

1 **Three dominant synoptic atmospheric circulation** 2 **patterns influencing severe winter haze in eastern China**

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12
13 **Abstract.** Previous studies indicated that, on synoptic scale, the severe haze in eastern China (EC)
14 is affected by the atmospheric circulation variations. However, it is still unclear what are the
15 dominant atmospheric circulation patterns influencing the severe winter haze conditions in EC and
16 what are the differences between them. To systematically determine the dominant synoptic
17 atmospheric circulation patterns of severe haze in different regions of EC, we use the Hierarchical
18 Clustering Algorithm to classify the local geopotential height anomalies at 500-hPa over the stations
19 with severe haze and obtained three dominant synoptic atmospheric circulation types based on
20 observed PM_{2.5} concentration and NCEP/NCAR reanalysis. Circulation Type1 is accompanied by
21 significant north wind component anomalies over northern China and causes severe haze pollution
22 over the Yangtze River valley. Although the local meteorological conditions are not conducive to
23 haze formation and accumulation, the severe haze in Yangtze River valley is related to the pollution
24 transportation caused by the north wind anomalies. During the haze days with circulation Type2,
25 the joint affection of East Atlantic-West Russia teleconnection pattern and winter East Asia
26 subtropical jet stimulate and maintain the anticyclonic anomalies over northeast Asia, which
27 provides meteorological conditions conducive to the occurrence of severe haze over the whole EC.
28 The circulation Type3 mainly caused severe haze events in northeast China through the
29 establishment of blocking high over the Okhotsk Sea. The results provide a basis for establishing
30 haze prediction and management policies applicable to different regions in EC.

31 1. Introduction

32 Severe haze could increase the risk of traffic accidents by reducing visibility and harm human
33 health by causing respiratory diseases (Xie et al., 2014; Hu et al., 2015; Wang et al., 2016). Haze
34 events in China are mainly caused by PM_{2.5} (particulate with an aerodynamic diameter less than
35 2.5 μ m; Cai et al., 2017; Shen et al., 2018; Wang et al., 2021). Researches show that the distribution
36 of haze days in China has the characteristics of uneven spatial distribution, with more in
37 economically developed eastern region and less in economically underdeveloped region (Wu et al.,
38 2013; Liu et al., 2015; Xu et al., 2015). With the rapid development of industrialization, urbanization
39 and increase in anthropogenic emission, eastern China (EC) has experienced more severe haze
40 events with long duration, large spatial scale, and serious harm in the past few decades (Monks et
41 al., 2009; Qian et al., 2009; Wang et al., 2009). Since the beginning of the 21st century, the uneven
42 spatial distribution of haze events in China have become more obvious (Sun et al., 2016), which has
43 led to the increasing incidence rate and mortality related to respiratory diseases in Beijing-Tianjin-
44 Hebei, the Yangtze River valley, and the Pearl River Delta (Tsaia et al., 2014; Ding et al., 2016; Fan
45 et al., 2019). Although haze pollution control in China has been improved to some extent with the
46 strict implementation of energy conservation and emission reduction policies after 2013 (Wang et
47 al., 2021), it still affects various socio-economic sectors and human health.

48 In addition to human activities, meteorological conditions are also considered as one of the most
49 important factors for determining regional air quality. Previous studies have indicated that, on the
50 weather scale, the formation and maintenance of haze days in eastern China (HD_{EC}) are closely
51 related to favorable weather conditions (Niu et al., 2010; Cai et al., 2017), including strong thermal
52 inversion potential, high relative humidity, negative sea level pressure anomaly, and weak wind
53 speed. Furthermore, the anticyclonic anomaly could lead to the sinking movement and weaker
54 thermal inversion potential, which inhibit the vertical diffusion of pollutants and affect the air quality
55 of the local or larger region (Wu et al., 2013; Xu et al., 2015). Many studies investigated the key
56 circulation system affecting HD_{EC} from an interannual scale or intraseasonal scale and suggested
57 that the weak East Asian Winter Monsoon (Li et al., 2015; Yin et al., 2015; Zhang et al., 2022), the
58 positive phase of Arctic Oscillation (Wang et al., 2015; Yin et al., 2015) and the positive phase of
59 East Atlantic-West Russia (EA/WR) teleconnection pattern (Yin et al., 2017) could result in more

60 haze days in China. On the synoptic scale, meteorological conditions could also significantly
61 regulate HD_{EC}. The weak synoptic circulation with a high-pressure or continuous low-pressure
62 system is beneficial for the accumulation of pollution, while the strong weather phenomena with a
63 large pressure gradient encourage the diffusion of pollutants (Li et al., 2019; Cai et al., 2020).
64 Furthermore, studies have shown that cold surges can dissipate and reduce local air pollutants by
65 bringing dry and clean cold air (Wu et al., 2017; Leung et al., 2018; Zhang et al., 2021).

66 A recent study classified the daily winter circulation anomalies and suggested that there are two
67 dominant climate drivers (i.e., EA/WR teleconnection pattern and Victoria mode of sea surface
68 temperature anomalies) conducive to the severe haze occurrence in North China (Li et al., 2022).
69 Existing studies have also investigated the synoptic circulation patterns conducive to haze pollution
70 in different regions of China (Chang and Zhang, 2017; Li et al., 2019; Liu et al., 2019; Liao et al.,
71 2020; Sun et al., 2020; Yang et al., 2021; Gong et al., 2022). Most of these studies produced the
72 classification based on low-level circulation anomalies, while the upper-level circulation also play
73 an important role in the generation and accumulation of haze (Wang et al., 2015; Yin and Wang,
74 2017; Zhong et al., 2019). In addition, due to the large spatial span in EC, if we assess the
75 classification of synoptic circulation patterns in a fixed region, it may lead to different effects of the
76 same classification pattern on different regions. Therefore, we classify the circulation anomalies
77 with severe haze days of each station in EC respectively, and finally obtain the dominant synoptic
78 atmospheric circulation pattern of each station. In general, the present study addresses the following
79 scientific questions: (1) what are the synoptic atmospheric circulation patterns that dominate severe
80 haze pollution in EC, (2) what are the differences in the action range of each circulation pattern, and
81 what are their possible mechanisms. These issues are addressed using a modified classification
82 algorithm (Hierarchical Clustering Algorithm) that is more suitable for studying the classification
83 of synoptic patterns in a large spatial range.

84 The remaining sections of this paper are structured as follows: Data and definitions are
85 introduced in section 2. Section 3 shows the dominant synoptic circulation patterns of severe HD_{EC}.
86 In section 4, we compare different circulation types associated with severe HD_{EC}. Finally, the
87 discussion and main conclusions are given in section 5.

