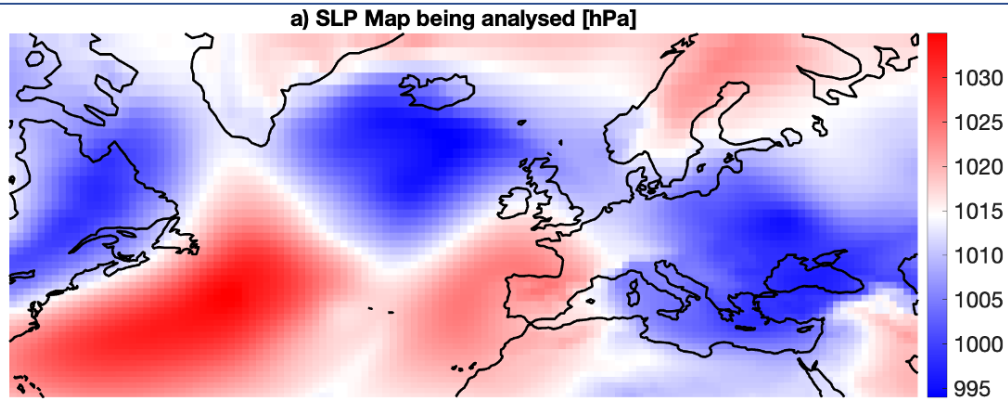


3) Fit a trend and compute confidence intervals.

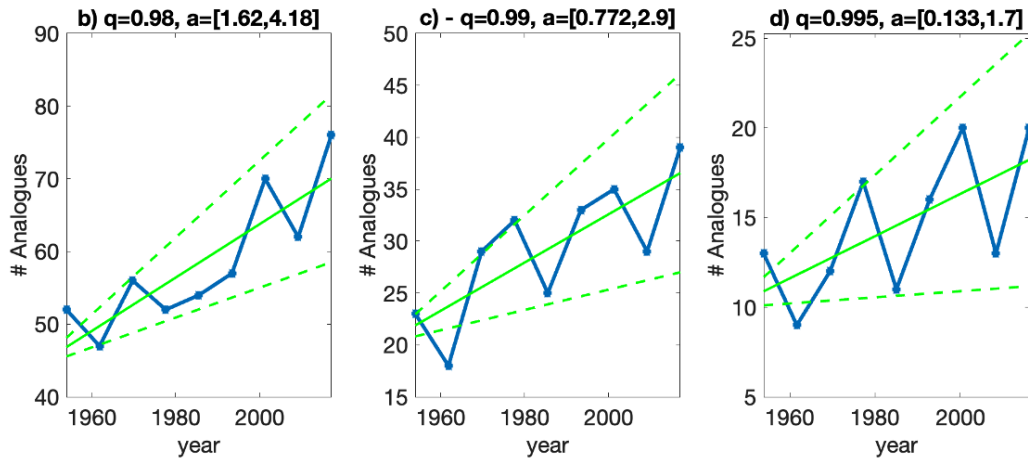
4) If all trend and confidence interval slopes have **positive** (**negative**) values, then this map is becoming increasingly **more** (**less**) frequent.

Fig. S1. Example of the procedure applied in the manuscript to detect daily sea-level pressure maps with increasing or decreasing occurrence trend: The sea-level pressure map for 01/01/1950 is shown in panel (a). The best 2% (b), 1% (c) and 0.5% (d) of analogues are searched in a database consisting of $n=26280$ daily sea-level pressure maps from ERA5 starting in January 1950. These correspond to the quantiles $q = 0.98, 0.99$ and 0.995 , respectively. The number of analogues (# Analogues) counted in each period of 7.5 years are then obtained (blue dots, panels b–d). A linear fit is performed and the confidence intervals for the slope a are computed (dotted lines and values in square brackets). Lines with positive (negative) slopes are coloured green (red). A daily sea-level pressure map displays a significant trend if all lines in panels (b–d) display the same color (red or green). This is not the case for the map 01/01/1950, taken in this example.

1) Select a sea-level pressure map to analyse.



2) Count the number of analogues per period for different quantiles.



3) Fit a trend and compute confidence intervals.

4) If all trend and confidence interval slopes have positive (negative) values, then this map is becoming increasingly more (less) frequent.

Fig. S2. Example of the procedure applied in the manuscript to detect daily sea-level pressure maps with increasing or decreasing occurrence trend: As Fig. S1, but for 04/01/1950 which does display a significant positive occurrence trend.

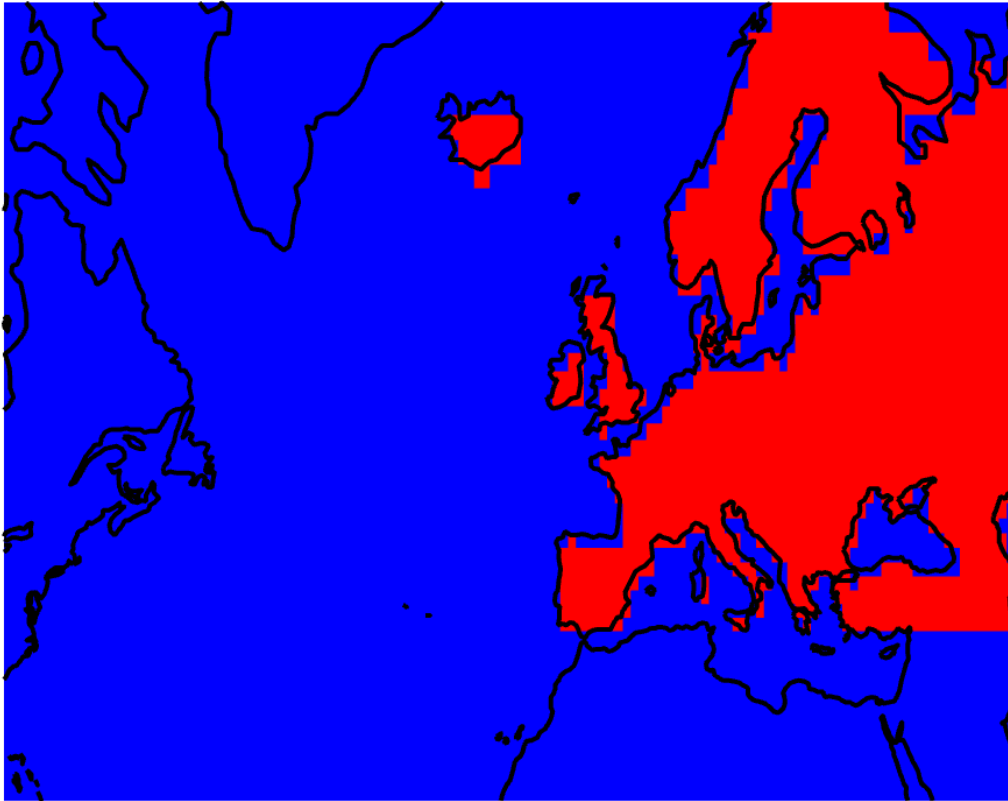


Fig. S3. European mask used to compute timeseries of continental European data: Red shading shows the gridboxes used to compute timeseries of continental European data shown in panels i, j of Figs. 1, 2 and S4–S19.

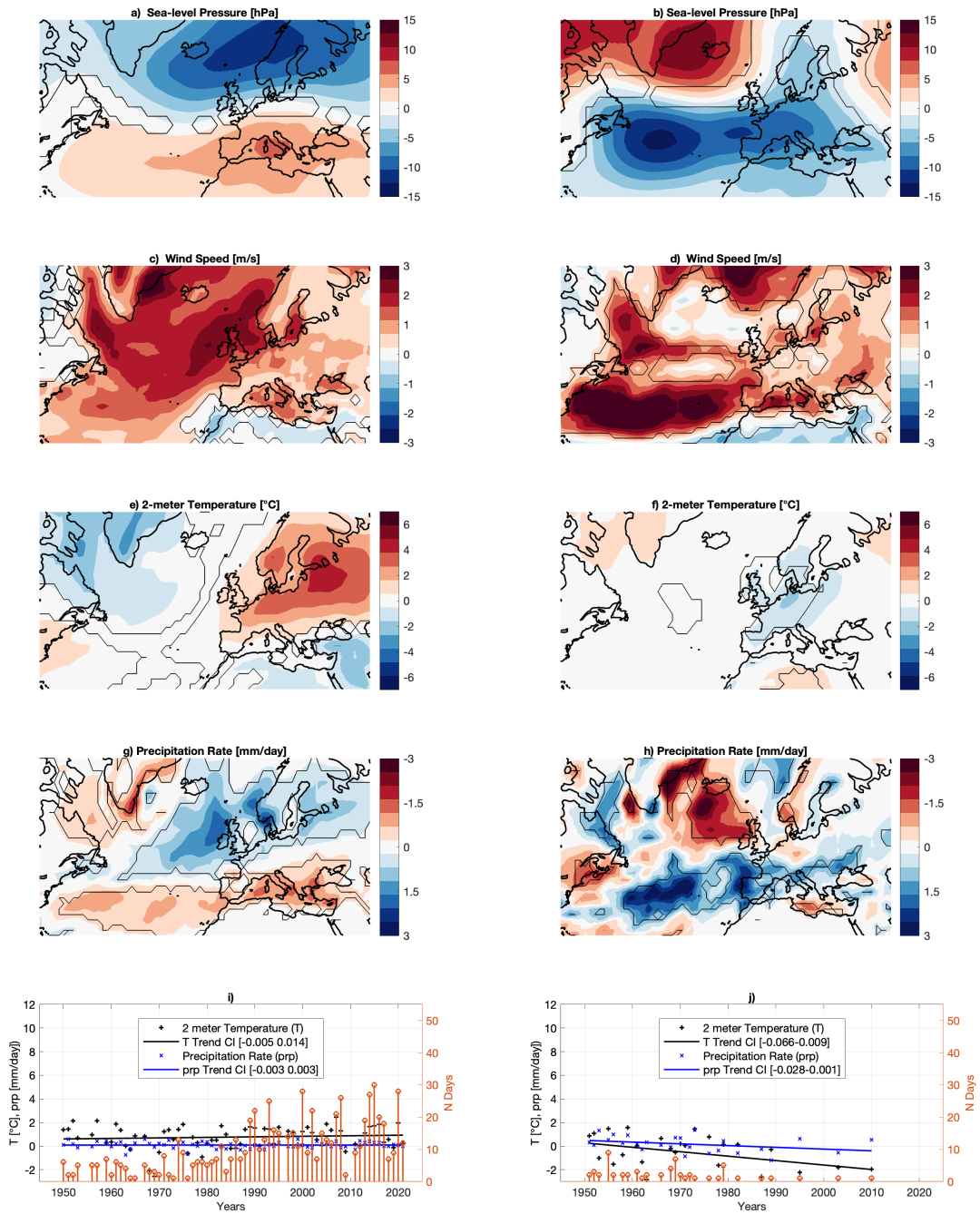


Fig. S4. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for NCEP-NCAR reanalysis data.

