

# Linkages between the atmospheric transmission originating from the North Atlantic Oscillation and persistent winter haze over Beijing

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**Abstract.** In this study, the persistent winter haze that occurred over Beijing during 1980 to 2016 is examined using reanalysis and station data. On both interannual and daily-to-weekly timescales, the winter haze weather in Beijing is found to be associated with a pronounced atmospheric teleconnection pattern from the North Atlantic to Eurasia (Beijing). A positive western-type North Atlantic Oscillation (WNAO+) phase and a positive East Atlantic/West Russia (EA/WR+) phase are observed as part of this teleconnection pattern (an arched wave train). This study focuses on the role of the WNAO pattern, because the WNAO+ pattern acts as the origin of the atmospheric transmission, 8–10 days before the persistent haze events. Further analyses reveal that the WNAO+ pattern can increase the number of haze days and persistent haze events on interannual and daily-to-weekly timescales. Specifically, strong WNAO+ winters (above the 95th percentile) can increase the number of haze days and persistent haze events by 26.0% and 42.3%, respectively. In addition, a high WNAO index for the five-day average (above the 95th percentile) predicts a 16.9% increase in the probability of haze days on Day 8 and a higher proportion of persistent haze days compared with an unknown WNAO state. Thus,

the WNAO+ pattern is as a necessary prior background condition for the formation of the wave train and is a skillful predictor for persistent hazy weather. Corresponding to the WNAO+ pattern, intensified zonal wind and a north–south sea surface temperature tripolar mode over the North Atlantic also appear before persistent haze events on the daily-to-weekly timescale. On the interannual timescale, winters with a greater number of persistent haze days are also associated with a tripolar SST mode over the North Atlantic that is situated farther northward.

## 1 Introduction

Beijing is the capital city of China and is situated in the northeast of the country. It covers 16,410 km<sup>2</sup> and has a permanent population of 21.542 million. In 2019, Beijing's gross domestic product reached 3,537 billion yuan, which was an increase of 6.1% over the previous year (<http://www.stats.gov.cn>). Along with economic development, Beijing has experienced more frequent hazy weather, especially persistent haze in winter, over the past 60 years (Wang et al., 2014; Su et al., 2015; Li et al., 2018a; Pei and Yan, 2018; Pei et al., 2018; Shi et al., 2019). Haze pollution is associated with a high concentration of particulate matter with a diameter of less than 2.5 micrometers (PM<sub>2.5</sub>) and low visibility, which is harmful to human health (e.g., cardiovascular and respiratory diseases and lung cancer) and puts pressure on, among other things, public transportation and economic activities (Wang et al., 2013).

The serious impacts of haze pollution in Beijing have been the topic of numerous studies. High emissions of haze pollutants (e.g., black carbon and organic matter) contribute greatly to the formation of hazy weather in Beijing (Li and Han, 2016; Wu et al., 2016; Li et al., 2017). In addition, the atmospheric and meteorological conditions as well as external and remote influences, such as Arctic sea-ice concentration (SIC), snow cover across Siberia and sea surface temperatures (SSTs), also need

49 to be taken into consideration (An et al., 2019; Wang et al., 2020; Yin and Wang 2016, 2018).  
50 Atmospheric circulations that are favorable for hazy weather in Beijing include a weak East Asian  
51 winter monsoon (EAWM), a shallow East Asian trough and a northward shift of the East Asian jet  
52 (Chen and Wang, 2015; Zou et al., 2017; Pei et al., 2018; Wang et al., 2020). These circulations tend  
53 to reduce the intrusion of cold air to Beijing, and hence result in poor ventilation in winter (Zou et al.,  
54 2017; Pei et al., 2018). Furthermore, teleconnection patterns and wave trains also have potential  
55 impacts on haze over Beijing (Yin and Wang, 2017; Yin et al., 2017; Chen et al., 2019; Zhang et al.,  
56 2019; Chen et al., 2020a, b; Lu et al., 2020). Yin et al. (2017) analyzed the role of the positive phases  
57 of the East Atlantic/West Russia (EA/WR+) pattern, the western Pacific pattern and the Eurasia pattern  
58 in the increased number of hazy days over the North China Plain. They found that these climatic  
59 anomalies can lead to meteorological conditions that are conducive to the formation of haze pollution  
60 through modulating the anticyclonic anomalies over North China (Yin and Wang, 2017; Yin et al.,  
61 2017). The positive phase of the Arctic Oscillation (AO+) pattern can also increase the number of hazy  
62 days in Beijing (Lu et al. 2020). Years with a high AO index are accompanied by a weakened East  
63 Asian trough and a weakened Siberian high, which suppress the [horizontal and vertical](#) diffusion of  
64 haze pollutants (Lu et al. 2020). Chen et al. (2019) stressed the role of the positive phase of the North  
65 Atlantic Oscillation (NAO+), which is related to AO+, in inducing an anticyclone over northeast China,  
66 which favors haze pollution in North China in spring. Chen et al. (2020b) also studied autumn haze,  
67 and found that the relative importance of the external drivers seems to differ across the individual  
68 months of September, October and November. In addition, the winter haze weather in Beijing is  
69 directly affected by the local meteorological conditions. Many studies have suggested that static and  
70 relatively warm air, weakened northerly or even southerly winds, decreased relative humidity,

71 temperature inversion and downward air motion in the planetary boundary layer can suppress the  
72 dispersal and advection of haze pollutants (Wang et al., 2014; Zhang et al., 2014; Zhang et al., 2016;  
73 Wu et al., 2017; Wang et al. 2019; Zhong et al., 2019; Callahan and Mankin, 2020).

74 In addition, a reduction in Arctic Sea ice in autumn (as documented by Simmonds and Li 2021)  
75 can increase the number of subsequent winter haze days through weakening wave activity over eastern  
76 China (Wang and Zhang, 2015; Zou et al., 2017). Zou et al. (2017) also revealed that increased  
77 Eurasian snowfall in earlier winter leads to regional circulations unfavorable to the ventilation of  
78 pollutants. In addition, autumn Beaufort Sea ice is closely connected with the number of early-winter  
79 haze days in North China (Yin et al., 2019a; Li and Yin, 2020), while an increase in early-winter  
80 Chukchi Sea ice can intensify February haze pollution in North China (Yin et al., 2019b). The changes  
81 in sea ice in both the Beaufort Sea and the Chukchi Sea are linked to hazy weather in North China via  
82 modulated large-scale atmospheric circulations (e.g., the East Asian trough and teleconnections). SSTs  
83 in both the Atlantic and Pacific oceans have potential impacts on the occurrence of hazy weather (Xiao  
84 et al., 2015; Pei et al., 2018; Wang et al., 2019). Winter haze days in China are also associated with  
85 SST anomalies over the North Atlantic on decadal and interannual timescales via the Atlantic  
86 multidecadal oscillation, and are also related to SSTs over the South Atlantic on the interannual  
87 timescale by anomalous southerly airflow (Xiao et al., 2015). Pei et al. (2018) found that positive SST  
88 anomalies over the northwest Pacific are conducive to more winter haze days in Beijing through  
89 weakening of the EAWM system. Wang et al. (2019) also suggested that the interannual variability in  
90 autumn haze days in the Beijing–Tianjin–Hebei region is associated with a wave train induced by SSTs  
91 in the subtropical North Atlantic and a local meridional cell induced by SSTs in the western North  
92 Pacific.

Persistent haze events correspond to continuous pollution for several days, which not only has a broad impact on human life (through traffic jams, for example) but also threatens human health in many ways. In this study, we gain a better understanding of the physical processes and mechanisms of the persistent haze over Beijing. Local meteorological conditions are directly associated with hazy weather and usually show diurnal variations (Zhang et al., 2014; Li et al., 2018b; Li et al., 2019), while external influences, which vary slowly, play key roles in explaining the interannual and interdecadal variabilities in hazy weather (Wang et al., 2020). On intraseasonal timescales, large-scale atmospheric circulations (e.g., teleconnections and wave trains) can bridge the timescales of local meteorological conditions and external forces. Large-scale atmospheric circulations can be modulated by external forces and can influence local meteorological conditions. In previous studies, the atmospheric patterns associated with hazy weather were mainly obtained from linear correlation or composite analysis based on interannual or longer timescales (Yin and Wang, 2017; Yin et al., 2017; Chen et al., 2019; Lu et al., 2020). However, the evolution of the atmospheric circulations and their roles in the formation of hazy weather (especially persistent haze events) from daily to intraseasonal timescales are not clear. Thus, a more in-depth analysis of daily-to-weekly changes is needed to determine a more robust relationship between haze and atmospheric circulation and to test whether certain circulation patterns can help to improve the predictability of haze days and persistent haze events.

In this study, we focus on persistent winter haze over Beijing and the corresponding large-scale atmospheric circulations from the perspective of interannual and daily-to-weekly timescales. In particular, we examine the role of a western-type NAO+ pattern in the increase of haze on the interannual timescale and the improvement of forecast skill for winter haze on the daily-to-weekly timescale. Furthermore, we investigate the SST and Arctic Sea ice conditions that have been proposed

as drivers of large-scale atmospheric circulations.