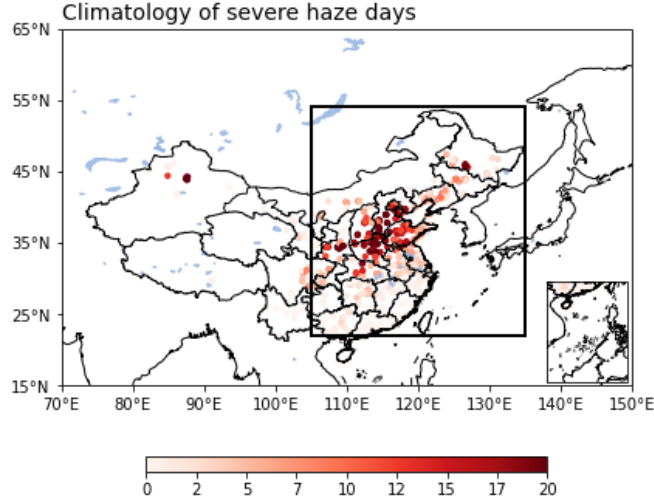
88 2. Data and Methods

89 2.1 Data

90 In this study, the daily meteorological data and the observed PM_{2.5} concentrations from 2014 to
91 2021 were used to analyze the dominant circulation patterns and their main causes of severe haze
92 in winter in EC. The daily NCEP/NCAR reanalysis was obtained from <https://psl.noaa.gov/>, which
93 includes sea level pressure (SLP), surface air temperature (SAT), the temperature in multiple
94 pressure levels, geopotential height (GPH), three-dimensional wind, relative humidity (RH) at 1000-
95 hPa, and vertical velocity (omega) at 850-hPa (Kalnay et al., 1996). The dataset has a horizontal
96 resolution of 2.5°× 2.5°. In this study, we defined the thermal inversion potential (TIP) as the air
97 temperature at 850-Pa minus SAT referring to Yin and Wang (2019). The Daily PM_{2.5} concentrations
98 for 935 meteorological stations in China (following Yin and Wang (2016) and Yin et al. (2021), the
99 stations with missing data more than 5% of are dropped; the stations with data lost continuously for
100 3 days or more is also discarded) were obtained from China National Environmental Monitoring
101 Centre (<https://quotsoft.net/air/>). The sporadic missing data (less than 3 days) were filled by cubic
102 spline interpolation.

103 2.2 Definition of severe HD_{EC}

104 In this study, the severe HD_{EC} is defined when PM_{2.5} concentration $\geq 150 \mu\text{g m}^{-3}$ (Cai et al., 2017;
105 Zhong et al., 2019). We focused on the haze days in the cool season (November to February of the
106 following year, abbreviated as NDJF), which accounts for more than 40% of the total haze days in
107 China in a year (Sun et al., 2013; Wang et al., 2015). Figure 1 shows the climatology of haze days
108 in China from 2014 to 2021 in NDJF. The severe haze days are mainly concentrated in the EC (east
109 of 105°E and south of 54 °N), which is selected as the target area in the present study. Thus, a subset
110 of 853 stations is selected.



111

112 **Figure 1.** Spatial distribution of the annual averaged severe haze days (unit: day) in China from
 113 2014 to 2021 in NDJF. The black box represents EC.

114 **2.3 Definition of blocking index**

115 In winter, the anticyclone anomaly over the Okhotsk Sea, usually related to atmospheric
 116 blocking, may lead to haze accumulation (Yun and Yoo 2019; Hwang et al., 2022). Thus, based on
 117 previous studies (Tibaldi et al., 1990; Fang and Lu, 2020), here we identify the blockings by
 118 northward gradients (GHGN) and southward gradients (GHGS) of Z_{500} at each grid point:

119
$$GHGN = \frac{z_{500}(\lambda, \phi + \Delta\phi) - z_{500}(\lambda, \phi)}{\Delta\phi} \quad (1)$$

120
$$GHGS = \frac{z_{500}(\lambda, \phi) - z_{500}(\lambda, \phi - \Delta\phi)}{\Delta\phi} \quad (2)$$

121 Where $\phi = 35^\circ, 35.5^\circ \dots, 75^\circ\text{N}$, $\lambda = 70^\circ, 70.5^\circ \dots, 160^\circ\text{E}$ and $\Delta\phi = 15^\circ$. A given longitude is
 122 defined as “blocked” at a particular time satisfies the following conditions:

123
$$GHGS > 0, \quad GHGN < -10 \text{ m (deg lat)}^{-1}$$

124 Based on these conditions, we can identify whether any grid in the range of 35°N - 70°N is blocked
 125 at any time.

126 **2.4 Plumb's wave activity flux**

127 Here we used the wave flux of Rossby to show the propagation of wave energy (Plumb, 1985).
 128 The two-dimensional Plumb's wave activity flux can be expressed by:

129
$$F_s = \frac{P}{P_0} \cos \varphi \times \left(\begin{array}{l} v'^2 - \frac{1}{2\Omega a \sin 2\varphi} \frac{\partial(v'\phi')}{\partial\lambda} \\ -u'v' + \frac{1}{2\Omega a \sin 2\varphi} \frac{\partial(u'\phi')}{\partial\lambda} \end{array} \right) \quad (3)$$

130 In Eq. (3), F_s (unit: $\text{m}^{-2} \text{s}^{-2}$) denotes the horizontal stationary wave activity flux, P means the
 131 pressure; $P_0 = 1000\text{-hPa}$, u' and v' are the zonal and meridional wind deviation, respectively.
 132 And the ϕ' is geopotential height. φ (λ) represents the latitude (longitude). a is the radius of Earth,
 133 and Ω means Earth's rotation rate.

134

135 2.5 Classification algorithm of synoptic atmospheric circulation

136 This paper uses the hierarchical clustering algorithm (HCA) to classify the severe HD_{EC} based
 137 on the associated circulations anomalies. Based on HCA (Rokach et al., 2005), we could create a
 138 clustering tree of data samples by calculating the Euclidean distance between different categories.
 139 The original data samples of different types are at the lowest level of the tree, and the root point of
 140 a cluster is at the top level of the tree.

141 Unlike Li et al. (2022), we only cluster the circulation anomalies of days with severe HD_{EC} ,
 142 which can ensure that all classification samples lead to $\text{PM}_{2.5}$ at least one station in EC exceeds the
 143 standard of severe haze pollution and produce more accurate classification types. Secondly, the
 144 circulation samples selected are not in a fixed region, but the rectangular regions of the same size
 145 centered on each station with severe haze. Since the upper-level circulation represented by 500-hPa
 146 geopotential height anomalies play an important role in the generation and accumulation of haze
 147 (Wang et al., 2015; Yin and Wang, 2017; Zhong et al., 2019), the GPH anomalies at 500-hPa in a
 148 rectangular region of 30 degrees from east, west, north, and south with each station as the center on
 149 the day of severe HD_{EC} were taken as the samples to perform HCA. It means that our classification
 150 results focus on the local circulation anomalies accompanied by haze, which can help us more
 151 accurately understand the impact of different local circulation patterns on different stations.
 152 Specifically, this clustering scheme can ensure that each station is located in the center of the
 153 circulation pattern when severe haze occurs, and avoid the impact of circulation pattern movement.
 154 The final composite results of the same pattern can reflect the average statement of the current type
 155 of circulation anomaly, which is helpful to investigate its possible physical mechanism.