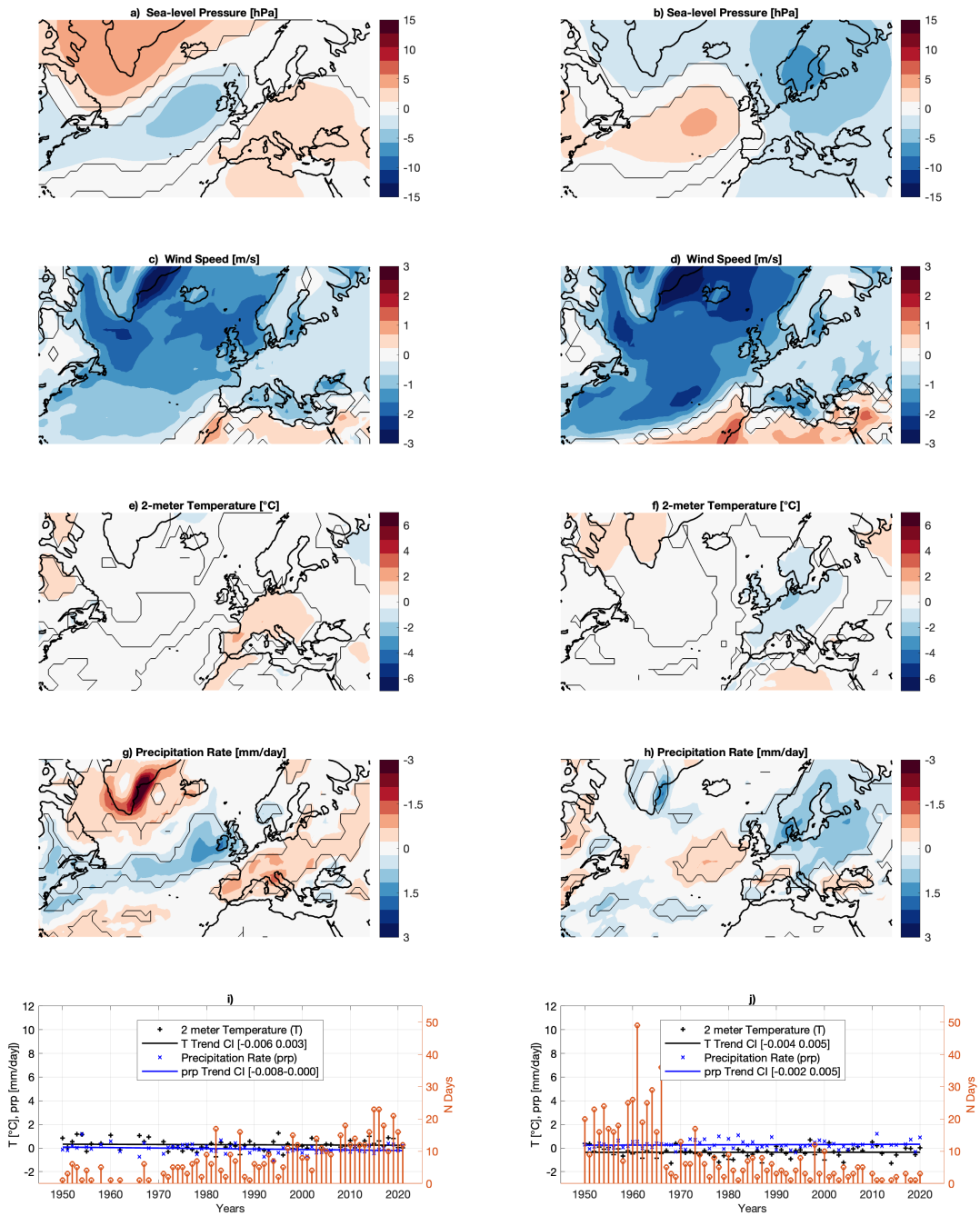


Fig. S5. Sea-level pressure summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but for NCEP-NCAR reanalysis data.

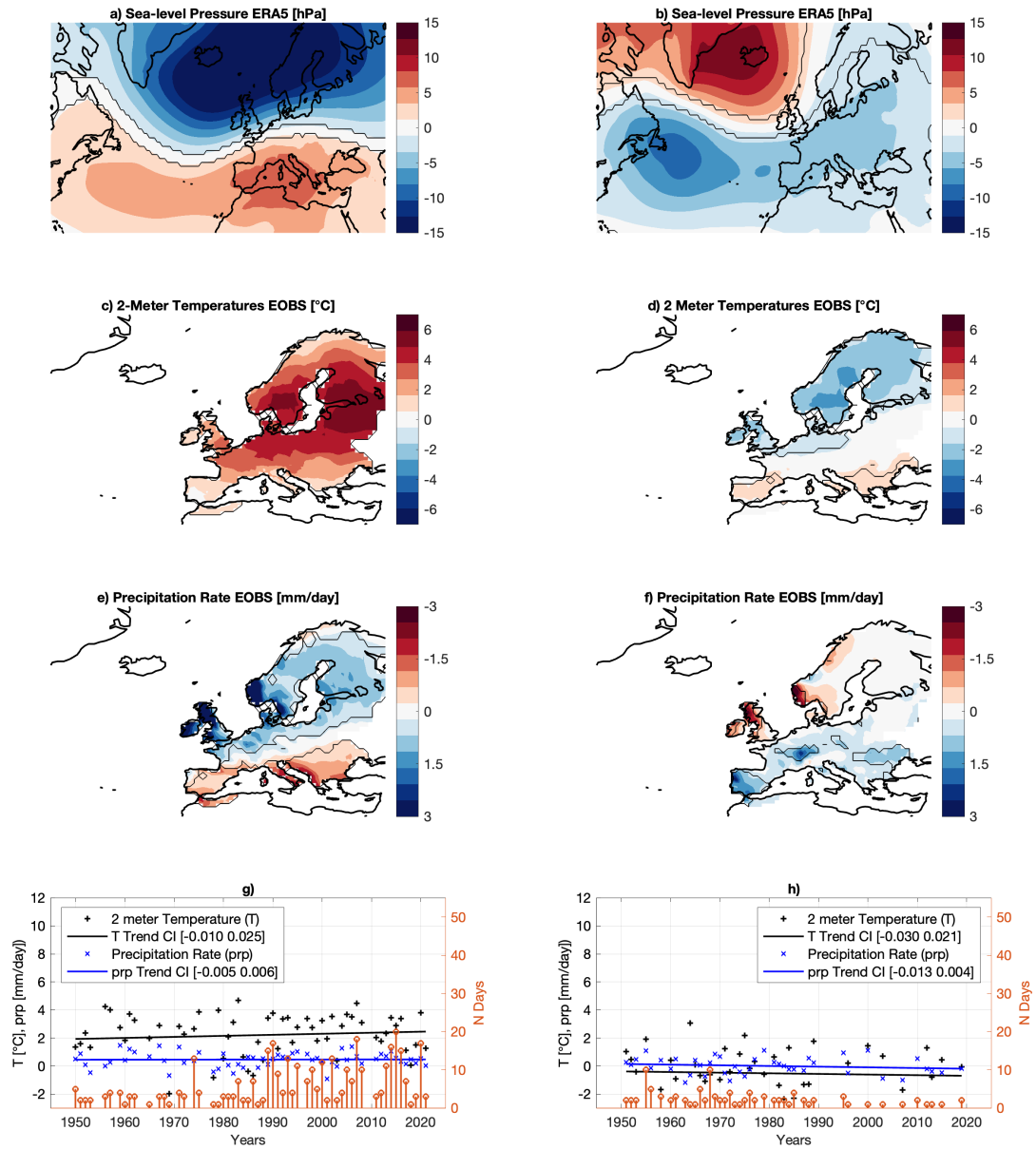


Fig. S6. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for ERA5 sea-level pressure data (a, b) and E-OBS temperature and precipitation (c–f). The figure does not include wind data.