## 2 Data and methods

### 2.1 Data

The observed relative humidity and visibility at 20 stations across Beijing at four local times (02:00, 08:00, 14:00 and 20:00) each day from 1980 to 2016 during the winter (December, January and February (DJF)) were obtained from quality-controlled station observations at the National Meteorological Information Center of China. The daily means of these variables were then calculated. Reanalysis data were taken from the European Centre for Medium-Range Weather Forecasts (ECMWF) ERA-Interim on a  $1^\circ \times 1^\circ$  grid (Dee et al., 2011). The variables included daily 500 hPa geopotential height (Z500), horizontal winds at 300 and 500 hPa, SST, SIC and sea-level pressure (SLP). We also used monthly SST and SIC data from the Hadley Centre Global Sea Ice and Sea Surface Temperature (HadISST) dataset (Rayner et al., 2003).

We used the normalized NAO index and EA/WR index from the National Oceanic and Atmospheric Administration/Climate Prediction Center (NOAA/CPC). Our investigation also employs a second NAO index, which is a modification of the NAO index proposed by Li and Wang (2003). Their index was taken as the difference in the normalized SLP, zonally averaged from  $80^\circ$  W to  $30^\circ$  E, between  $35^\circ$  N and  $65^\circ$  N. They commented that this measure “provides a much more faithful and optimal representation of the spatial–temporal variability associated with the NAO”. We have slightly modified the Li and Wang (2003) definition by conducting our sector averaging over  $80^\circ$  W– $10^\circ$  W, instead of the original  $80^\circ$  W– $30^\circ$  E, and the newly calculated index is called the western-type NAO (WNAO) index. This modification is made because the longitudinal range of Li and Wang (2003) includes both the North Atlantic and part of the European continent. Atmospheric circulations

associated with Beijing haze in our investigation involve both a blocking anticyclone over Europe and an NAO+ pattern (Figure 1a). When the anticyclone over Europe and the NAO+ pattern occur concurrently, the center of the north pole of the NAO+ pattern is shifted westward, which displays a western-type NAO+ type pattern. Although the NAO pattern is coupled with the anticyclone over Europe, the modified WNAO index could better reflect the strength of the north–south dipole over the Atlantic, regardless of the impact of the anticyclone over Europe. This concept was first proposed by Yao and Luo (2014), who divided the NAO into the eastern-type NAO (ENAO) and the western-type NAO (WNAO). They proposed this definition for the purpose of revealing the different relationships between these two types of NAO and the downstream blocking (e.g. the Euro–Atlantic blocking), temperature and precipitation. Some detailed NAO studies (Yao and Luo 2014; Luo et al. 2014; Yao et al. 2016) have also revealed the disadvantages of the conventional NAO indices in the identification of the spatial structures of the NAO pattern, and have suggested that the zonal position and inclination of the NAO dipole could change the occurrence of extreme weather events. Below, we will see that the WNAO index has a closer relationship to the Beijing haze frequency and better represents the background circulation of the persistent haze events.

## **2.2 Definition of persistent haze events**

Haze is generally defined in terms of relative humidity and visibility (World Meteorological Organization, China Meteorological Administration, UK Met Office). Some studies also define haze in terms of thresholds of PM<sub>2.5</sub> concentrations (Cai et al. 2017; Callahan and Mankin 2020). As the PM<sub>2.5</sub> data are not available for a long time period and thus cannot cover our study period, we use a definition more applied to meteorological range. In the “Specifications for surface meteorological observations” compiled by the China Meteorological Administration (2004), haze is defined as a

weather phenomenon with a large amount of extremely fine dust particles evenly floating in the air that reduces the horizontal visibility to less than 10 km. As visibility can also be reduced by fog, a relative humidity threshold is also applied to distinguish hazy weather from fog. Many previous studies (Wu, 2006, 2008; Yang et al., 2016; Pei et al., 2018; He et al., 2018) have proposed that fog is associated with a relative humidity greater than 90%. Therefore, we use a daily mean relative humidity of less than 90% and a visibility of less than 10 km as the criteria to define winter haze days in this study (Pei et al., 2018; Chang et al., 2020). Persistent haze events are defined as periods for which haze occurs for at least five consecutive days. The day with the minimum visibility within a persistent haze event is defined as Day 0, which is the most prominent haze day, and the days before (after) Day 0 signify the growth (decay) of the haze event.

### 3 Large-scale atmospheric circulations

#### 3.1 Interannual timescale

To show the context of the relationship between Beijing winter haze weather and atmospheric circulations, in Fig. 1a we show the time series of the number of winter haze days (blue line) and detrended winter haze days (red line) from 1980 to 2016. The linear trend (+1.9 days per decade) of the number of winter haze days in Beijing is not significantly different from zero ( $P = 0.10$ ), which is consistent with previous studies (Chen and Wang, 2015; Pei et al., 2018, 2020), and its interannual variability is quite large. Many factors could affect the interannual variation in the number of winter haze days, including changes in emissions, emission reduction measures and climate variables (e.g., meteorological conditions, atmospheric circulation and SSTs) (Dang and Liao, 2019; Wang et al., 2020; Pei et al. 2020). Pei et al. (2020) pointed out that anthropogenic emissions showed an increasing trend before 2012 and a decreasing trend thereafter, which can be attributed to a Clean Air Action Plan



181 introduced in 2013, and these trends with opposite sign could explain the absence of a significant trend  
182 in the number of hazy days. Thus, we undertake our investigation with detrended data to explore the  
183 influences that are associated with the interannual variation in the number of hazy days. In addition,  
184 the long-term trends are removed for winter haze days and other variables in the following analyses.

185 We focus first on the role of atmospheric circulation in inducing hazy weather on the interannual  
186 timescale. The detrended winter haze days are significantly correlated with circulation patterns in the  
187 Z500 anomaly field (Fig. 1b). A wave train of wavenumber-3 structure dominates the mid-high  
188 latitudes; three cyclones are situated over the Greenland region, the Ural region and the Sea of Okhotsk  
189 with positive geopotential height anomalies in between. The NAO+ and EA/WR+ teleconnections  
190 patterns have been suggested to be connected with winter haze days in Beijing (Yin and Wang, 2017;  
191 Yin et al., 2017; Chen et al., 2019). Barnston and Livezey (1987) have shown that the NAO+ pattern  
192 is made up of negative geopotential height anomalies in the high latitudes of the North Atlantic  
193 (Greenland) and positive anomalies over the central North Atlantic, extending into the eastern United  
194 States and western Europe. The EA/WR+ pattern is associated with positive geopotential height  
195 anomalies over Europe and northern China, and negative anomalies located over the central North  
196 Atlantic and the north of the Caspian Sea. Following Barnston and Livezey (1987), we can say that  
197 Fig. 1b demonstrates a western-type NAO+ pattern over the North Atlantic, as this pattern is situated  
198 farther west and co-occurs with a blocking anticyclone over Europe. In addition, a quadrupole mode  
199 from the North Atlantic to northern China shares some similarities with an EA/WR+ pattern, but there  
200 are some differences. On the one hand, two negative geopotential height anomalies are located over  
201 the northern North Atlantic and the northern Ural region, which are farther north than the two cyclones  
202 of the EA/WR+ pattern. On the other hand, the EA/WR+ pattern shows a zonal wave train structure;

and only when the cyclone (north center of the western-type NAO+ pattern) over the North Atlantic is excluded can a tripolar mode from Europe to northern China manifest a zonal wave train structure.

To further understand the relationships between winter haze days in Beijing and the teleconnection patterns of the NAO and EA/WR, we present the annual variations in winter haze days, the NAO index and EA/WR index from NOAA/CPC, and the WNAO index in Fig. 2. Their correlation coefficients are calculated over the time periods 1980–2016, 1980–1999 and 2000–2016. The correlations between winter haze days and the NOAA/CPC NAO index are 0.27, 0.16 and 0.33, respectively, and only the correlation for the entire 1980–2016 period is significant at the 90% confidence level (Fig. 2a). Correlation coefficients between winter haze days in Beijing and the WNAO index are considerably larger for the 1980–2016 and 2000–2016 periods, being 0.42 ( $P < 0.01$ ) and 0.61 ( $P < 0.01$ ), respectively, while the correlation coefficient (0.08) is much smaller (and not significant) for the period 1980–1999 (Fig. 2b). This indicates that the WNAO index more strongly reflects the relationship between the north–south dipole mode over the North Atlantic and winter haze days in Beijing. The correlation with the EA/WR index is slightly lower than that with the WNAO index for the entire period (0.36,  $P < 0.05$ ), but considerably higher (0.57,  $P < 0.01$ ) when only the first 20 years of the record are considered (Fig. 2c). Overall, we see that the EA/WR+ pattern has a closer relationship with hazy weather in Beijing before 2000, whereas the western NAO+ pattern is dominant after that date. It is also worth noting that the extreme numbers of winter haze days more closely correspond with the magnitude of the WNAO index (Fig. 2d–f). We can see that the extremely low numbers of hazy days in the winters of 1995, 2003 and 2010 occur simultaneously with low (normalized) WNAO values of  $-1.11$ ,  $-1.11$  and  $-1.61$ , respectively, whereas the winter of 2013 with an extremely high number of hazy days has a large WNAO value of 2.42 (Fig. 2e). However, such

correspondences are weaker for the NOAA NAO index (Fig. 2d) and could not be found in the case of the EA/WR index (Fig. 2f).