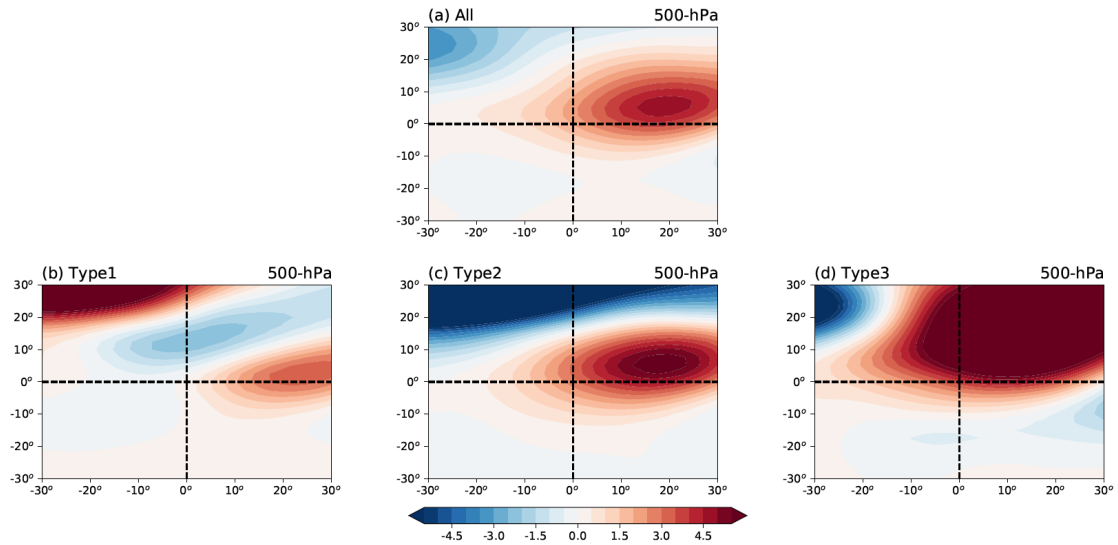
156 We use the silhouette coefficient to determine the optimal classification result (Rousseeuw,
157 1987). For any sample i , the silhouette coefficient $s(i)$ is defined as:

$$158 \quad s(i) = \frac{b(i)-a(i)}{\max\{a(i),b(i)\}} \quad (4)$$

159 $a(i)$ means the average distance from sample i to all other samples in the cluster it belongs to,
160 and $b(i)$ means the lowest average distance from sample i to all samples in any other cluster. The
161 silhouette coefficient of the clustering result is the average of the silhouette coefficients of all
162 samples. The closer to 1, the better the classification results. Figure S1 shows the clustering tree and
163 its associated silhouette coefficient of this study.

164 **3. Dominant synoptic atmospheric circulation patterns of severe HD_{EC}**

165 Figure 2a shows the composite anomalies of 500-hPa GPH during all severe HD_{EC} in 853
166 stations. Generally, the stations with severe haze are located in the southwestern parts of the
167 anticyclonic anomaly center, which is consistent with previous studies (Zhong et al., 2019; Wang
168 and Zhang, 2020). Then we performed the HCA as described in Section 2.5 and obtained three types
169 of dominant local circulation anomalies associated with the severe HD_{EC} (Figure 2b, c, d).
170 Circulation Type1 shows a wave-train structure of '+ - +', and the stations are located in the west
171 of anticyclonic anomaly and the south of cyclonic anomaly. Circulation Type2 shows the circulation
172 anomalies similar to Figure 2a. Finally, circulation Type3 denotes that the stations are located south
173 of the anticyclonic anomaly, and the intensity and range of the anticyclonic anomaly are
174 significantly stronger than the other two patterns. The differences between the types imply that
175 severe HD_{EC} may be related to different causes.

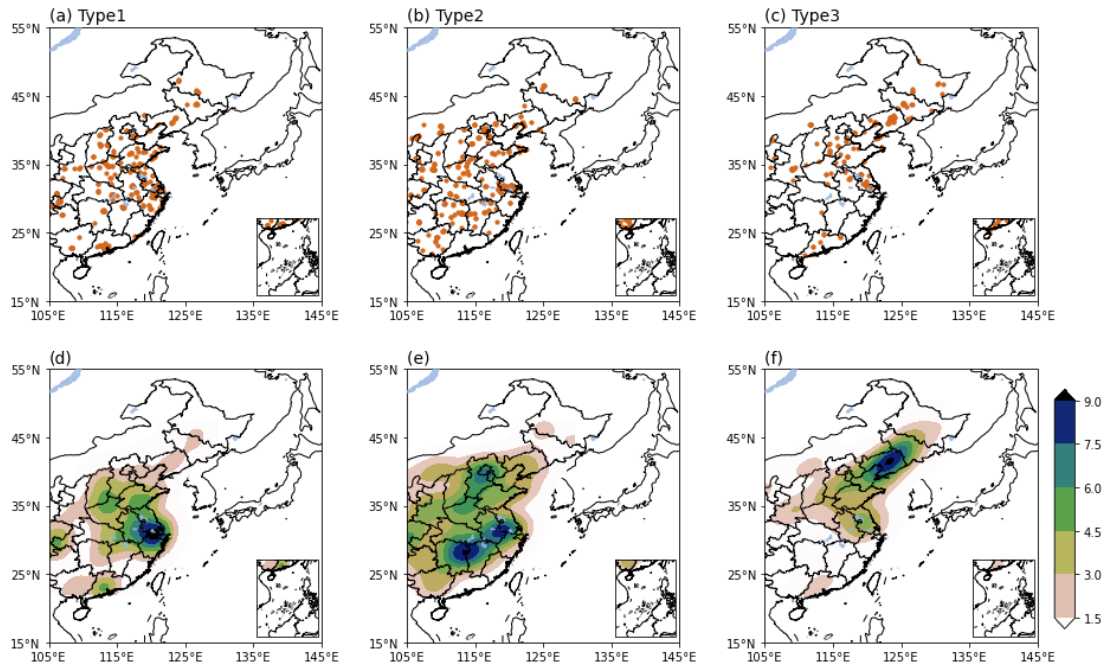


176

177 **Figure 2.** (a) Composite anomalies of GPH at 500-hPa (units: gpm) during all severe HD_{EC} in 853
 178 stations. (0°, 0°) represents the location of stations. (b), (c), and (d) are same as (a) but for three sub-
 179 types.

180

181 For each station, when the probability of a certain circulation type is greater than the sum of
 182 the other two types, we define this type as the dominant type of the station. Figure 3 shows the
 183 leading circulation types of severe HD_{EC} for 853 stations and the weighted probability density
 184 distribution of three circulation types (the weight of each station is the probability of the
 185 corresponded dominant type occurring at the station). Stations dominated by the circulation Type1
 186 are mainly distributed in the Yangtze River valley (YRV). The stations dominated by the circulation
 187 Type2 cover almost the whole EC, with two centers in South China (SC) and Beijing-Tianjin-Hebei
 188 region. The stations dominated by the circulation Type3 are mainly located in Northeast China
 189 (NEC). In general, the stations in the north of EC are accompanied by higher PM_{2.5} concentration
 190 and more haze days (Figure S2). These results suggest significant differences in the circulation
 191 patterns of severe haze in different regions of EC.