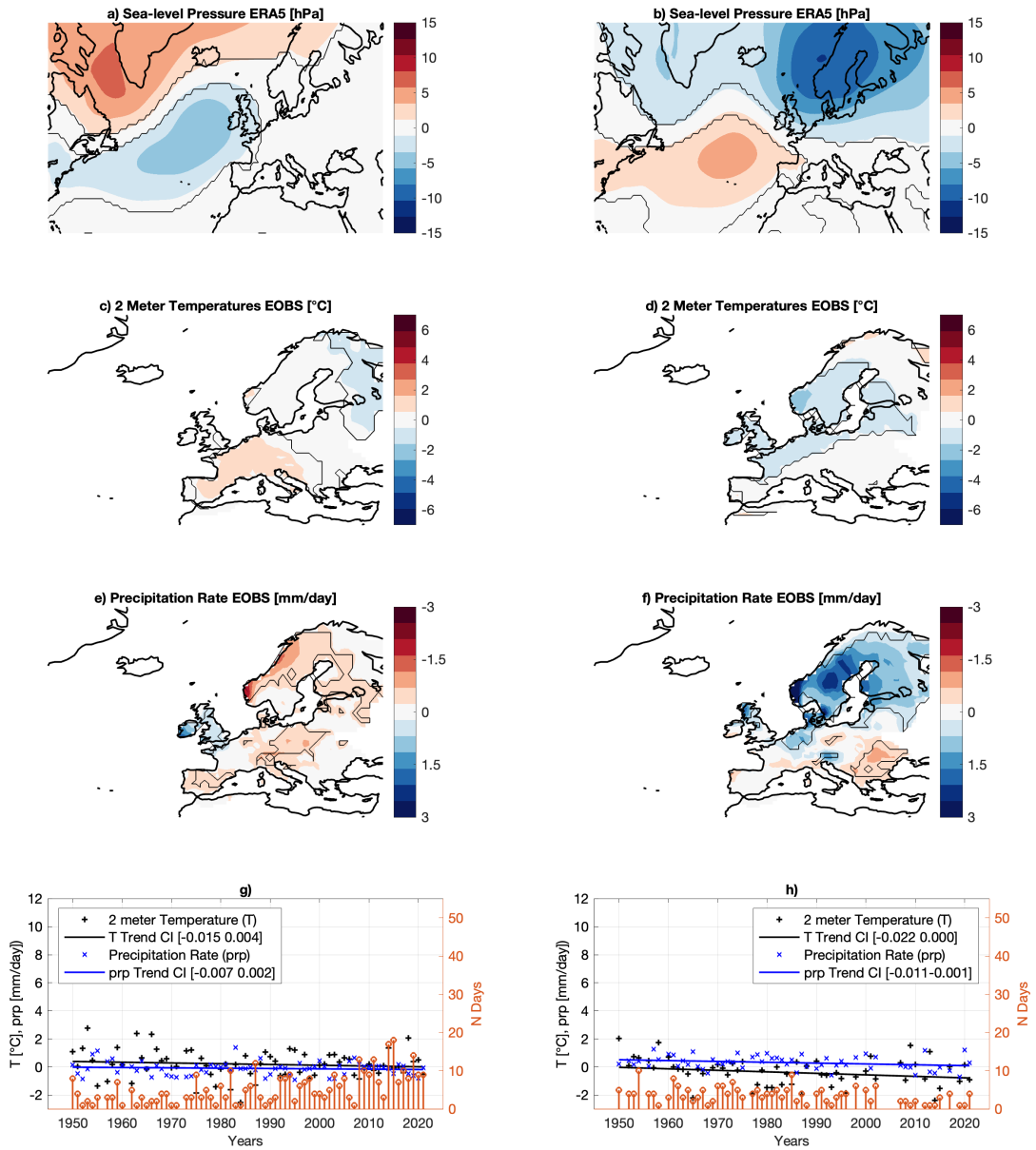


Fig. S7. Sea-level pressure summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but for ERA5 sea-level pressure data data (a, b) and E-OBS 2-metre temperature and precipitation (c-f). The figure does not include wind data.

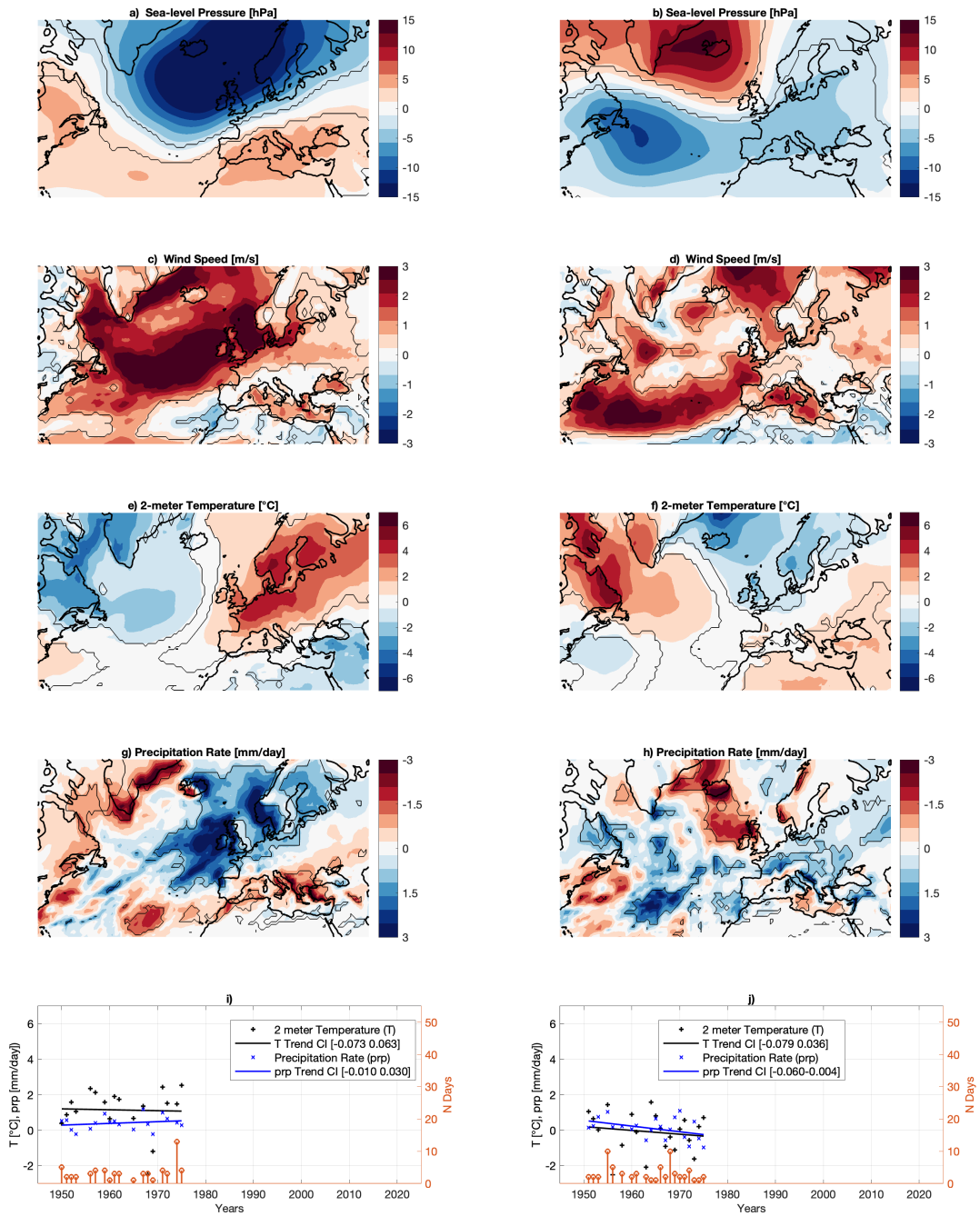


Fig. S8. First 25 years of the sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for the earliest 25 years in the dataset.

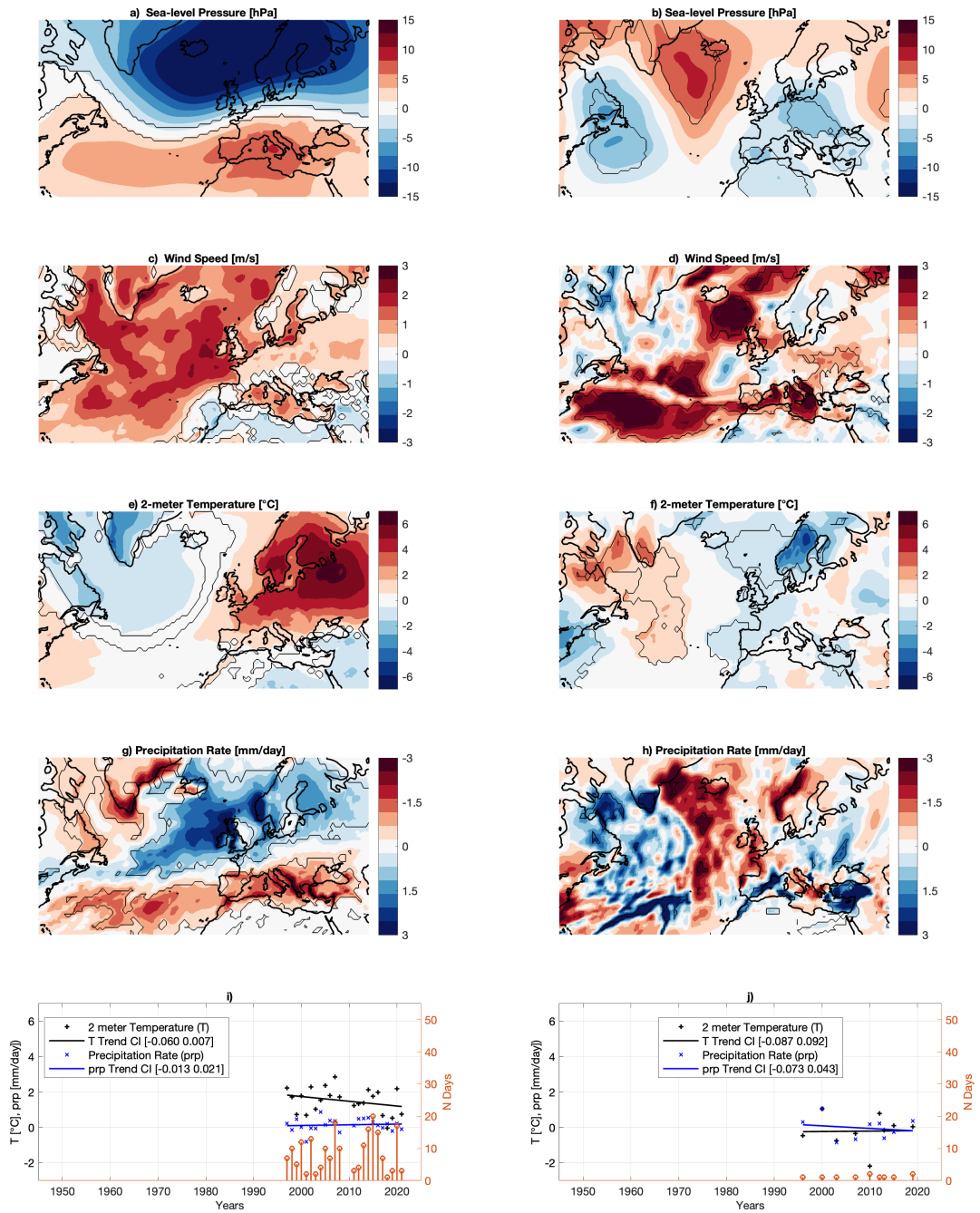


Fig. S9. Last 25 years of the sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for the latest 25 years in the dataset.