### 3.2 Daily-to-weekly timescale

To cast further light on the above results, we have identified all the persistent haze events during 1980–2016 and explored the daily progression of the circulation structures that led up to (up to 10 days before) and followed (out to 8 days) the 65 identified persistent haze episodes. To accomplish this, we formed composites (across the 65 episodes) of daily Z500 anomalies and the horizontal components of wave activity flux (Takaya and Nakamura, 2001) (Fig. 3). From Day –10 to Day –8, a WNAO+ pattern can be identified over the North Atlantic, and the wave activity flux propagates downstream from the northern pole of the WNAO+ pattern to a weak anticyclone over Europe. Compared with the NAO pattern in the study by Li and Wang (2003), the north–south dipole mode of the WNAO pattern here is situated farther to the west and is accompanied by an anticyclone over Europe. These structures further emphasize the value of using our modified NAO index in this study. From Day –6 to Day –4, the weak anticyclone over Europe strengthens and a cyclone develops over western Russia following the propagation of wave activity. Simultaneously, the WNAO+ pattern weakens and shows a northeast–southwest inclination. From Day –3 to Day –2, an EA/WR+ pattern starts to appear in the mid to high latitudes. At this time, a zonal wave train that propagates from a cyclone over the North Atlantic, through an anticyclone over Europe and a cyclone over the west of Lake Baikal, to East Asia leads to the formation of an anticyclone over northeastern China. On Day –1, an anticyclone forms over the Gulf Stream region, which continues to provide wave activity to the development of downstream circulation. This anticyclone sustains the wave train propagating to northeastern China. As a result, the anticyclone over northeastern China persists for a total of 7 days from Day –3 to Day

3 and then moves eastward to the North Pacific. The wave train structure dissipates from Day 4, and the anomalous atmospheric circulation around Beijing becomes weaker.

From the daily evolution of atmospheric circulation, we find that certain teleconnection patterns (WNAO+ and EA/WR+) and wave trains lead to persistent haze events in Beijing. In the whole process, the WNAO+ pattern seems to be the origin of the transmission and the EA/WR+ pattern acts as a subsidiary mode that transfers the wave activities from the WNAO+ pattern to the downstream anticyclone over northeastern China. This sequence is consistent with the variations of zonal winds, as they act as waveguides for the propagation of wave trains (Ambrizzi et al., 1995; Fang et al., 2001; Athanasiadis et al., 2010; Wang and Zhang 2015; Martinez-Asensio et al., 2016; Wirth et al., 2018; Rudeva and Simmonds 2021). Zonal wind anomalies at 300 hPa are composited between Day -10 and Day 5 for the 65 persistent haze events to explain the evolution of large-scale circulations and to ascertain whether potential predictors exist (Fig. 4a). Corresponding to the propagation of the WNAO+ and EA/WR+ patterns, significantly intensified or weakened anomalous zonal wind centers can be clearly identified from the North Atlantic to East Asia. The negative–positive–negative zonal wind tripolar mode over the North Atlantic reflects the WNAO+ pattern, whereas the arched structure of the anomalous zonal wind from Greenland, passing through high latitudes to northern China, manifests in the EA/WR+ pattern. To track the daily variations in anomalous zonal winds, significantly increased zonal wind anomalies over the central North Atlantic, the Scandinavian Peninsula and the north of China are used as three indicators (Fig. 4b). The amplitude of the anomalous zonal wind over the central North Atlantic peaks at Day -8, which corresponds to the dominance of the WNAO+ pattern from Day -8 to Day -10 in Fig. 3. Although the zonal wind in this region weakens after Day -8, it retains significantly high values until Day 4. Simultaneously with the significantly increased zonal

269 wind over the central North Atlantic, the difference in the zonal wind over the south of the  
270 Scandinavian Peninsula and the zonal wind over North China also show significant positive anomalies  
271 starting from Day -11 and Day -5, respectively. The difference in the zonal wind over the south of the  
272 Scandinavian Peninsula reaches its highest point on Day -2 when the EA/WR+ pattern can be clearly  
273 identified in Fig. 3, whereas the anomalous zonal wind over the north of China reaches a peak on Day  
274 -1. From the analyses above, we can see that the three indicators of anomalous zonal winds well reflect  
275 the sequential occurrence of teleconnection patterns related to persistent haze events.

276 The above analysis demonstrates that the formation of the persistent haze weather in Beijing is  
277 closely related to the WNAO+ and EA/WR+ patterns. Although both patterns are essential in the  
278 persistent haze formation, the WNAO+ pattern plays the role of the starting point in the atmospheric  
279 transmission and leads to the formation of the persistent haze over a longer time period. In the next  
280 section, we determine whether the WNAO+ pattern is a harbinger of more persistent haze on  
281 interannual timescales and explore the extent to which the WNAO+ pattern can improve the haze  
282 prediction on the daily-to-weekly timescale.

## 283 **4 The WNAO pattern and persistent Beijing haze**

### 284 **4.1 Interannual timescale**

285 We first compare some characteristics of winter haze that occur in WNAO+ winters and WNAO-  
286 winters. The WNAO+ winters and WNAO- winters are classified with four sets of criteria with a view  
287 to making our results robust. We do this by stratifying the winters into above the 50th, 75th, 90th and  
288 95th percentiles (WNAO+ winters) and below the 50th, 25th, 10th and 5th percentiles (WNAO-  
289 winters). Figure 5a shows that there are more haze days in WNAO+ winters than in WNAO- winters.  
290 For the 50th, 75th, 90th and 95th percentiles, the average number of haze days for WNAO+ winters

291 are 4.5%, 23.2%, 57.5% and 38.6% more than their counterparts for WNAO– winters, and 2.3%, 7.4%,  
 292 22.8% and 26.0% more than the average numbers of all winters (the baseline) (Fig. 5a). This suggests  
 293 that the WNAO+ winters, especially winters with large-magnitude positive WNAO indices, are  
 294 conducive to haze days. For the persistent haze days (that is, haze days within persistent haze events),  
 295 their numbers for the WNAO+ winters are also greater than for WNAO– winters, and the same is true  
 296 for the persistent haze events (Fig. 5b,c). The differences are especially large for the 90th/10th and  
 297 75th/25th percentiles. For these two sets of criteria, the numbers of persistent haze days in WNAO+  
 298 winters are 167.9% and 101.5% more than that in WNAO– winters (Fig. 5b), whereas the numbers of  
 299 haze events in WNAO+ winters are 80.0% and 80.0% more than that in WNAO– winters (Fig. 5c).  
 300 Compared with the average numbers of persistent haze days (events) for all the winters, the numbers  
 301 for WNAO+ winters are increased by 14.2%, 24.3%, 57.7% and 80.8% (7.9%, 13.9%, 28.1% and  
 302 42.3%) at the 50th, 75th, 90th and 95th percentiles, respectively. The duration of haze events can be  
 303 derived by dividing the number of persistent haze days by the number of haze events. As can be seen,  
 304 the WNAO+ winters also correspond to longer persistent haze events, with the first two longest average  
 305 durations, 8.3 and 8.6 days, in the winters with WNAO indices larger than the 90th and 95th percentiles,  
 306 which are 2.9 and 2.7 days longer than that of winters with WNAO indices less than the 10th and 5th  
 307 percentile, respectively (Fig. 5d). These findings indicate that the WNAO+ pattern is conducive to the  
 308 occurrence and the persistence of winter haze in Beijing, whereas the WNAO– pattern militates against  
 309 these. These findings indicate that if we predict a strong WNAO+ winter (over the 95th percentile),  
 310 we are able to prognose a 26.0% increase in haze days, a 42.3% increase in haze events and 1.8 day  
 311 increase in haze duration compared with the condition that we do not know the prior state of the  
 312 WNAO.