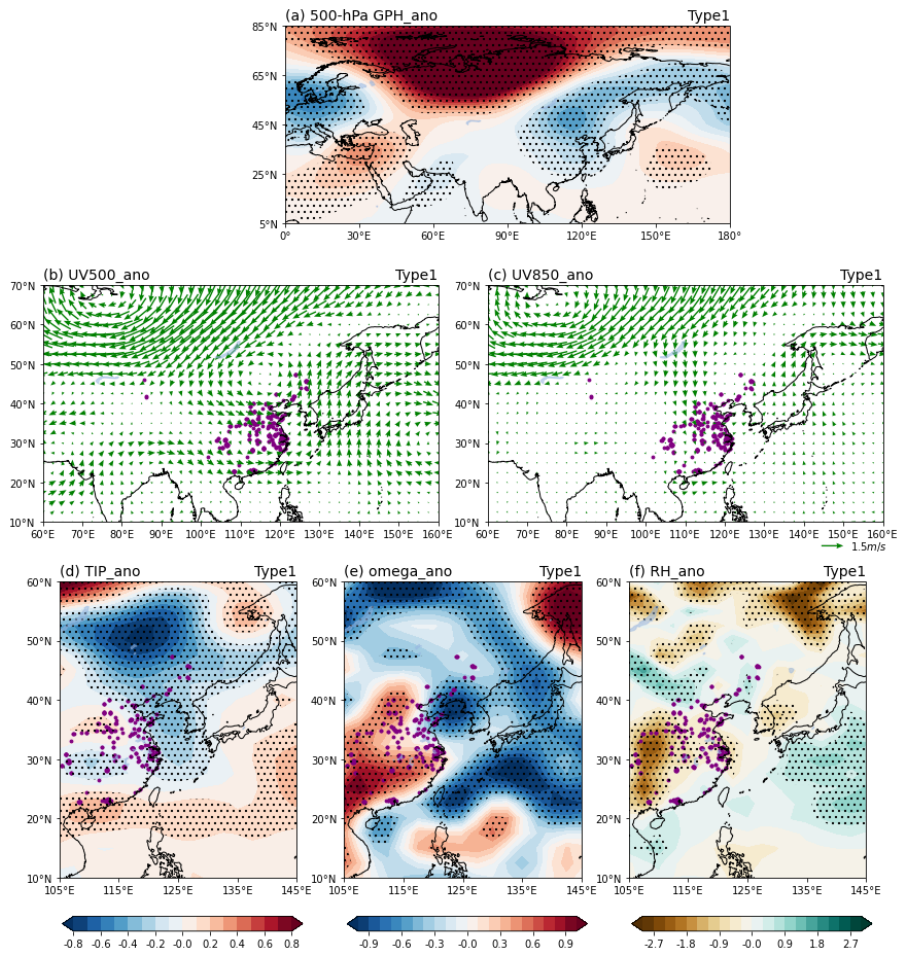


192

193 **Figure 3.** Distribution of stations dominated by (a) Type1, (b) Type2, and (c) Type3 synoptic
 194 circulation pattern. Weighted probability density distribution of stations dominated by (d) Type1, (e)
 195 Type2, and (f) Type3 synoptic circulation pattern.

196 **4. Comparison of different circulation types associated with severe HD_{EC}**

197 Figure 4a, b, and c show the composite anomalies of circulation Type1 at 500-hPa and 850-hPa.
 198 The circulation Type1 is associated with the upper troposphere's wave-train structure of “- + -”.
 199 Unlike previous studies (Zhong et al., 2019; Wang and Zhang, 2020), there are no significant
 200 anticyclonic anomalies in the mid-troposphere over YRV, but with substantial north wind
 201 component in the lower troposphere over northern China. The TIP, sinking movement, and RH
 202 anomalies over the YRV are weak (Figure 4d, e, f). Therefore, it can be inferred that it is not the
 203 local circulation anomalies that promote the formation and accumulation of haze pollution, but the
 204 regional haze transportation caused by the north wind component anomalies that leads to the severe
 205 haze in the YRV.



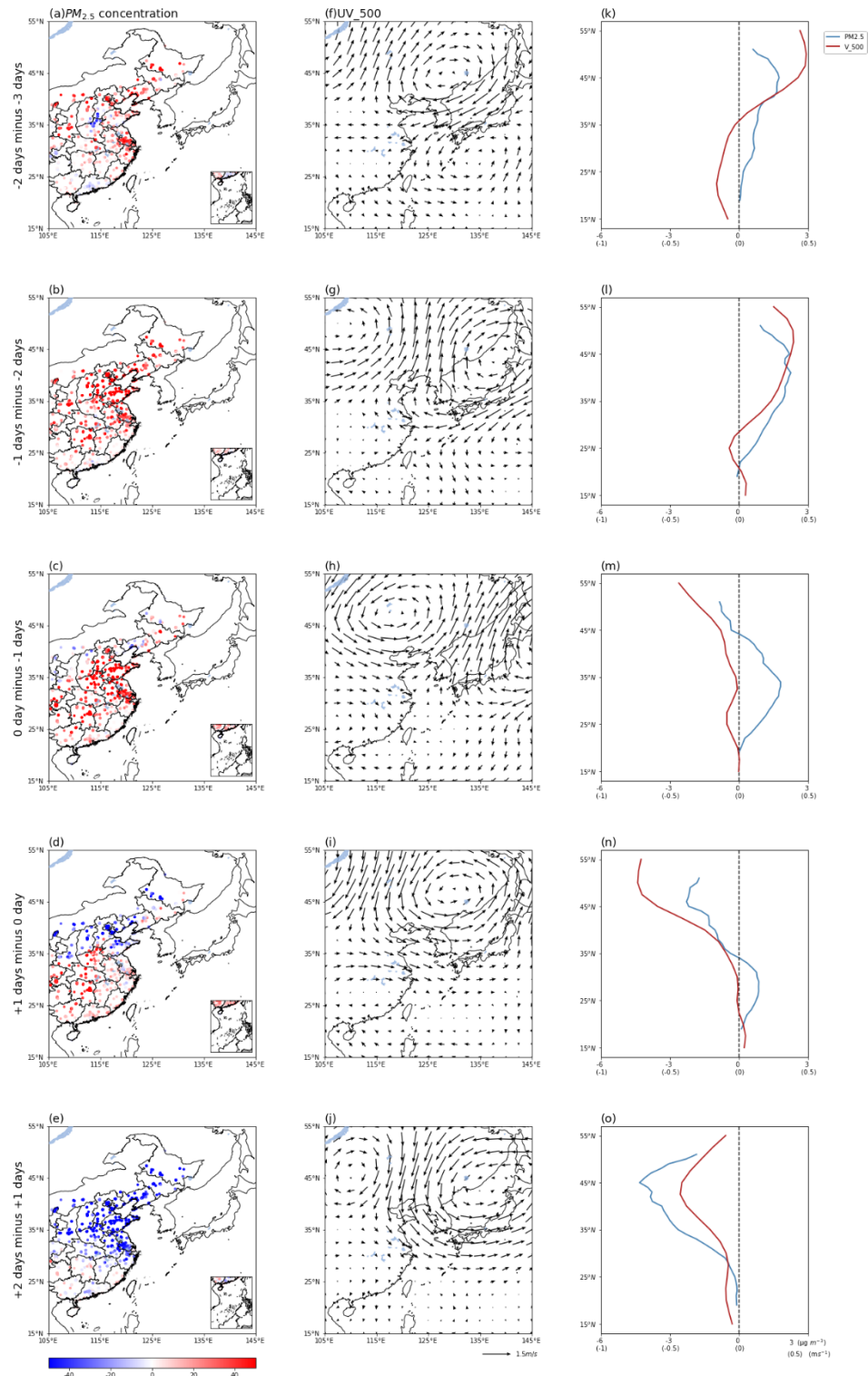
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207 **Figure 4.** Composite anomalies of (a) GPH at 500-hPa (unit: gpm), horizontal wind (unit: m s^{-1}) at
 208 (b) 500-hPa and (c) 850-hPa, (d) TIP (unit: K), (e) omega (unit: $10^{-2} \text{ Pa s}^{-1}$), and (f) RH (unit: %)
 209 for circulation Type1. Dotted areas are statistically significant at the 95% confidence level. The
 210 purple dots represent the stations dominated by circulation Type1.