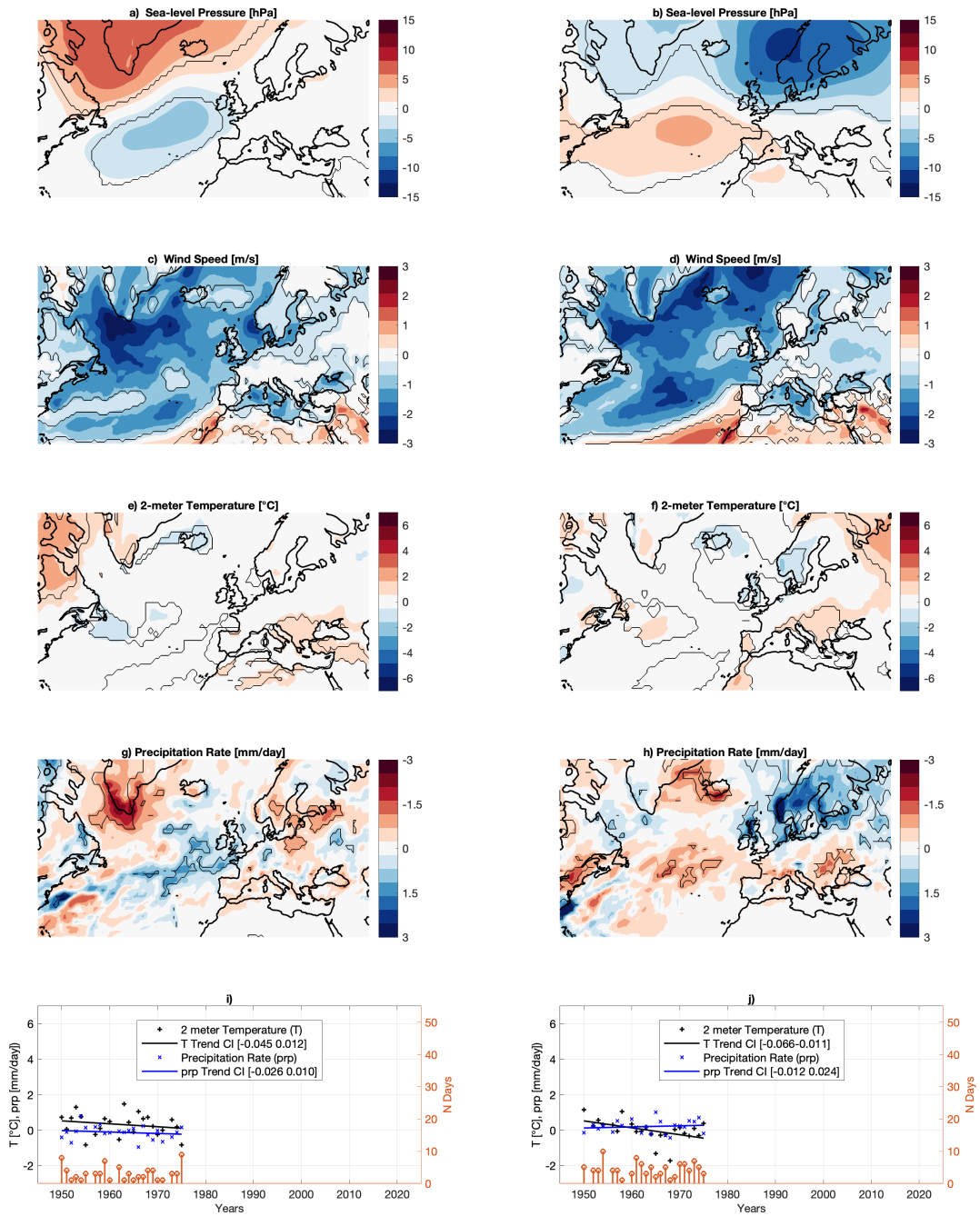


Fig. S10. First 25 years of the sea-level pressure summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but for the earliest 25 years in the dataset.

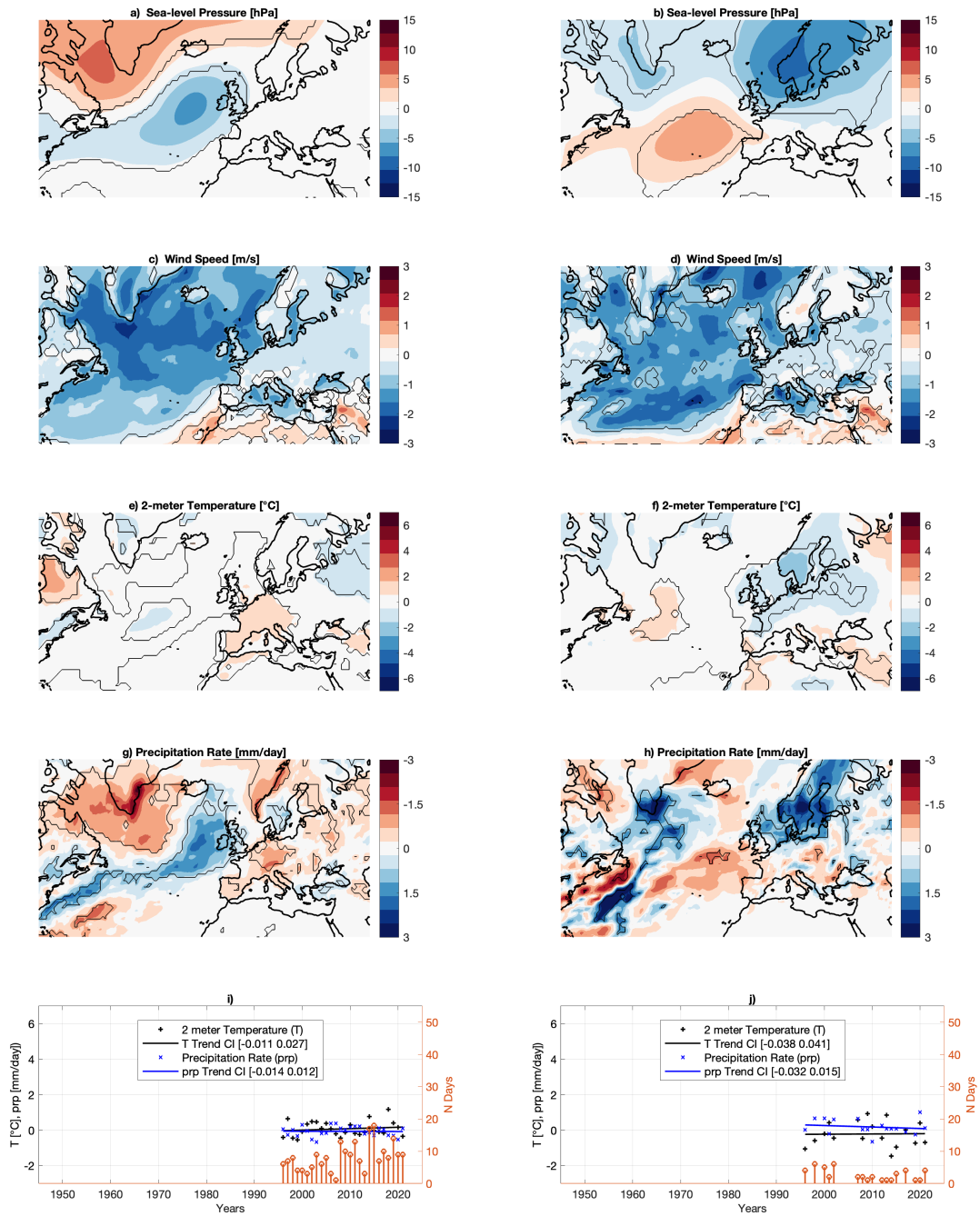


Fig. S11. Last 25 years of the sea-level pressure summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but for the latest 25 years in the dataset.