## 4.2 Daily-to-weekly timescale

To test whether the predictability of haze days and haze events can be improved via the WNAO, we first compare the probability of haze days under three categories of conditions: the state of WNAO is unknown (the baseline), the WNAO is positive and the WNAO is negative. The WNAO we use here is a five-day (a pentad) moving average calculated by dividing the consecutive five-day sum of the WNAO index by five. To determine the probabilities of haze at different lead times, we take the last day of a pentad as Day 0, take the first day after the pentad as Day 1 and so on. We present the baseline probability of haze and the change in probability relative to the baseline when the WNAO is above (below) the 95th, 90th, 75th and 50th percentiles (the 50th, 25th, 10th and 5th percentiles) from Day 0 to Day 10 in Fig. 6. The baseline probability of a haze day for the 37 winters ranges between 42% and 43% (Fig. 6a). Moreover, the probabilities of haze for all percentiles are increased (decreased) given the positive (negative) WNAO condition, and most are significant at the 0.01 level (Fig. 6b,c). Compared with the 50th and 75th percentiles, larger increases in probability are found for the 95th and 90th percentiles, the peaks of which are at Day 8, being 16.9% and 13.0%, respectively. This suggests that the predictability of Beijing haze can be improved by 16.9% (13.0%) at Day 8 if the earlier five-day average WNAO is known to be larger than the 95th (90th) percentile. Moreover, the day on which the maximum probability occurs corresponds well with the obvious WNAO pattern at Day -8 of the composite haze events (Fig. 3). Conversely, a small WNAO index implies fewer winter haze days, especially at Day 5 for the 5th and 10th percentiles, being 19.0% and 12.0%, respectively. Hence, the earlier WNAO condition is conducive to the improvement of the prediction of winter haze day in Beijing.

As an extension of Fig. 6, Fig. 7 shows the proportions of persistent haze days in the haze days

under different WNAO states. If the WNAO state is unknown, the percentage of persistent haze days in the haze days varies between 29% and 32% (Fig. 7a). However, if the WNAO index is larger than the 95th, 90th, 75th and 50th percentiles, the percentage of persistent haze days increases to 50.6%, 45.1%, 45.8% and 37.5% on Day 7, Day 7, Day 8 and Day 6, respectively (Fig. 7b). This suggests that the positive WNAO pattern not only favors more haze days but also predicts a greater number of persistent haze days and haze events. However, if the WNAO is negative, persistent haze days and haze events are less likely to occur (Fig. 7c).

## **5 External influence of SST and SIC**

### **5.1 Interannual timescale**

We now determine the SST pattern in the North Atlantic and the SIC pattern in the Greenland Sea and Barents–Kara Sea, which might be related to persistent Beijing haze on the interannual timescale. Below we use the number of persistent haze days to categorize winters with more haze days and fewer haze days. This measure reflects the number of persistent haze events as well as their duration. The composite distributions of SST anomalies and SIC anomalies of 11 winters (1988, 1996, 1998, 2005, 2007, 2008 and 2011–15) with more than 15 persistent haze days and 11 winters (1980, 1982–84, 1986, 1987, 1994, 1995, 1999, 2003 and 2010) with no more than 5 persistent haze days, and their differences are shown in Fig. 8. The SST anomalies associated with the most persistent haze days show a north–south tripolar mode in the North Atlantic, with positive anomalies in the Gulf Stream region and negative anomalies at the southwest of the Canary Islands and regions around Greenland, which is consistent with the WNAO+ pattern (Fig. 8a). Simmonds and Govekar (2014) noted that warm SSTs in the Gulf Stream region give rise to a teleconnection pattern with a node in the Barents Sea and another over Eurasia. Significant SST anomalies can also be seen in the northwestern Pacific, with



negative SST anomalies from Yellow Sea to the Sea of Japan and positive SST anomalies to the east of the Philippines. For SST anomalies of winters with fewer persistent haze days, a north–south tripolar mode in the North Atlantic opposite to that in Fig. 8a is found, although the anomalies are not significant (Fig. 8b). In addition, significant positive SST anomalies are located in the Greenland Sea and the Barents Sea (Fig. 8b). The structure of the SST anomalies in the northwest Pacific is similar to that in Fig. 8a, but with much smaller magnitude. Figure 8c, which shows the difference between these two anomaly plots, further highlights the distribution of SST anomalies that are associated with more persistent haze days: a prominent north–south tripolar mode over the North Atlantic and negative SST anomalies over the Greenland Sea and Barents Sea.

SIC anomalies in the Greenland Sea and Barents–Kara Sea also show significant differences between winters with the greatest number of persistent haze days and winters with the lowest number of persistent haze days (Fig. 8d–f). The Greenland Sea and the southern part of the Barents Sea are dominated by positive SIC anomalies and the Kara Sea shows negative SIC anomalies, although they are not significant (Fig. 8d). The only positive SIC anomalies that differ significantly from zero are found in the north of the Barents–Kara Sea and the southeast of Greenland (Fig. 8d). By contrast, for the lowest number of hazy days, composite significant negative SIC anomalies are located in the Greenland Sea and the south part of the Barents Sea (Fig. 8e). The difference between Fig. 8d and Fig. 8e also suggests that the decrease in sea ice in the Greenland Sea and the southern Barents Sea is not conducive to a greater number of persistent haze days or haze events (Fig. 8f).

## 5.2 Daily-to-weekly timescale

From the analyses on the interannual timescale, we can see that SST anomalies in the North Atlantic and SIC anomalies in the Greenland Sea and Barents–Kara Sea are associated with the

379 occurrence of persistent haze days or haze events. To examine the potential causalities, here we  
380 analyzed the composite SST anomalies and SIC anomalies averaged over the period of 10 days prior  
381 to the first day of the 65 haze events. Prior to the persistent haze events, SST anomalies exhibit a north–  
382 south tripolar mode in the North Atlantic (Fig. 9a). This pattern differs from that shown in Fig. 8a; the  
383 tripolar mode is located farther south and has significant positive anomalies around the Florida  
384 peninsula and even into the Caribbean Sea and significant negative SST anomalies off the northeast  
385 coast of the United States (Fig. 9a). Figure 9a also shows significant negative SST anomalies in the  
386 Greenland Sea and the Barents–Kara Sea. For the SIC composite anomalies for these 65 events, we  
387 observe significant increases in the Greenland Sea and the Barents–Kara Sea (Fig. 9b). These prior  
388 distributions of SST and SIC anomalies are conducive to setting up the atmospheric circulations and  
389 zonal winds that are favorable for the formation of persistent Beijing haze events. The SST tripolar  
390 mode in the North Atlantic is favorable for the formation of a WNAO+ pattern, and the strongly  
391 increased zonal wind in association with the WNAO+ pattern encourages downstream wave train  
392 propagation. In concert with this, the decreased SST and increased SIC in the Barents–Kara Sea region  
393 potentially enable a strengthening of the cyclone upstream of the anticyclone over northern China,  
394 which is an essential part of the downstream wave train. Furthermore, as in Fig. 8a, the significant  
395 negative SST anomalies in the northwestern Pacific are not compatible with the anticyclone directly  
396 controlling persistent haze events in Beijing.

397 In light of the above, it is noted that, even though regression and composite analyses (e.g., Fig. 1)  
398 might suggest a strong link between haze events and the WNAO+, it is important to remember that  
399 such analyses identify *necessary* but not *sufficient* conditions (Boschat et al., 2016). This leads us to  
400 the question of whether there are persistent haze events related to the WNAO– pattern and what the

differences are between these haze events and those related to the WNAO+ pattern. First, the daily evolutions of the NAO index for the 65 persistent haze events (black line), 43 WNAO+-related persistent haze events (red line) and 22 WNAO--related persistent haze events (blue line) are shown in Fig. 10. The composite WNAO index for all haze events reaches its highest value on Day -8, so we select a WNAO+-related (WNAO--related) persistent haze event if its averaged WNAO index during the period of Day -10 to Day -6 is above (below) zero. Figure 10a shows that the daily evolution of the composite WNAO index for WNAO+-related and WNAO--related persistent haze events generally shows the opposite character. However, the peak value (0.90) of the WNAO index for WNAO+-related persistent haze events appears on Day -8, the absolute value of which is larger than that of the valley value (-0.68) of the WNAO index for WNAO--related persistent haze events on Day -10. Corresponding SST and SIC anomalies of the WNAO+-related case and the WNAO--related case are also compared in Fig. 10b-d. We use the difference in the SST anomalies between the two regions A (40° N-55° N, 70° W-20° W) and B (22° N-37° N, 85° W-35° W) marked in Fig. 9a to signify the SST tripolar pattern in the North Atlantic. Daily evolutions of the SST anomaly difference show significant increases starting from Day -11 and Day -9 for the WNAO+-related case and all persistent haze events, respectively, whereas the difference in SST anomalies for the WNAO--related case shows only a weak fluctuation (Fig. 10b). As for the SIC anomaly in the Barents-Kara Sea, it has a significant maximum amplitude leading the Day 0 of WNAO+-related persistent haze events by 12 days (Fig. 10d), whereas the SIC anomaly in the Greenland Sea shows significant high values during the period from Day -15 to Day 15 of WNAO--related persistent haze events and its peak appears on Day -3 (Fig. 10c). We can conclude that the SST difference in the North Atlantic and SIC anomalies in the Barents-Kara Sea show preceding signals for WNAO+-related persistent haze events, and the

WNAO--related persistent haze events are associated mainly with increased SIC anomalies in the Greenland Sea.