211

212 To further explore the relationship between Type1 severe HD_{EC} and north wind component
 213 anomalies, we present the evolution of $\text{PM}_{2.5}$ concentration variations ($\text{PM}_{2.5}$ concentration on Day_i
 214 minus that on Day_{i-1}) from -3 days to 2 days of Type1 severe HD_{EC} occur (Figure 5a, b, c, d, e) and
 215 the corresponding horizontal wind variations at 500-hPa (Figure 5 f, g, h, i, j). $\text{PM}_{2.5}$ concentration
 216 tends to increase at first and then dissipate showing an obvious transportation process from north to
 217 south. Accordingly, the horizontal wind changes from anticyclonic anomalies to cyclonic anomalies,
 218 with the south wind turning to the north wind. Here we average the $\text{PM}_{2.5}$ concentration variations
 219 in Figure 5a, b, c, d, e, and meridional wind variations in Figure 5f, g, h, i, j along latitudes (Figure
 220 5k, l, m, n, o). The result shows that $\text{PM}_{2.5}$ concentration gradually increased from north EC to south

221 EC and began to decrease after severe HD_{EC} occurred. With the variation in PM_{2.5} concentration,
 222 the south wind in the north EC gradually weakens and turns to the north wind when severe HD_{EC}
 223 occurs. With the dry and cold air from the north invading southward, the haze dissipates rapidly,
 224 and EC can maintain high air quality weather. Therefore, although circulation Type1 will lead to
 225 severe haze in YRV, its circulation anomalies do not match the conditions to maintain haze pollution.

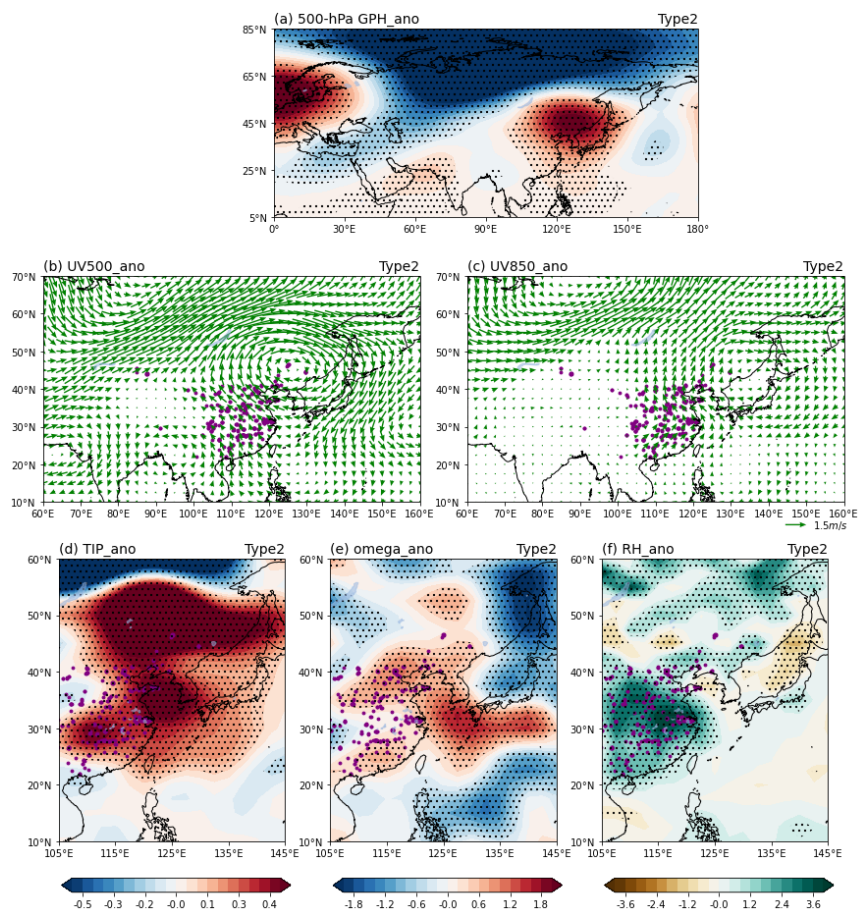


226

227 **Figure 5.** Composite anomalies of (a-e) the spatial distribution of PM_{2.5} concentration (unit: $\mu\text{g m}^{-3}$)
 228 ³) from -3 days to 2 days related to Type1 severe HD_{EC} occur and (f-j) the corresponding horizontal
 229 wind (unit: m s^{-1}) at 500-hPa. (k-o) shows the zonal averaged PM_{2.5} concentration variations (unit:
 230 $\mu\text{g m}^{-3}$) and meridional wind variations (unit: m s^{-1}) in the range of 15-55°N, 105-135°E.

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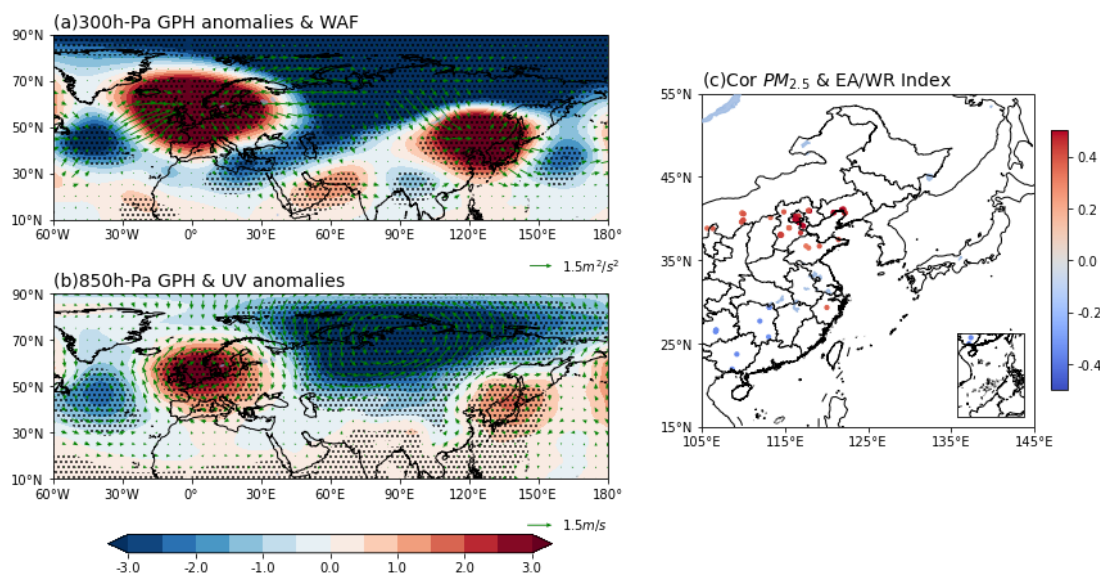
232 During the occurrence of circulation Type2, there was an anticyclonic anomaly with a quasi-
 233 barotropic structure over Northeast Asia, and the EC was located in the southwest of the anticyclone
 234 (Figure 6a, b, c). The significant positive TIP, sinking movement, and positive RH anomalies control
 235 the region over EC (Figure 6d, e, f). With the increase in TIP and the warm and humid air from the
 236 sea transports to the EC, the horizontal and vertical dispersion of pollutants was restrained, while
 237 higher surface RH exacerbated the formation of particulates. Such circulation anomalies are
 238 beneficial for the formation and maintenance of haze pollution.