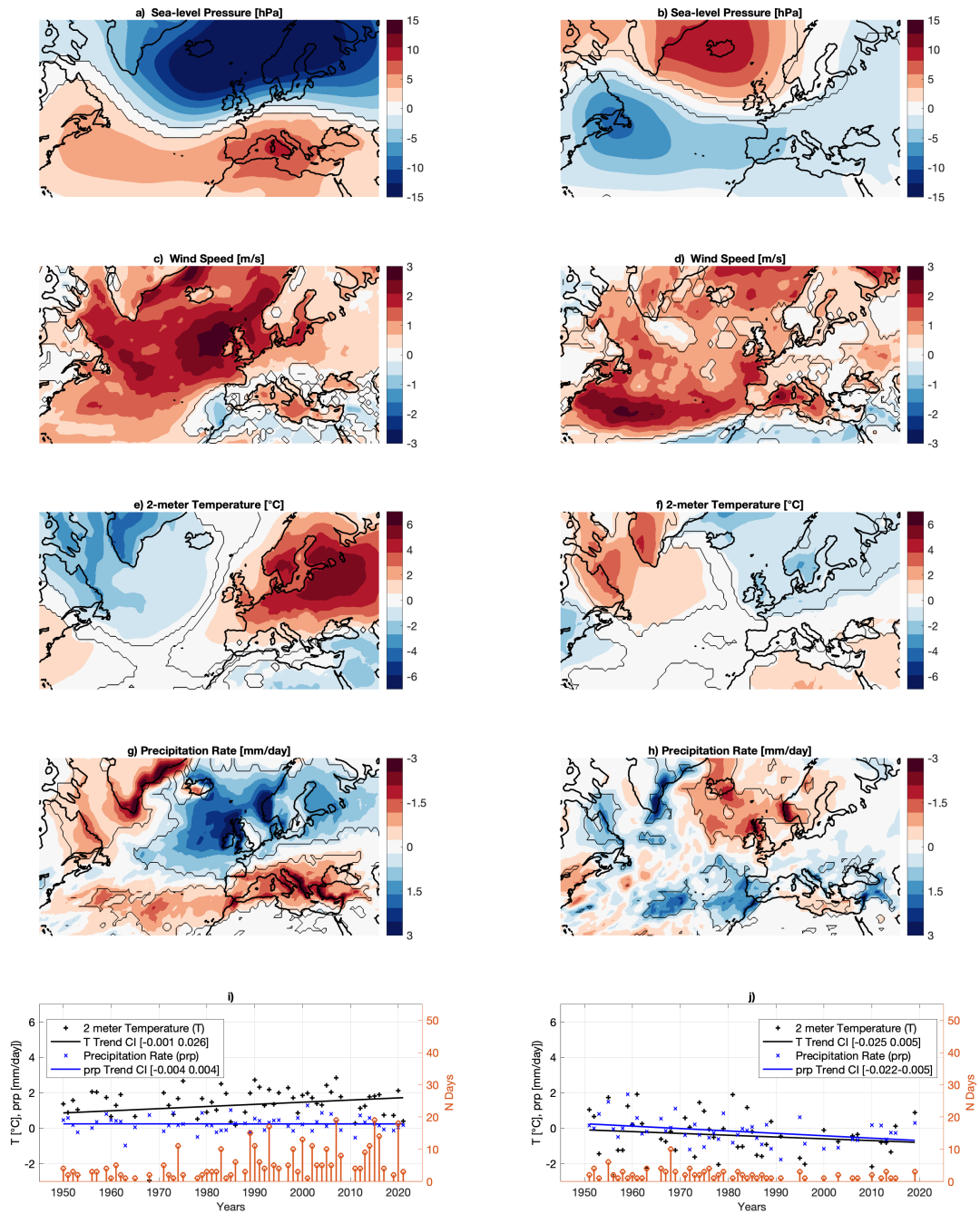


Fig. S12. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but dividing the time interval of 73 years in 14 periods of roughly 5 years.

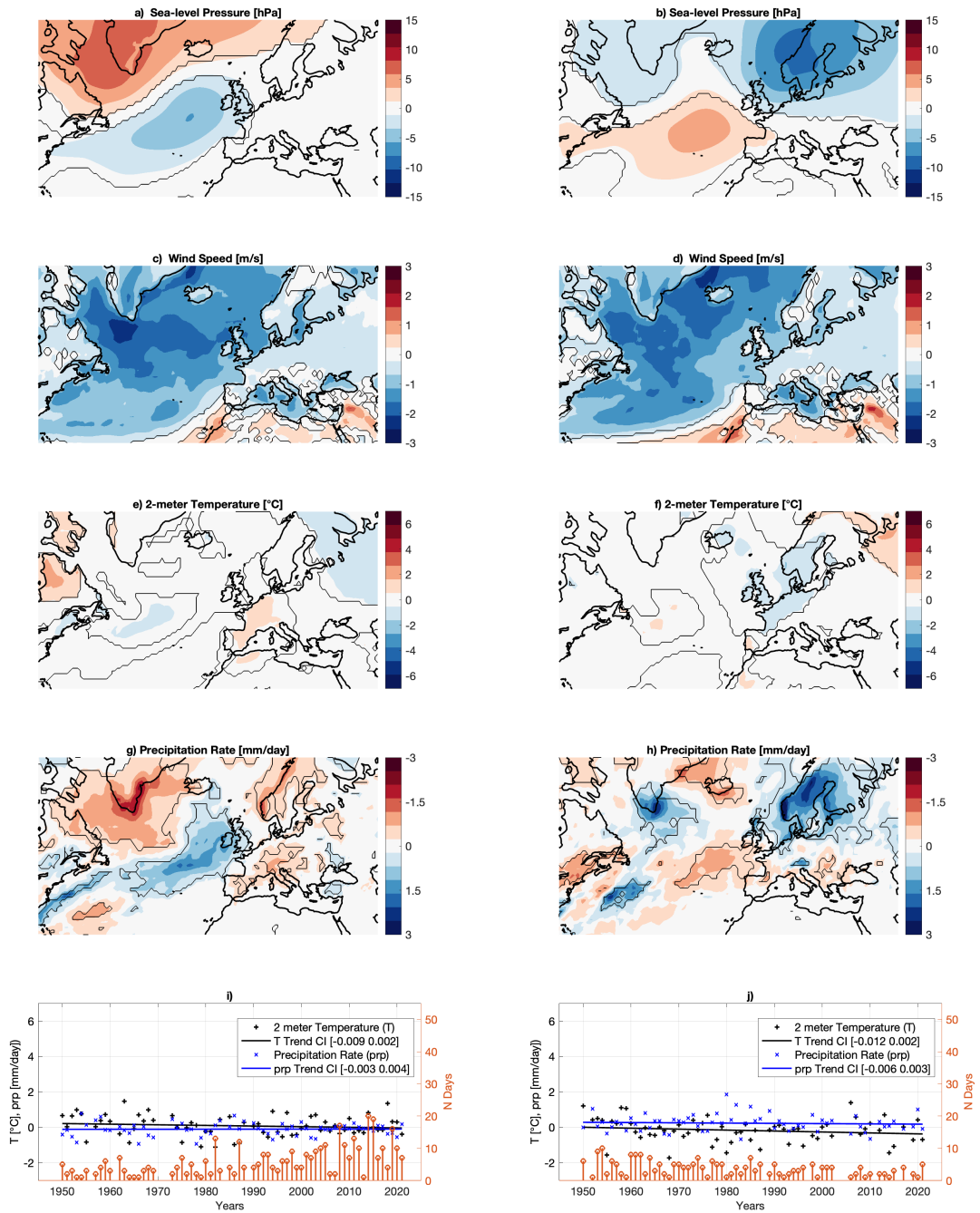


Fig. S13. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but dividing the time interval of 73 years in 14 periods of roughly 5 years.

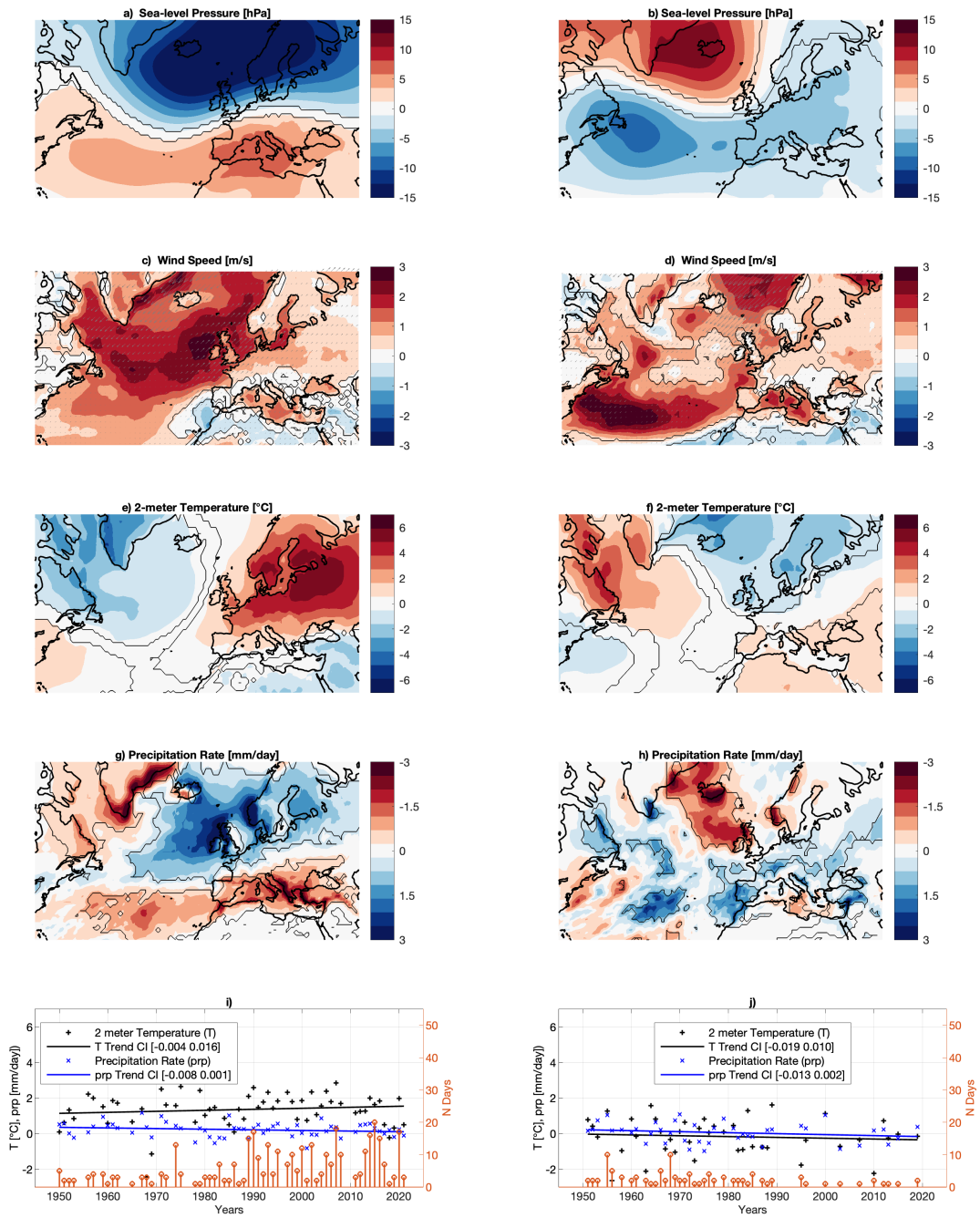


Fig. S14. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but when applying cubic fits (instead of linear) to the data.