## 6 Conclusions and discussions

In this study, we have focused on persistent winter haze in Beijing from the perspective of large-scale atmospheric circulation (e.g., teleconnections and wave trains). We first presented the atmospheric circulation correlated with winter haze in Beijing on the interannual timescale. It is found that the frequency of winter haze is related to a western-type NAO+ pattern and an EA/WR+ pattern on interannual timescales. Notably, the number of winter haze days is more relevant to the WNAO+ pattern after 1999, whereas it is more associated with the EA/WR+ pattern before 2000. Then, the daily evolution of the composite circulation patterns and wave activities for 65 persistent haze events is used to represent the relationship between winter haze weather in Beijing and atmospheric circulation on the daily-to-weekly timescale. A transmission of atmospheric circulation demonstrated a wave train propagating from North Atlantic to northeast China. The WNAO+ pattern is the initiator of this transmission from Day -10 to Day -8, followed by an EA/WR+ pattern from Day -3 to Day -2. We also highlighted the daily variations in zonal winds linked to atmospheric circulation. It was revealed that the zonal wind over the central North Atlantic, which is coupled with the WNAO+ pattern, reached its largest value on Day -9, and the zonal wind downstream over the Scandinavian Peninsula and the north of China showed significantly increased values after Day -5, indicating the propagation of the wave train.

As the WNAO+ pattern acted as the original driver of this atmospheric transmission, we were interested in whether it can improve the predictability of winter haze on both interannual and daily-to-weekly timescales. From a comparison of the number of haze days and haze events, and the average

445 duration of haze events for WNAO+ winters, WNAO− winters and all winters with four sets of criteria,  
446 we found that the haze days and haze events are more likely to occur in WNAO+ winters. The WNAO+  
447 winters contribute to a 22.8% (26.0%) increase in winter haze days, a 28.1% (42.3%) increase in winter  
448 haze events, 1.6- (1.8-) days increase in average haze duration relative to the baselines for the 90th  
449 (95th) percentile. Moreover, when compared to the WNAO− winters, these values are much higher.  
450 Thus, we propose that if the WNAO state can be predicted for a winter, then a general predication of  
451 the probability of haze days and haze events as well as the duration of haze events can be obtained.

452 On the daily-to-weekly timescale, our results also suggest that the WNAO+ pattern significantly  
453 improves the predictability of haze days and haze events. For a set of lead times (from Day 1 to Day  
454 10) and four sets of criteria, the probabilities of haze days lagging the WNAO+ condition are all  
455 increased compared with the baseline or the WNAO− condition. The largest increase occurs for the  
456 95th and 90th percentiles, and the largest increase, 16.9% and 13.0%, can be predicted at Day 8.  
457 However, if the WNAO index is lower than the 5th (10th) percentile, a (19.0%) 12.0% decrease of  
458 haze days can be predicted at Day 5. This provides more information for decision-makers for a given  
459 day, that is, if the five-day average of WNAO index of the day is known, they can determine the  
460 probability of haze for a set of days behind this day.

461 Further analyses revealed that the SST anomalies in the North Atlantic and SIC anomalies in the  
462 Greenland Sea and Barents–Kara Sea are also related to winter haze weather in Beijing on both  
463 interannual and daily-to-weekly timescales. On the interannual timescale, winters with a greater  
464 number of persistent haze days correspond to a north–south tripolar mode over the North Atlantic. On  
465 the daily-to-weekly timescale, a more southerly SST tripolar mode is also prominent over the North  
466 Atlantic, and the Greenland Sea and Barents–Kara Sea also show significant SST and SIC anomalies

10 days prior to the persistent haze events. For a more in-depth understanding, the relationship between the preceding WNAO pattern, SST and SIC anomalies and the persistent haze events was highlighted by comparing the WNAO+-related haze events and the WNAO--related events. We found that the WNAO+-related case corresponded to prior SST signals in the North Atlantic and increased SIC anomalies in the Barents–Kara Sea, whereas the WNAO--related case had large SIC anomalies only in the Greenland Sea.

Previous studies have shown that the large-scale atmospheric circulations are closely associated with the winter haze in Beijing from decadal timescales down to intra-seasonal timescales (Yin and Wang 2017; Yin et al. 2017; Chen et al. 2019). These studies well captured certain modes of atmospheric circulations that are significant during the haze episodes, but did not reveal their lead–lag relationships on a daily timescale as well as the role of atmospheric circulations in the prediction of winter haze. By analyzing the day-by-day evolutions, our investigation has revealed that a WNAO+ pattern, part of an atmospheric transmission, shows potential prediction abilities for persistent haze events. One step further, we find that prior knowledge of the WNAO state can significantly improve the predictability of haze days, and implies an increase in the predictability of persistent haze. As the external forces have longer memories and can affect the state of the NAO by air–sea interaction (Okumura et al. 2001; Frankignoul and Kestenare, 2005; Peng et al. 2002, 2003; Nie et al. 2019), it is possible to further improve the predictability of winter haze by predicting the WNAO state through external forces on both interannual and daily-to-weekly timescales. In addition, it is also worthwhile to determine the reasons why the WNAO index (EA/WR index) has a stronger correlation with winter haze weather in Beijing after 1999 (before 2000) from the perspective of climate change.

**Data availability.** Daily atmospheric and land-surface data were downloaded from the ECMWF ERA-Interim data archive (<http://www.ecmwf.int/en/research/climate-reanalysis/era-interim>) (ERA-Interim, 2017). Monthly sea surface temperature data and sea-ice concentration data are available from the Met Office Hadley Centre datasets (<https://www.metoffice.gov.uk/hadobs/hadisst/data/download.html>) (Met Office, 2017). The ground observations were taken from the National Meteorological Information Center of China (<http://data.cma.cn/>) (CMA, 2017). The NAO and EA/WR indices are from the NOAA Climate Prediction Center (<http://www.cpc.ncep.noaa.gov/data/teledoc/telecontents.shtml>) (CPC, 2017). The modified WNAO index can be obtained from the authors.

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

**Author contributions.** YY designed the study. ML created the figures. ML and YY wrote the manuscript. IS, DL and LZ gave constructive comments. LP provided and analyzed the observational station data.