239

240 **Figure 6.** Composite anomalies of (a) GPH at 500-hPa (unit: gpm), horizontal wind (unit: m s^{-1}) at
 241 (b) 500-hPa and (c) 850-hPa, (d) TIP (unit: K), (e) omega (unit: $10^{-2} \text{ Pa s}^{-1}$), and (f) RH (unit: %)
 242 for circulation Type2. Dotted areas are statistically significant at the 95% confidence level.

243 Here we investigate the dynamic mechanism of the circulation Type2 by compositing the GPH
 244 and WAF anomalies in the upper troposphere. The circulation anomalies show two quasi-zonal wave
 245 trains over the mid-high latitudes. The one is characterized by a ‘-+-+’ pattern of GPH anomalies
 246 from the south of Greenland across Siberia to Northeast China, with positive GPH anomalies in the
 247 second and fourth centers. Such anomalies are similar to the positive phase of EA/WR
 248 teleconnection, which can strengthen stable weather conditions over EC (Wu et al., 2016; Yin and
 249 Wang, 2016) by causing weak wind speed, higher RH, and strong TIP (Niu et al., 2010; Ding and
 250 Liu, 2014; Cai et al., 2017). Figure 7c shows the correlation coefficients between PM_{2.5}
 251 concentration during the occurrence of circulation Type2 and the EA/WR index (The EA/WR index
 252 was computed by the NOAA climate prediction center according to the rotated principal component
 253 analysis used by Barnston and Livezey (1987)). The results show significant positive correlations
 254 between the two in north EC and weak negative correlations in south EC. However, the circulation
 255 Type2 caused the severe HD_{EC} for almost the whole EC, which is not completely consistent with
 256 the results of Figure 7c. Therefore, we speculate that the other wave-train may lead to haze pollution
 257 in south EC.

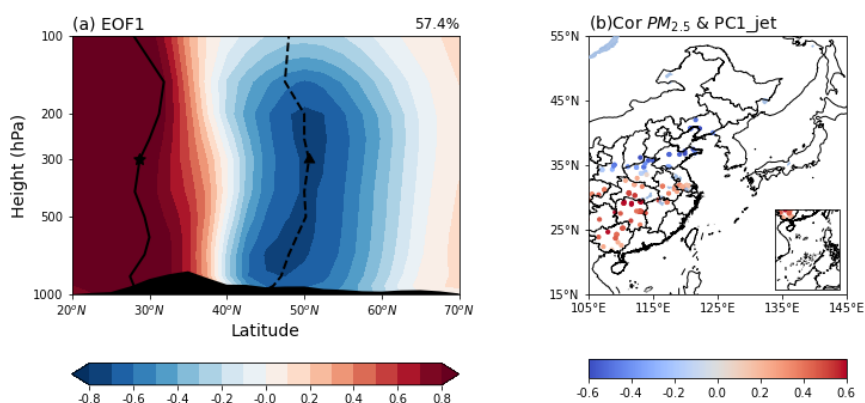


258
 259 **Figure 7.** Composite anomalies of (a) GPH (shading; gpm) and WAF (vectors; $m^2 s^{-2}$) at 300-hPa,
 260 and (b) GPH (shading; gpm) and horizontal wind (vectors; $m s^{-1}$) at 850-hPa for Type2. Dotted areas
 261 are statistically significant at the 95% confidence level. (c) Correlation coefficients between Type2
 262 PM_{2.5} concentration and EA/WR index.

263

264 It can be found that the second wave-train reaches EC from Europe along with southern Asia,
 265 forming a '+- + - +' pattern of GPH anomalies. The formation of such a wave-train is closely related
 266 to the winter East Asia subtropical jet (EASJ) (Xiao et al., 2016; An et al., 2020; Zhang et al., 2022).
 267 Here we use an Empirical orthogonal function (EOF) analysis of zonal wind from 1980 to 2021 to
 268 determine the leading modes of winter EASJ (Xiao et al., 2016). The variance of the first mode
 269 (EOF1) accounts for 57.4% of the total variance and indicates the intensity of EASJ (Figure 8a),
 270 which could significantly affect the haze pollution in EC (An et al., 2020; Zhang et al., 2022).

271 The correlation coefficients between daily $PM_{2.5}$ concentration and the first principle component
 272 (PC1_jet) during the occurrence of circulation Type2 is shown in Figure 8b, which has significant
 273 positive correlations in south EC and negative correlations in north EC. It indicates that the
 274 circulation Type2 may cause severe haze pollution in most areas of EC under the joint affection of
 275 EA/WR teleconnection and winter EASJ. The results suggested that when discussing the impact of
 276 an anticyclonic anomaly in Northeast Asia on haze pollution in EC, we should comprehensively
 277 consider the joint affection of signals from high and middle latitudes.