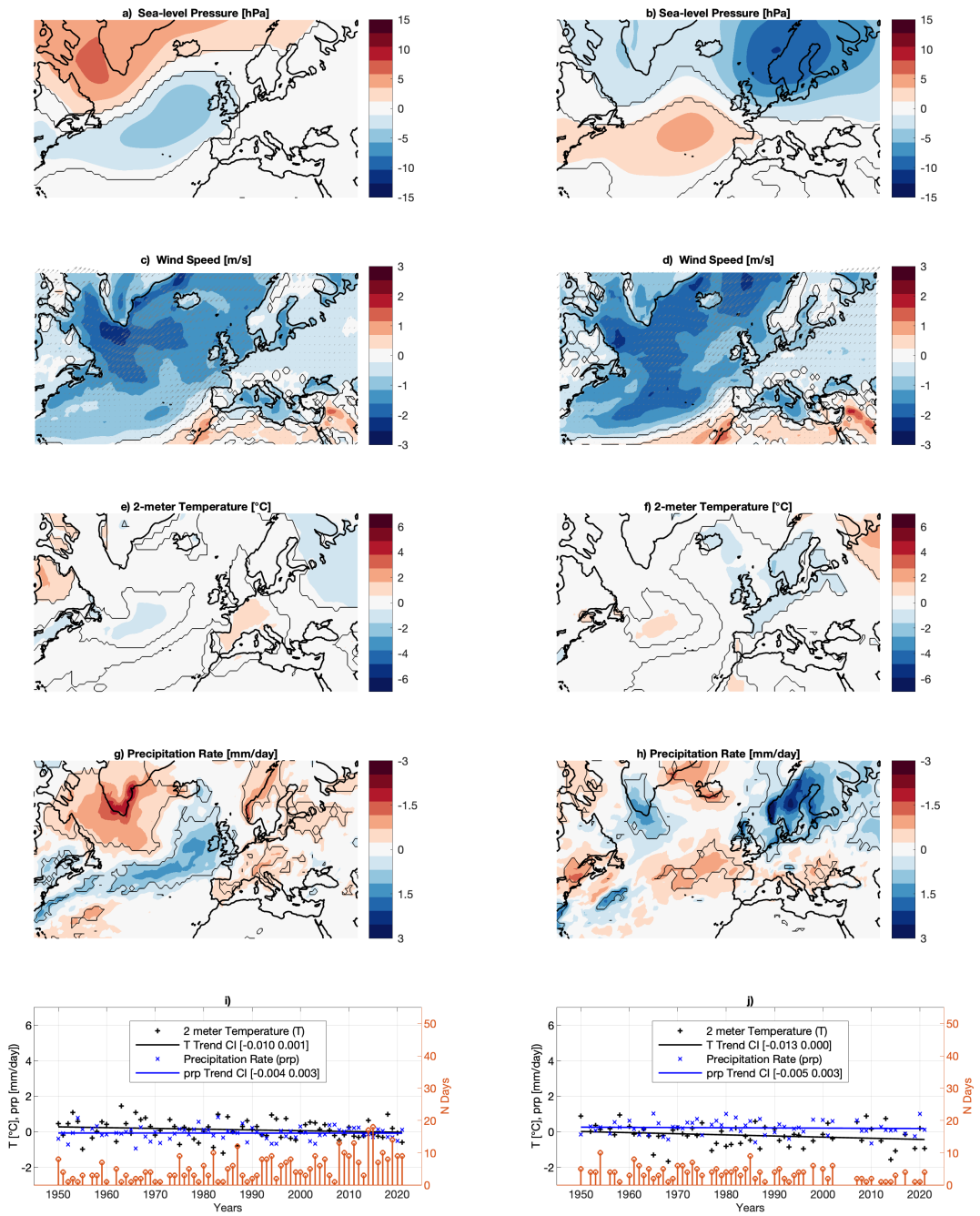


Fig. S15. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but when applying cubic fits (instead of linear) to the data.

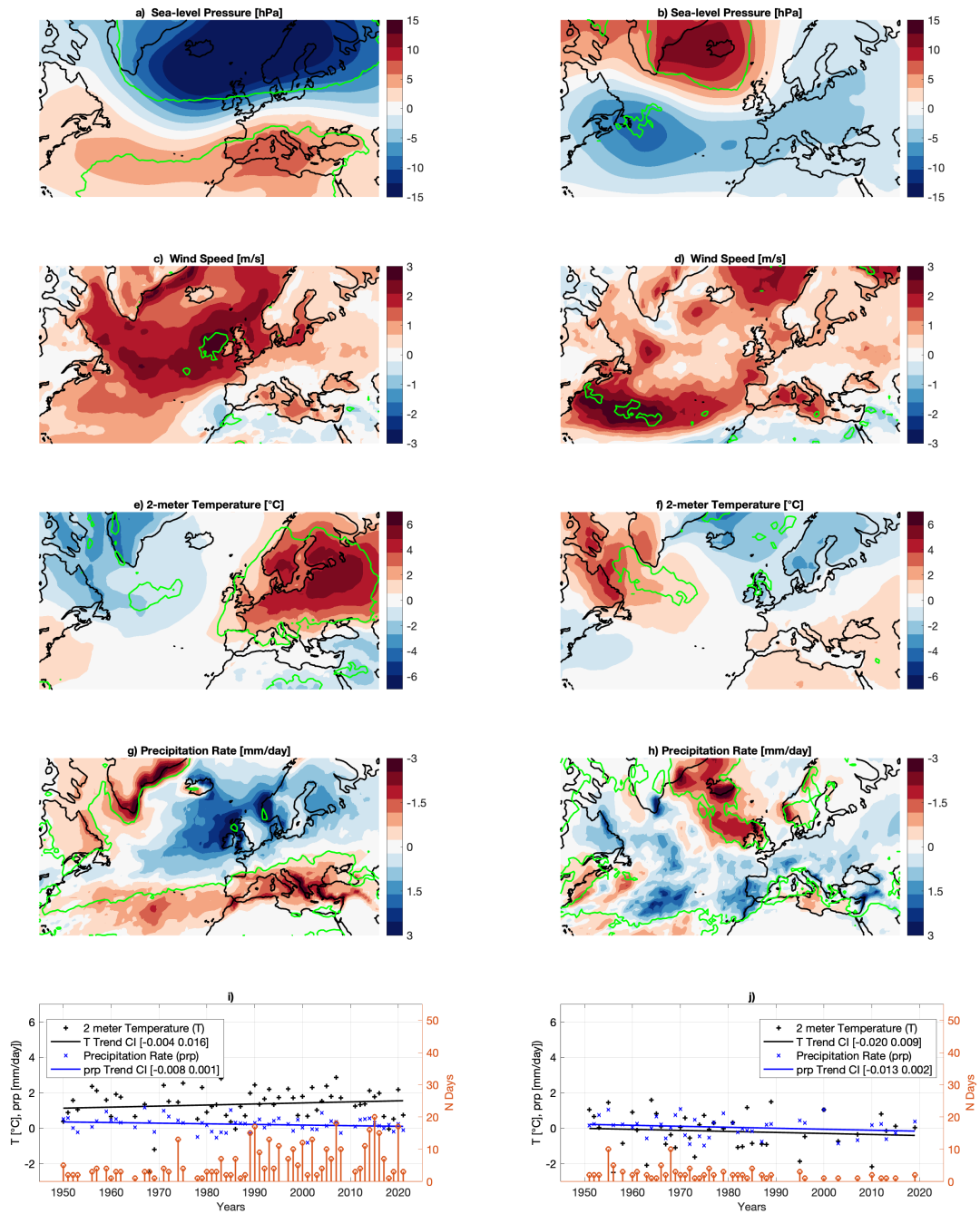


Fig. S16. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but where green contours indicate regions where 75% of the days display anomalies of the same sign.

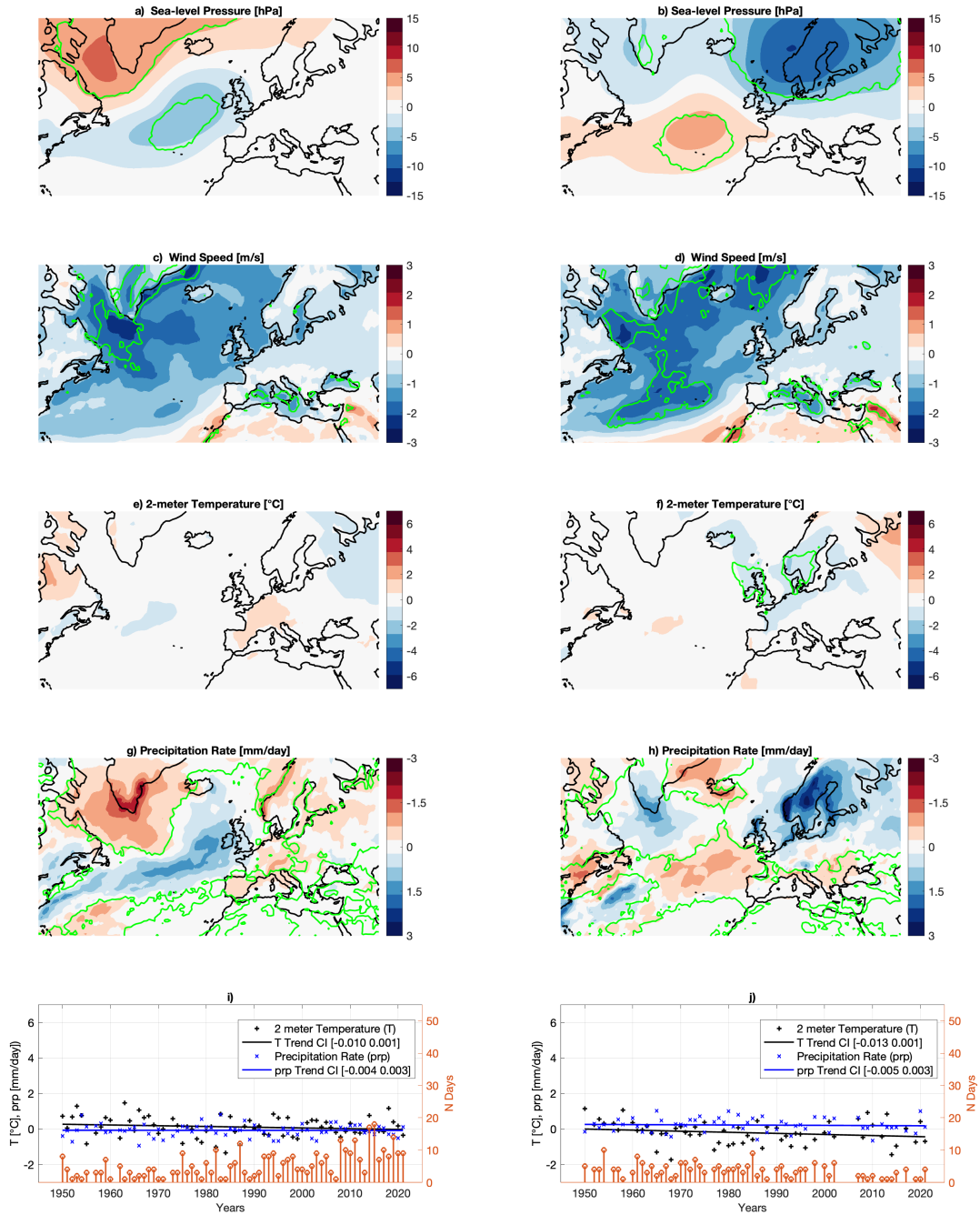


Fig. S17. Sea-level pressure wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but where green contours indicate regions where 75% of the days display anomalies of the same sign.