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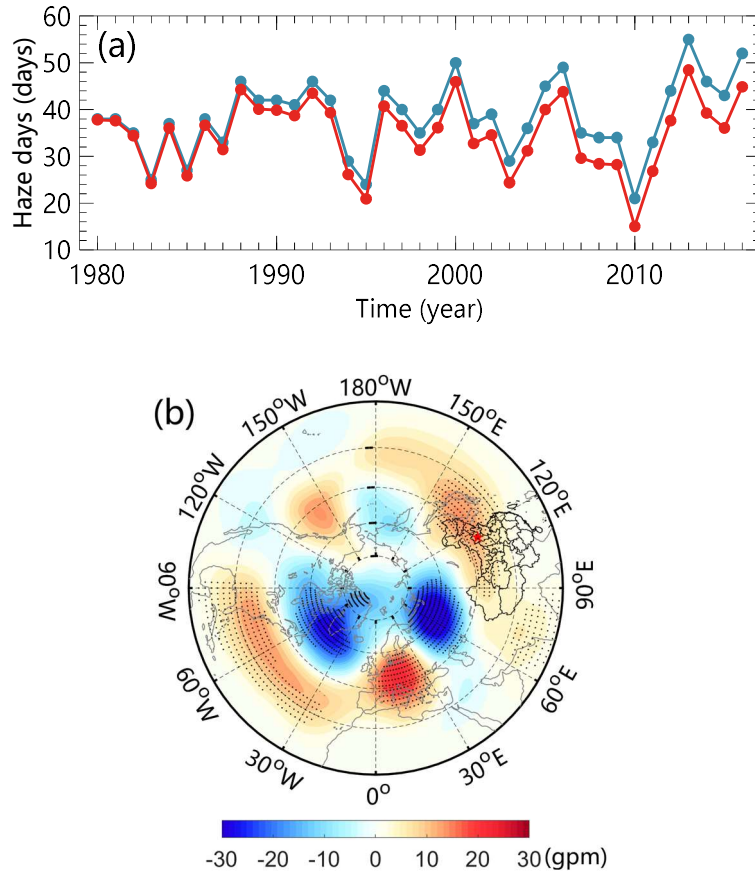
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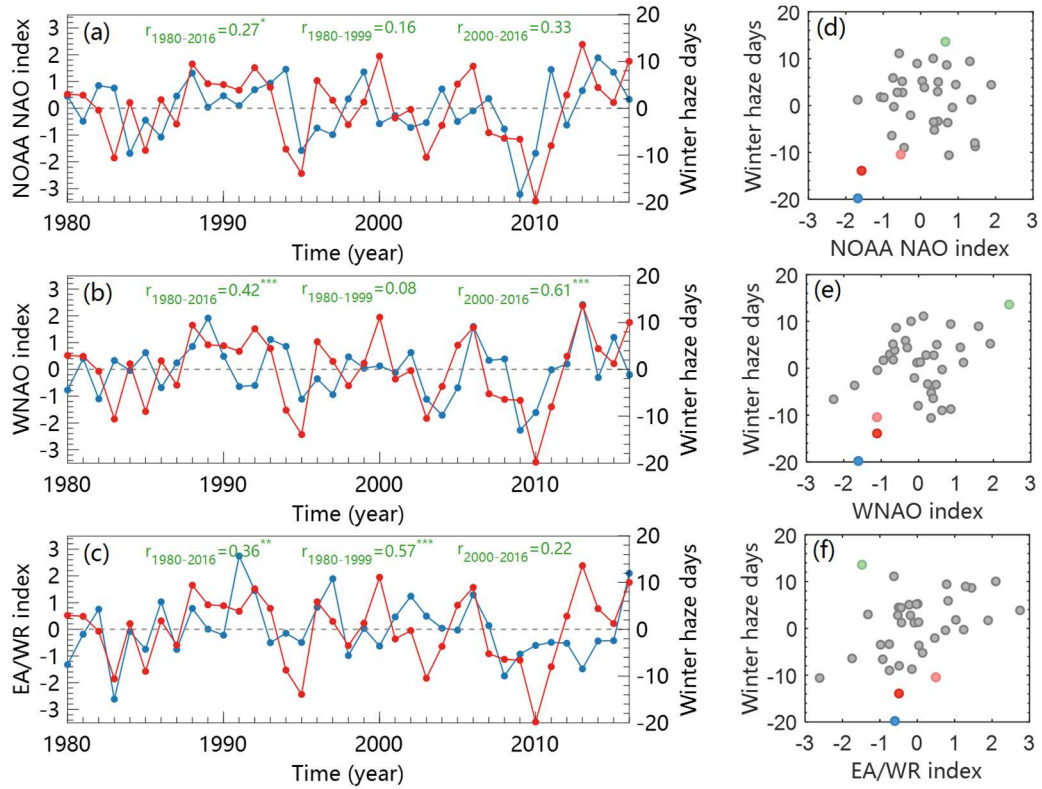
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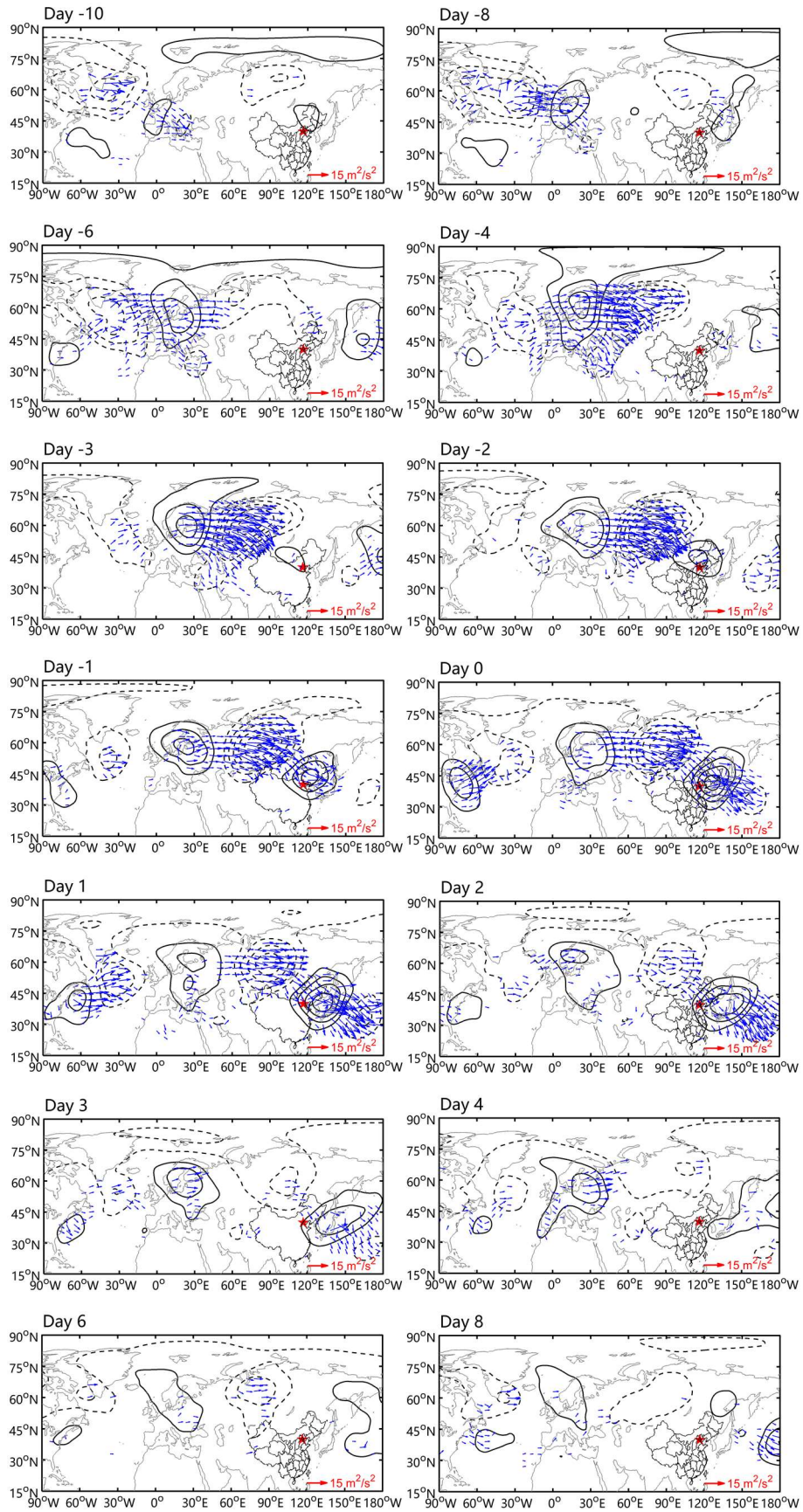




**Figure 1. (a)** Time series of winter haze days (blue) and detrended winter haze days (red) in Beijing from 1980 to 2016. To aid visualization, a fixed value is added to all the detrended data so that the first values of both series coincide. **(b)** Regressed Z500 anomalies (shading; units: gpm) against the detrended winter haze days. The dotted areas show values that are above the 95% confidence level based on a two-sided Student's  $t$ -test. For all plots in this study, the red star denotes the location of Beijing.



**Figure 2.** Time series of detrended winter haze days (red) and (a) NOAA NAO index (blue), (b) WNAO index (blue) and (c) NOAA EA/WR index (blue). Correlation coefficients between winter haze days and the three indices during the periods 1980–2016, 1980–1999 and 2000–2016 are labelled at the top of each panel. The 90%, 95% and 99% confidence levels for the Student’s *t*-test are denoted by one, two and three asterisks, respectively. Scatter diagrams of winter haze days against the (d) NOAA NAO index, (e) WNAO index and (f) EA/WR index, in which the red, pink blue and green dots represent the year 1995, 2003, 2010 and 2013, respectively.



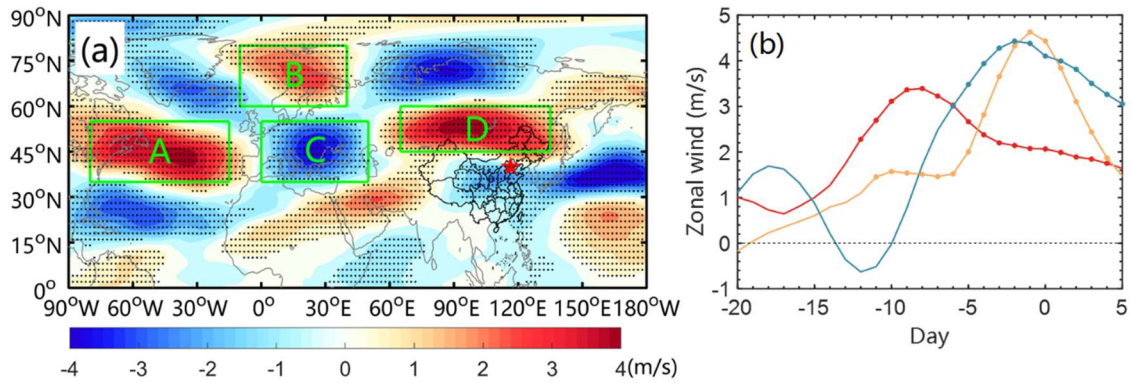
**Figure 3.** Instantaneous fields of the composite daily Z500 anomalies (contours, interval = 20; units:

740 gpm) and horizontal components of wave activity developed by Takaya and Nakamura (2001) (arrows;  
 741 units:  $\text{m}^2 \text{s}^{-2}$ ) from Day -10 to Day 8 for the 65 persistent haze events.