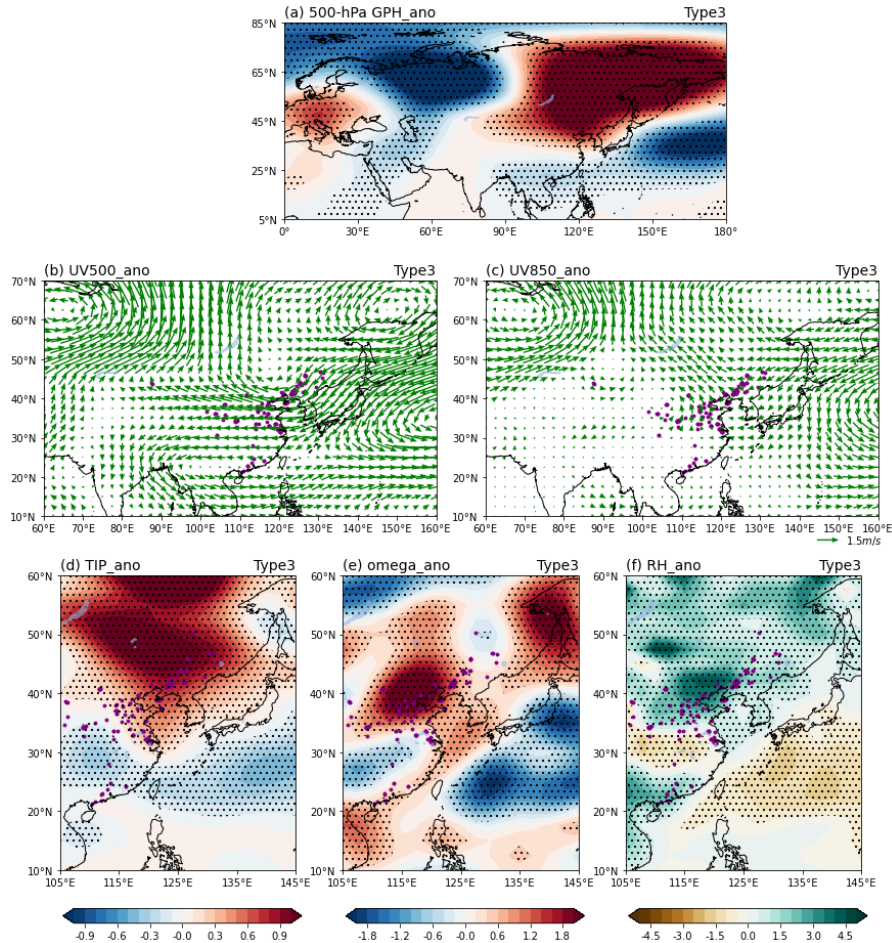


278
 279 **Figure 8.** (a) The first EOF mode of zonal wind (EOF1, $m s^{-1}$) averaged from 60°E to 160°E in
 280 NDJF. The star and circular at 300-hPa denote the subtropical jet and polar-front jet cores,
 281 respectively. The zonal mean orography is dark-shaded. (b) Correlation coefficients between Type2
 282 $PM_{2.5}$ concentration and PC1_jet.

283

284 Compared with circulation Type2, the range and intensity of anticyclonic anomalies in Northeast
 285 Asia circulation Type3 are more robust, and the location is more northerly (Figure 9a). Such
 286 circulation anomalies lead to southeasterly wind anomalies at 850-hPa, strong TIP, and abundant
 287 moisture that induces severe haze over NEC (Figure 9d, f). In addition, the ascending motion over

288 the south EC and the descending motion over the Beijing–Tianjin–Hebei region and NEC formed
 289 meridional circulation cell anomalies (Figure 9e), which are conducive to the accumulation of
 290 severe HD_{EC} over the NEC.

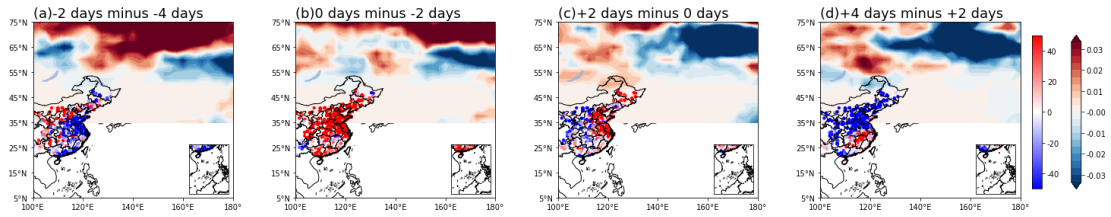


291
 292 **Figure 9.** Composite anomalies of (a) GPH at 500-hPa (unit: gpm), horizontal wind (unit: m s^{-1}) at
 293 (b) 500-hPa and (c) 850-hPa, (d) TIP (unit: K), (e) omega (unit: $10^{-2} \text{ Pa s}^{-1}$), and (f) RH (unit: %)
 294 for circulation Type3. Dotted areas are statistically significant at the 95% confidence level.

295

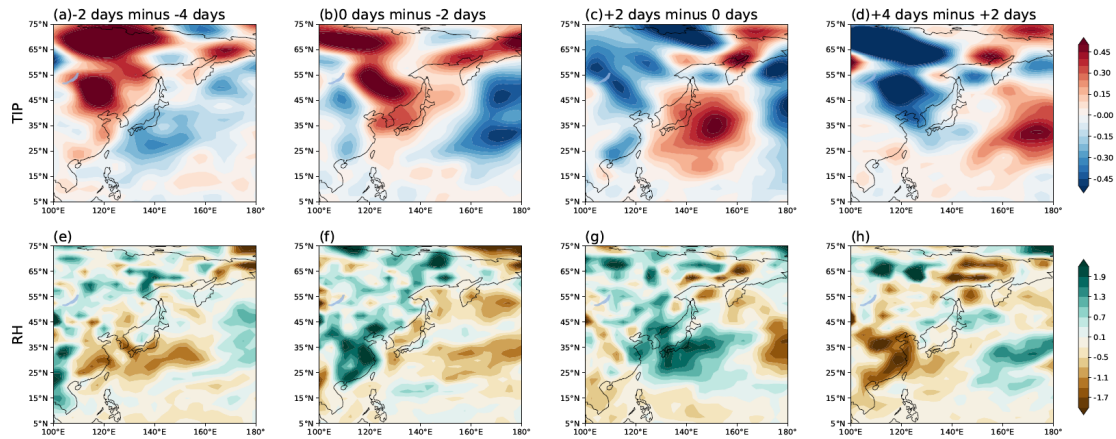
296 In winter, the anticyclonic anomalies over the Okhotsk Sea are usually related to atmospheric
 297 blocking (Yun and Yoo 2019; Fang et al., 2020; Hwang et al., 2022). Therefore, we calculated the
 298 daily atmospheric blocking introduced in section 2.3 to investigate the relationship between Type3
 299 severe HD_{EC}. Figure 10 shows that when Type3 severe HD_{EC} occurs, the PM_{2.5} concentration
 300 increases with the blocking anomalies in the high-latitudes build-up, dissipating with the blocking
 301 anomalies crash. The blocking anomalies strengthen the TIP and sufficient RH in the lower
 302 atmosphere (Figure 11), causing severe HD_{EC} in NEC.

303



304 **Figure 10.** Composite anomalies of (a-d) the spatial distribution of PM_{2.5} concentration variations
 305 (unit: $\mu\text{g m}^{-3}$) and blockings from -4 days to 4 days related to Type3 severe HDEC occur.

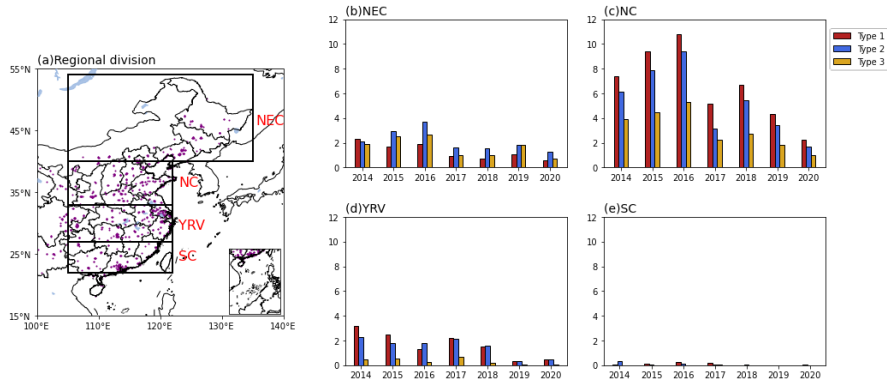
306



307 **Figure 11.** Composite anomalies of (a-d) TIP variations (unit: K) and (e-h) RH variations (unit: %)
 308 from -4 days to 4 days related to Type3 severe HDEC occur.