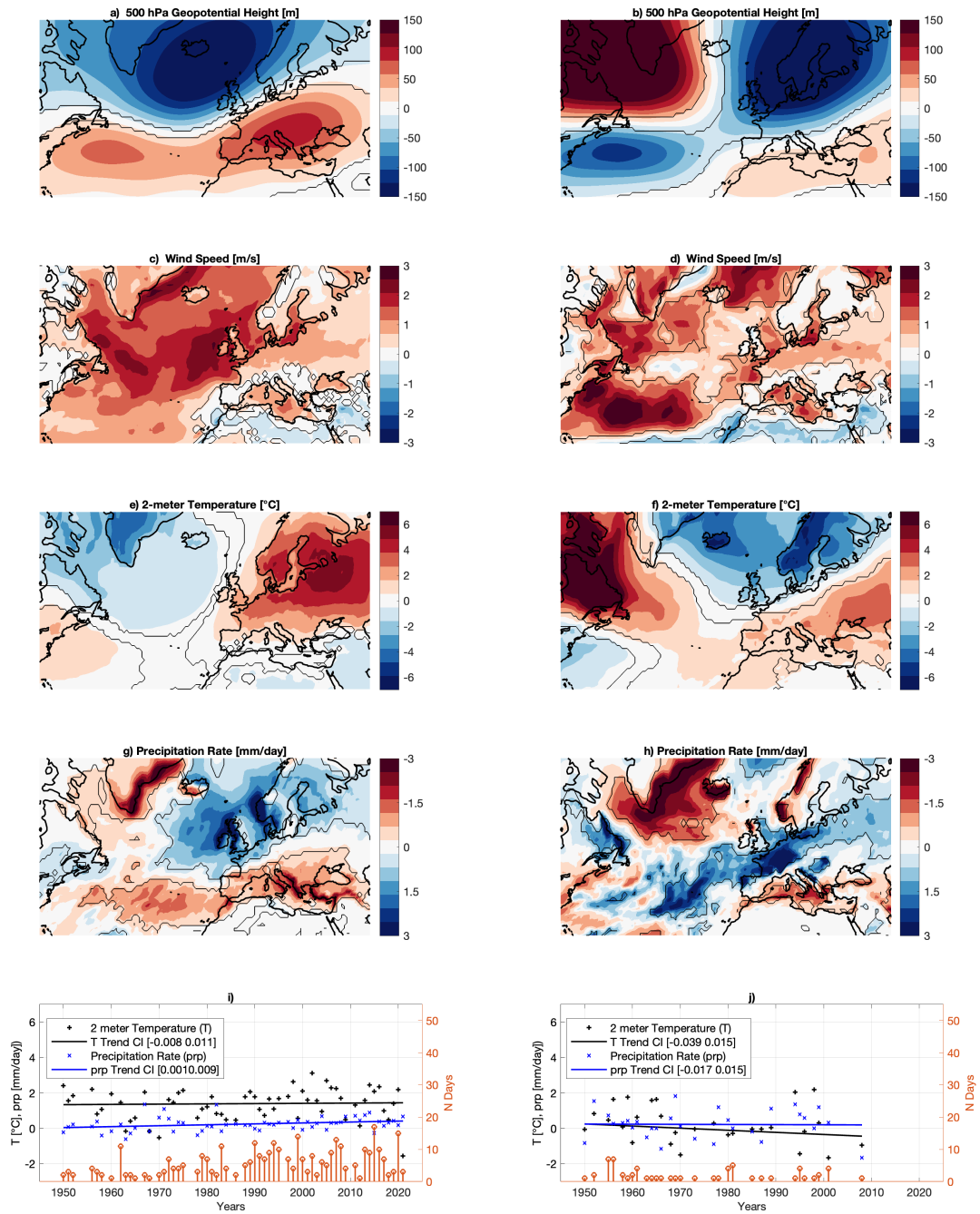


Fig. S18. 500 hPa geopotential height wintertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 1, but for 500 hPa geopotential height data from ERA5 (a,b).

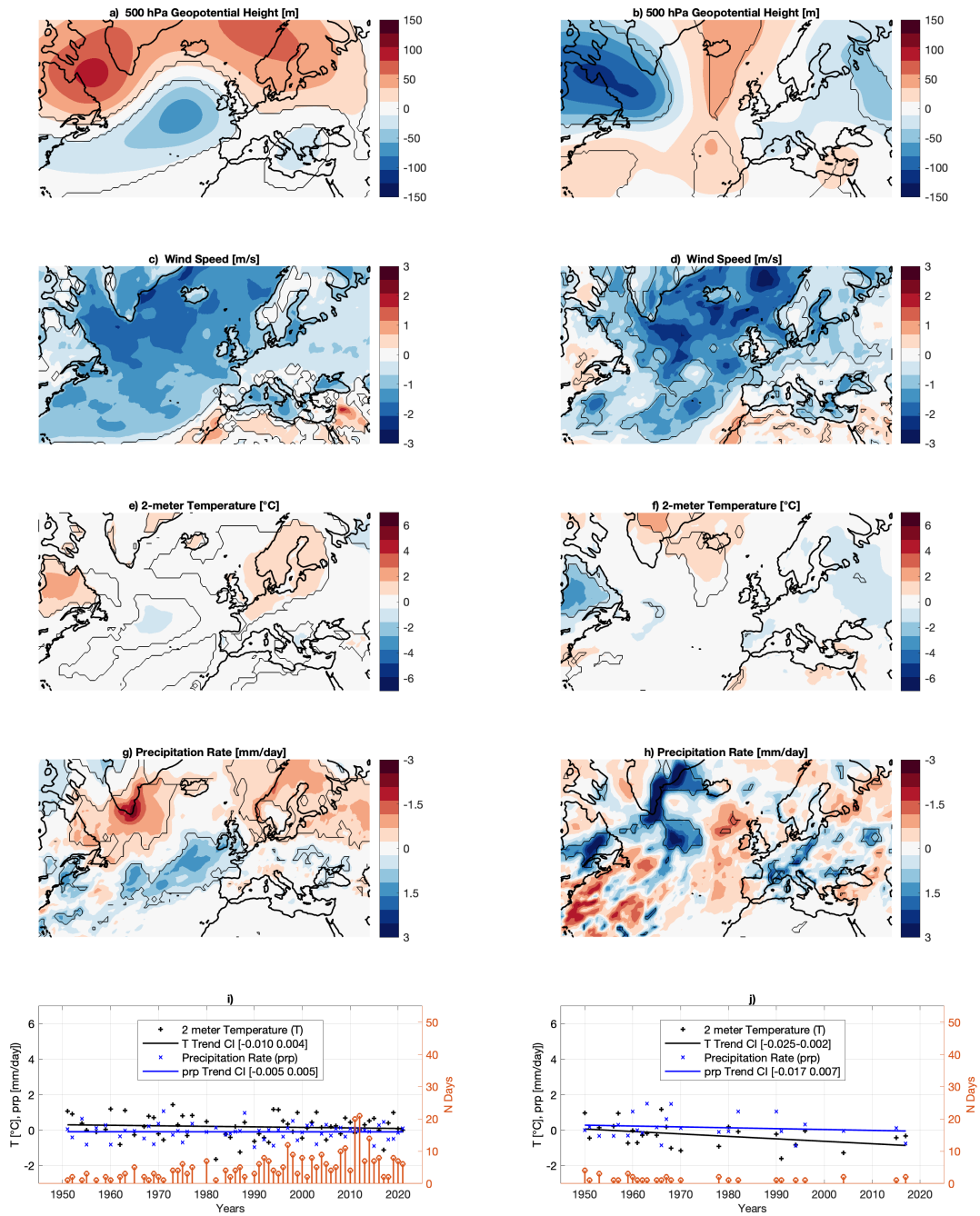


Fig. S19. 500 hPa geopotential height summertime atmospheric circulation patterns with significant occurrence trends and associated surface anomalies: As in Fig. 2, but for 500 hPa geopotential height data from ERA5 (a,b).