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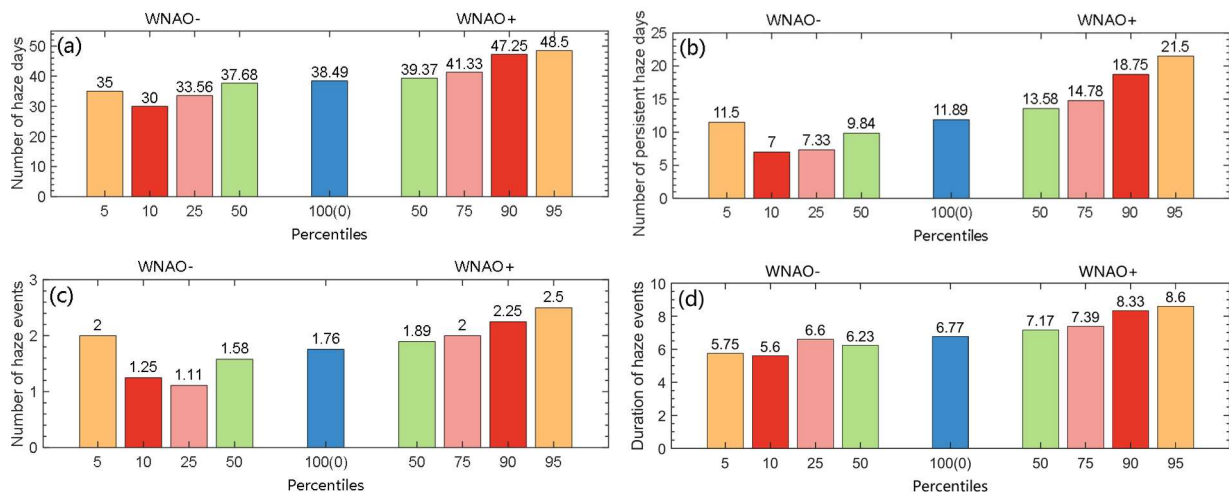
746 **Figure 4. (a)** Composite 300 hPa zonal wind anomalies (shading; units:  $\text{m s}^{-1}$ ) between Day -10 and  
 747 Day 5 for the 65 persistent haze events. The dotted areas show values that are above the 95%  
 748 confidence level based on a two-sided Student's *t*-test. **(b)** Composite daily time series of region-  
 749 averaged 300 hPa zonal wind anomalies in region A (80° W–15° W, 35° N–55° N) (red) and region D  
 750 (65° E–135° E, 45° N–60° N) (orange) and the difference in region-averaged zonal wind anomalies  
 751 between region B (10° W–40° E, 60° N–80° N) and region C (0° W–50° E, 35° N–55° N) (blue). The  
 752 dots denote the days with values that are above the 95% confidence level for the two-sided Student's  
 753 *t*-test.

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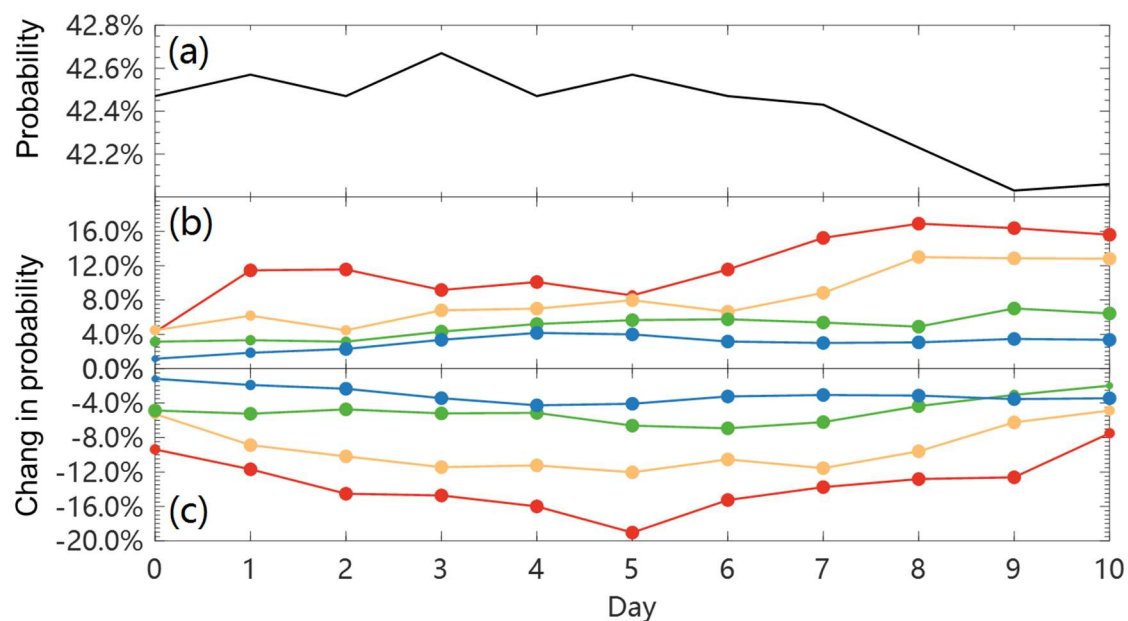
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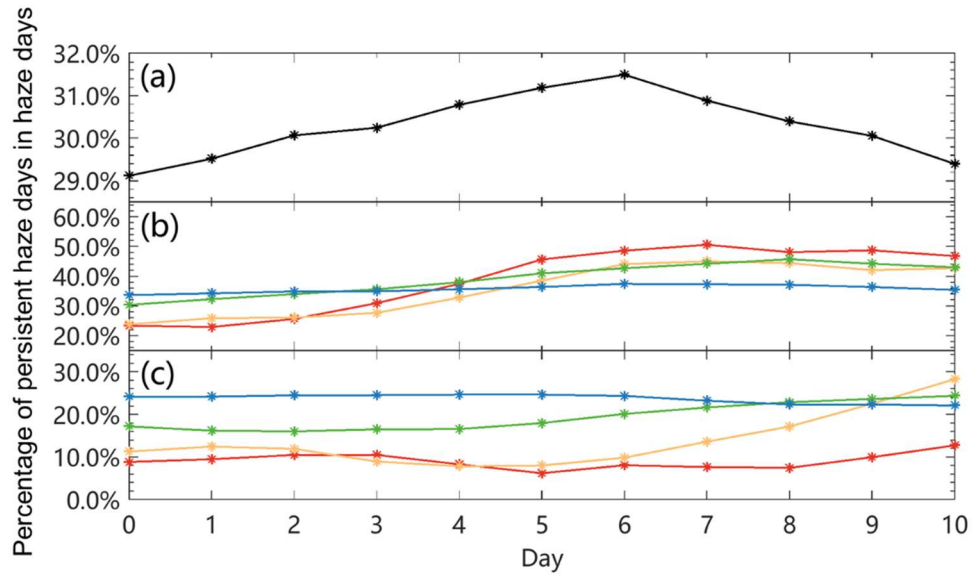


**Figure 5.** Average number of (a) haze days, (b) persistent haze days, (c) haze events and (d) average duration of haze events (days) during WNAO- winters and WNAO+ winters. The orange, red, pink and green bars on the left and right denote winters with a WNAO index larger than the 5th, 10th, 25th and 50th percentiles (WNAO- winters) and winters with the WNAO index smaller than the 95th, 90th, 75th and 50th percentiles (WNAO+ winters), respectively. The blue bars in the middle denote the results of all the winters. The persistent haze days are defined as haze days belonging to haze events.