309

310 Based on the previous studies and the differences in the influence range of the three circulations
 311 types in this study, we divided the EC into NEC (40°N-54°N, 105°E -135°E), North China (NC;
 312 33°N-40°N, 105°E -122°E), the YRV (27°N-33°N, 105°E -122°E), and SC (22°N-27°N, 105°E -
 313 122°E) to analyze the temporal characteristics of three HDEC types in different subregions of EC
 314 (Figure 12a). Figure 12b, c, d, and e display the annual regional averaged frequency of the three
 315 HDEC types in the four subregions. The results show that severe haze pollution mainly occurs in NC
 316 and less in SC. The frequency of severe haze generally shows a downward trend in the four
 317 subregions.

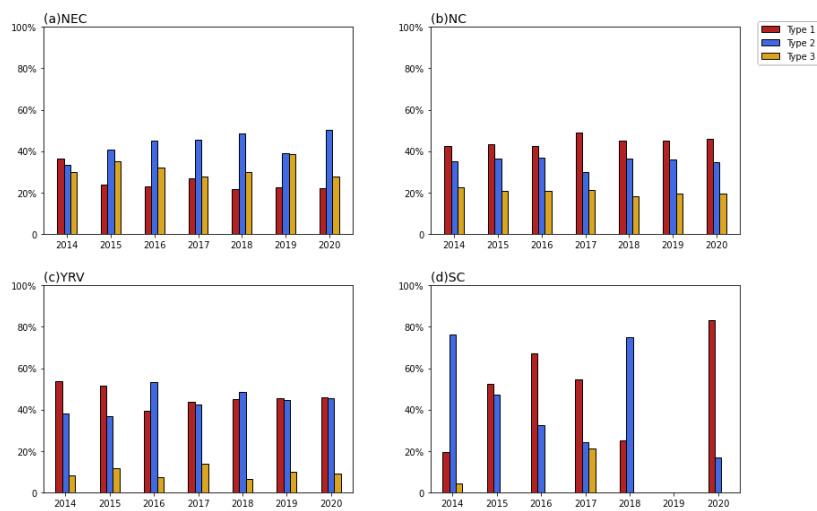


318

319 **Figure 12.** (a) The four subregions of EC. The purple dots are the stations. (b-e) Frequency of three
 320 types of cool season severe HD_{EC} in NEC, NC, YRV, and SC.

321

322 We further calculated the proportion of the frequency of each circulation type in the total annual
 323 severe haze frequency in the four subregions (Figure 13). For NEC, the proportion of the three
 324 circulation types is almost equal. It should be noted that the proportion of the circulation Type3 is
 325 much larger than that in the other three subregions. In NC, the proportion of the circulation Type1
 326 is more than 40%, while the proportion of the circulation Type3 is about 20%. For YRV, circulation
 327 Type1 and Type2 lead the severe haze pollution. There are relatively few severe haze pollution in
 328 SC. Therefore, the dominant circulation type in SC has strong interannual variation and is hardly
 329 affected by the circulation Type3. Overall, on the weather scale, the HD_{EC} is affected by a variety
 330 of synoptic circulations, and the areas affected by each synoptic circulation are also different.



331

332 **Figure 13.** Annual percentage of the three types of severe HD_{EC} in (a) NEC, (b) NC, (c) YRV, and
 333 (d) SC.

334 5. Conclusions and discussion

335 In this study, the Hierarchical Clustering Algorithm was used to investigate three dominant
336 circulation types that could lead to severe HD_{EC}. We cluster the circulations over the stations in EC
337 on the severe haze days from 2014 to 2021, which eliminates the interference of the circulations of
338 non-severe haze days on the cluster results. The results show that three dominant circulation types
339 associated with severe HD_{EC} are obtained, which are mainly characterized by a local anticyclonic
340 anomaly but also present obvious spatial variation on large scale circulations. The circulation Type1
341 with wave-train structure of “-+” in the upper troposphere mainly causes severe haze pollution in
342 the YRV through the low-level north wind anomalies over NC. Although the sinking movement,
343 TIP, and RH anomalies over the YRV are weak or not significant, the regional haze transportation
344 leads to the severe haze in the YRV. The circulation Type2 is characterized by two quasi-barotropic
345 Rossby wave trains at 300-hPa, which may be stimulated and sustained by the joint affection of
346 EA/WR teleconnection and the winter EASJ. One travels from the south of Greenland across Siberia
347 to NEC, forming a ‘-+-’ pattern of GPH anomalies, and the other travels from Europe along with
348 southern Asia, forming a ‘+--+’ pattern of GPH anomalies, which led an anticyclonic over
349 northeastern Asia and conducive to the accumulation of haze. The circulation Type3 is characterized
350 by blocking anomaly over Okhotsk Sea, which influences the severe HD_{EC} over NEC with
351 southeasterly wind at 850-hPa, strong TIP, and abundant moisture. The temporal characteristics of
352 three circulation types in NEC, NC, YRV, and SC were further analyzed. The result shows that on
353 the synoptic scale, HD_{EC} is affected by various synoptic atmospheric circulations, and the regions
354 affected by each synoptic atmospheric circulation are also different.

355 The study shows that circulation patterns and key systems that contribute to severe HD_{EC} are
356 complex and diverse revealing the dominant circulation patterns of severe haze in different regions
357 of EC. These three dominant atmospheric circulation patterns could be potentially used to establish
358 severe winter haze prediction models for different regions of EC (e.g., project the future variations
359 of severe haze in different regions of EC by identifying similar circulation patterns through machine
360 learning or regression fitting). Due to the limitation of data, it is difficult to carry out the work of
361 circulation classification over a longer period. Therefore, whether there is an interannual or
362 interdecadal connection between the dominant circulation types of severe haze and its key

363 circulation system needs further investigation. In addition, considering the latitude difference of
364 PM_{2.5} concentrations in EC and the decreasing of PM_{2.5} concentrations due to implementation of
365 the Air Pollution Prevention and Control Action Plan since 2013, the flexible threshold to identify
366 haze day is suggested to use in the further studies. And we will further carefully compare the impact
367 of emissions and meteorological factors on haze in subsequent work.

368 This study shows that different circulation types may lead to severe haze in different regions
369 of EC, and further studies are needed to investigate whether there are differences in persistence or
370 intensity among them.

371 **Data availability**

372 The Daily PM_{2.5} concentrations for 935 meteorological stations in China are collected by the
373 China National Environmental Monitoring Centre archive at: <https://quotsoft.net/air/> (last access:
374 16 May 2022). Daily mean meteorological data are obtained from the NCEP/NCAR reanalysis data
375 archive at: <https://psl.noaa.gov/data/gridded/data.ncep.reanalysis.pressure.html> (last access: 16
376 May 2022, NCEP/NCAR, 2022). The monthly EA/WR Index (CPC, 2022) can be downloaded from
377 NOAA's Climate Prediction Center: <http://www.cpc.ncep.noaa.gov/data/teledoc/telecontents.shtml>
378 (last access: 16 May 2022).

379 **Competing interests**

380 The authors declare that they have no conflict of interest.

381 **Author contributions**

382 SZ and GZ put forward the conception of this paper, TW improved the research and manuscript.
383 SZ, XY and IV performed research. SZ wrote the manuscript with contributions from all co-authors.

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