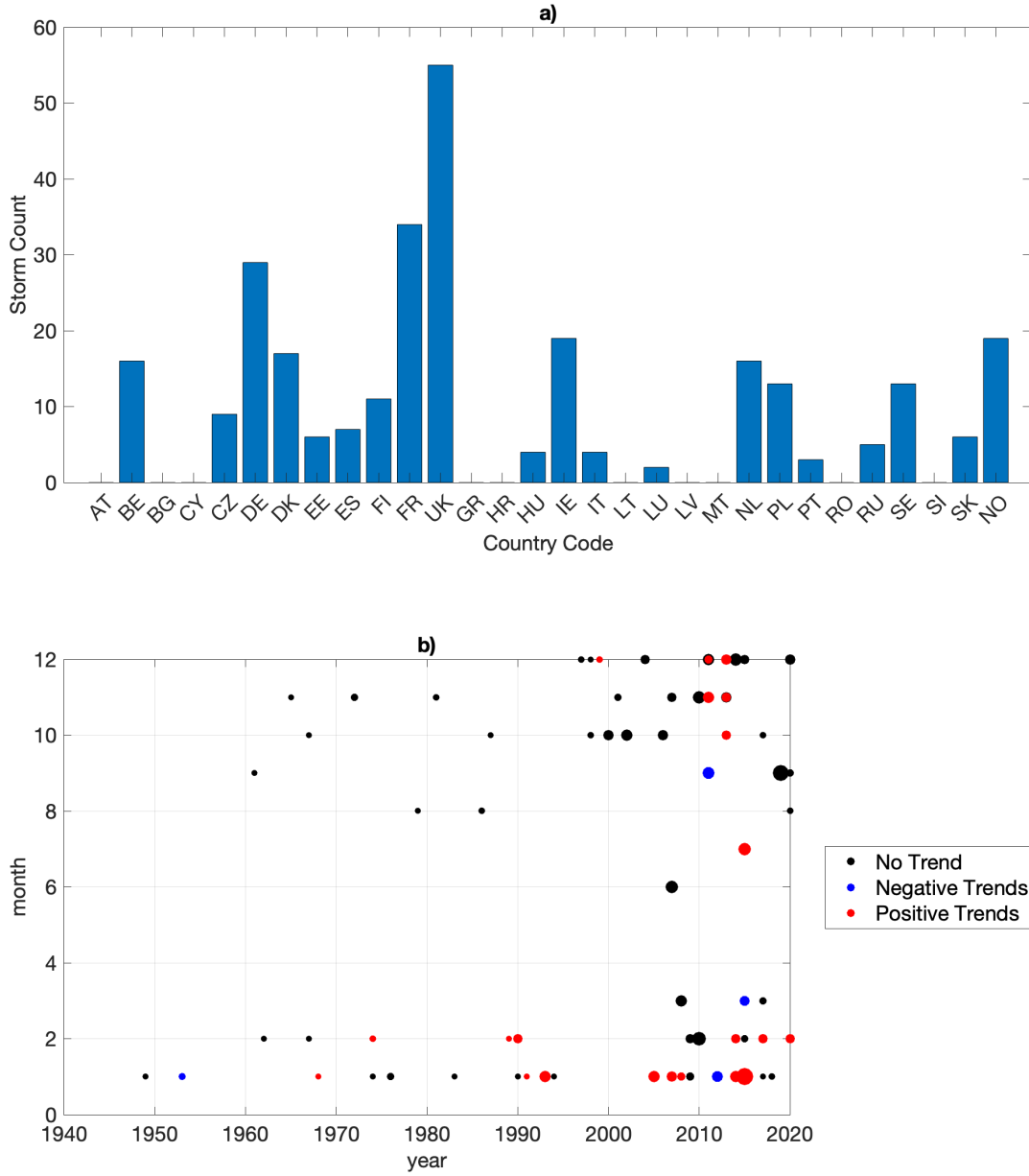


Fig. S20. Spatial and temporal distribution of European windstorms: Number of European Windstorms per country (a). Year and month of occurrence of each storm and duration (size of the balls). Colours show storms associated with atmospheric patterns displaying no (black), negative (blue) or positive (red) occurrence trends (b).

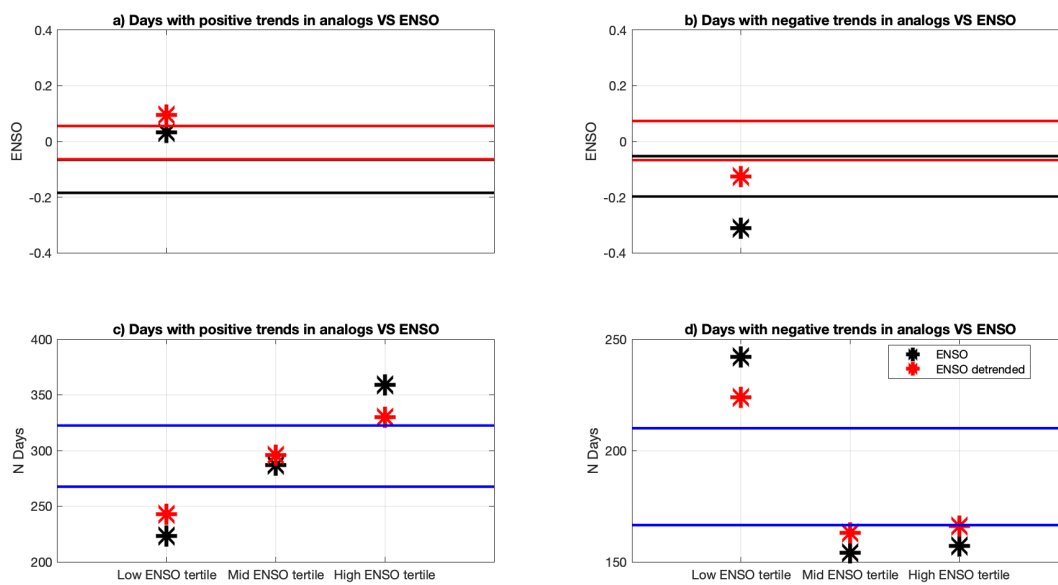


Fig. S21. Monthly El Niño – Southern Oscillation values conditioned on days with significant occurrence trends and count of days with significant occurrence trends corresponding to different El Niño – Southern Oscillation phases: Averages of monthly ENSO 3.4 values conditioned on positive (a) or negative (b) occurrence trends (stars). Number of days with positive (c) or negative (d) occurrence trends during months with ENSO 3.4 values in the upper (right), middle (center) and lower (left) tertiles of the ENSO 3.4 index distribution. Results for both the raw (black) and the linearly detrended (red) ENSO timeseries are shown. The continuous lines correspond to the 5% significance bounds obtained by a bootstrapping procedure with 500 iterations.

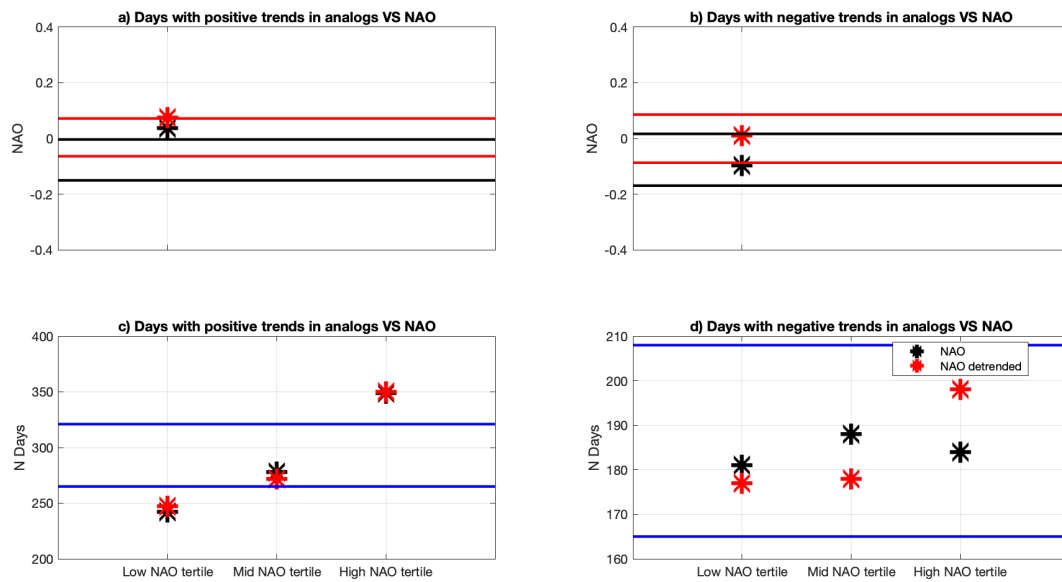


Fig. S22. Monthly North Atlantic Oscillation values conditioned on days with significant occurrence trends and count of days with significant occurrence trends corresponding to different North Atlantic Oscillation phases: As in Figure S21 but for the North Atlantic Oscillation Index.

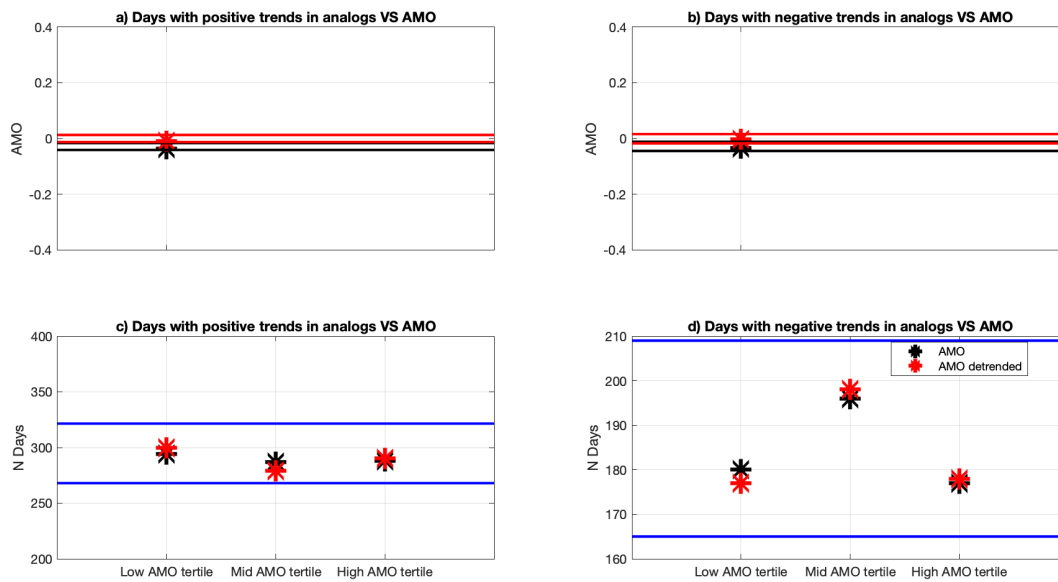


Fig. S23. Monthly Atlantic Multidecadal Oscillation values conditioned on days with significant occurrence trends and count of days with significant occurrence trends corresponding to different Atlantic Multidecadal Oscillation phases: As in Figure S21 but for the Atlantic Multidecadal Oscillation index.