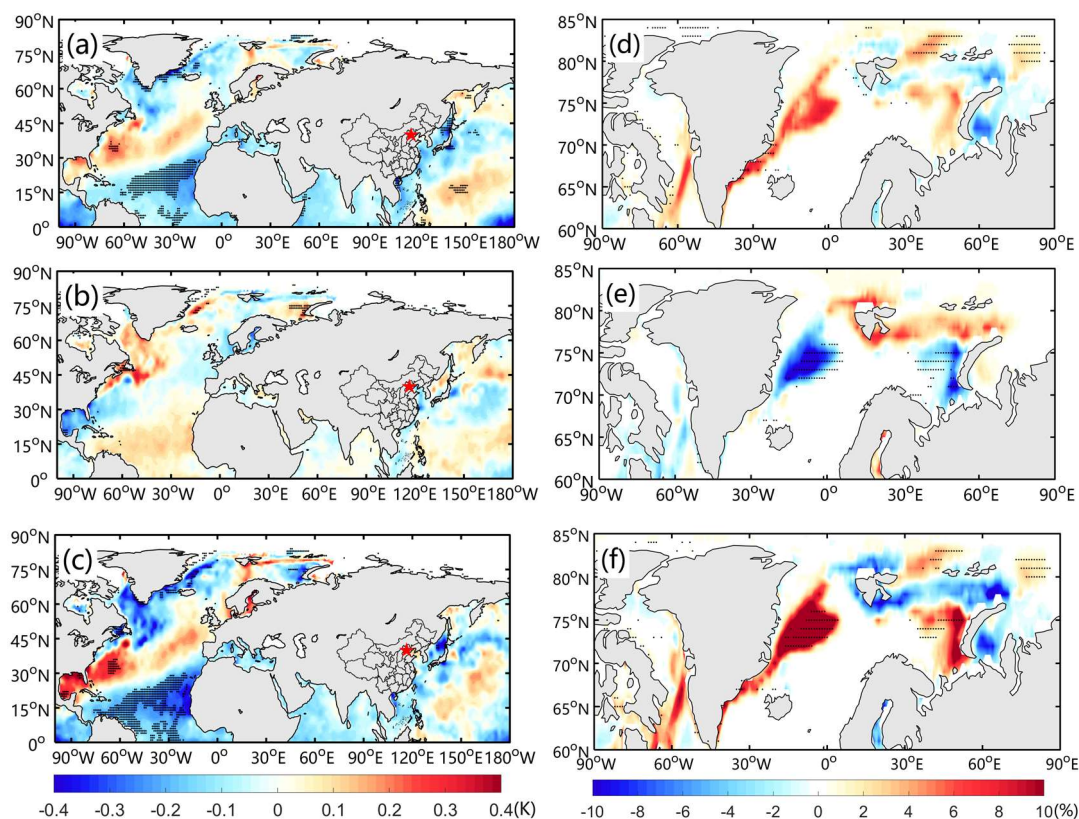




**Figure 6.** The probability of haze day at Day  $i$  ( $i = 0 \dots 10$ ) when (a) the state of the WNAO is unknown. The change in probability of haze day compared with (a) on the condition that (b) the five-day moving average WNAO index is above the 95th (red), 90th (yellow), 75th (green) and 50th (blue) percentile of the moving average series, and (c) the five-day moving average WNAO index is below the 50th (blue), 25th (green), 10th (yellow) and 5th (red) percentiles of the moving average series. The five-day moving average is the five-day sum of the WNAO index divided by five. Day 0 is defined as the last day of each consecutive five days. The small, medium and large dots represent the probability changes that are above the 90%, 95% and 99% confidence levels, respectively, for a Monte Carlo test with 10,000 simulations.

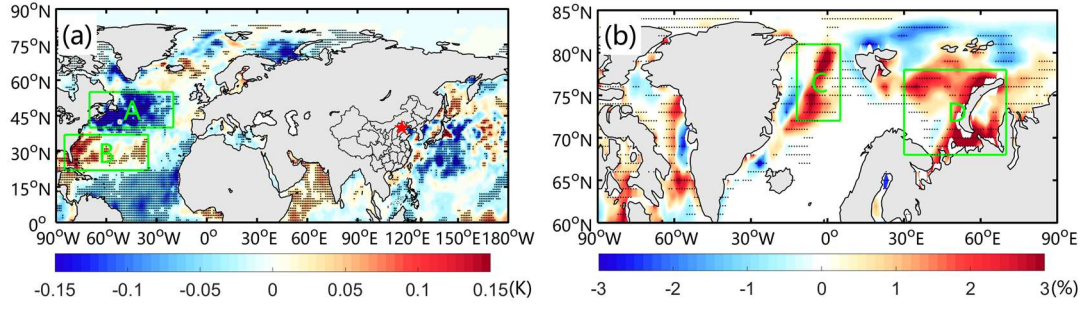


**Figure 7.** The percentage of persistent haze days in haze days at Day  $i$  ( $i = 0 \dots 10$ ) on the condition that (a) the state of the WNAO is unknown, (b) the five-day moving average WNAO index is above the 95th (red), 90th (yellow), 75th (green) and 50th (blue) percentiles of the moving average series, and (c) when the five-day moving average WNAO index is below the 50th (blue), 25th (green), 10th (yellow) and 5th (red) percentiles of the moving average series.

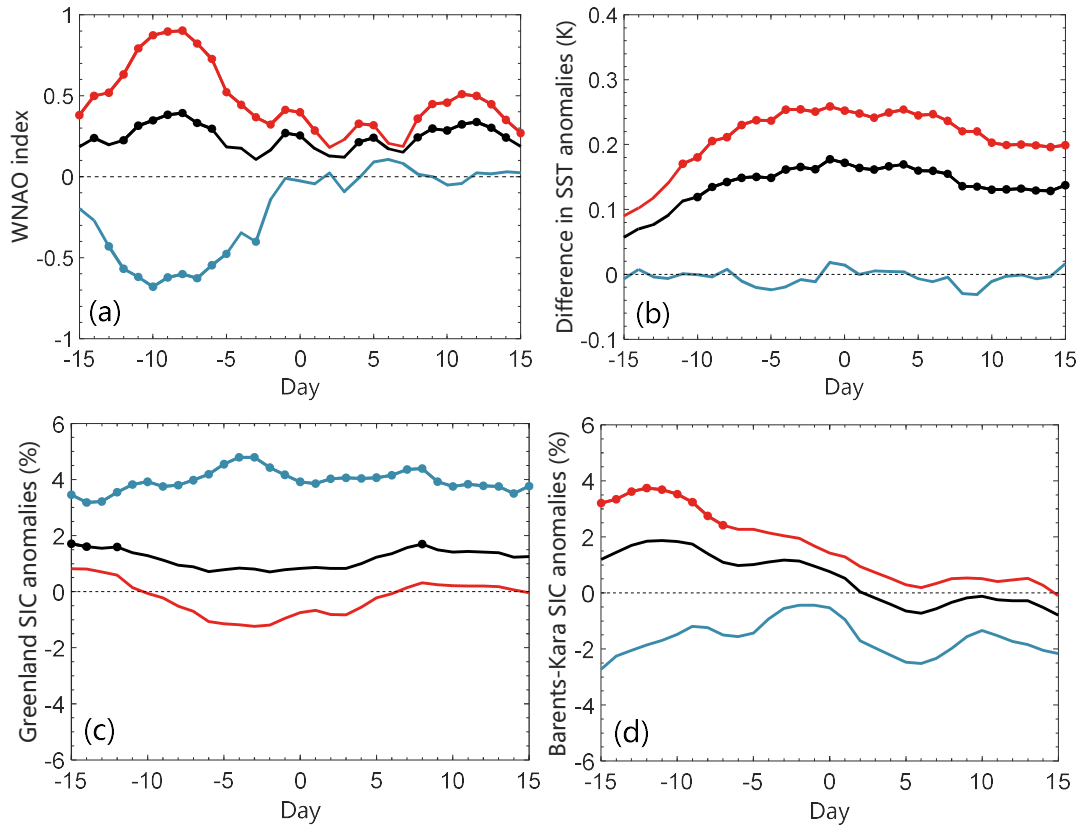


**Figure 8.** Composite SST anomalies and SIC anomalies for **(a, d)** winters (1988, 1996, 1998, 2005, 2007, 2008, 2011–2015) with the greatest number of persistent haze days and **(b, e)** winters (1980, 1982–1984, 1986, 1987, 1994, 1995, 1999, 2003, 2010) with the lowest number of persistent haze days, and **(c, f)** their differences. The dotted areas show values that are above the 95% confidence level based on a Monte Carlo test with 10,000 simulations.





**Figure 9. (a)** Composite SST anomalies and **(b)** SIC anomalies for the 10 days before the first day of the 65 persistent haze events. The dotted areas show values that are above the 95% confidence level based on a two-sided Student's  $t$ -test. The green rectangles A and B in **(a)** denote a north region ( $40^{\circ}$  N– $55^{\circ}$  N,  $70^{\circ}$  W– $20^{\circ}$  W) and a south region ( $22^{\circ}$  N– $37^{\circ}$  N,  $85^{\circ}$  W– $35^{\circ}$  W) over the North Atlantic, and the green rectangles C and D in **(b)** denote the Greenland Sea ( $12^{\circ}$  W– $5^{\circ}$  E,  $72^{\circ}$  N– $81^{\circ}$  N) and the Barents–Kara Sea ( $30^{\circ}$  E– $70^{\circ}$  E,  $68^{\circ}$  N– $78^{\circ}$  N), respectively.



**Figure 10.** Composite daily time series of the (a) WNAO index, (b) difference in the SST anomalies over the two regions, as presented in Figure 9a, (c, d) region-averaged SIC anomalies in the (c) Greenland Sea (marked by C in Figure 9b) and (d) Barents–Kara Sea (marked by D in Figure 9b) for the 65 persistent haze events (black), 43 WNAO+-related persistent haze events (red) and 22 WNAO--related persistent haze events (blue) from Day -15 to Day 10. The difference in the SST anomalies is calculated by subtracting the region-averaged SST anomalies in the north (rectangle A) from the region-averaged SST anomalies in the south (rectangle B) (that is, B minus A). A persistent haze event that is related to the WNAO+ (WNAO-) is selected when its averaged WNAO index is positive (negative) from Day -10 to Day -6 of the persistent haze event.