# Possible linkage between winter extreme low temperature over central-western China and autumn sea ice loss

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#### Abstract

Based on reanalysis datasets and sea-ice sensitivity experiments, this study has pointed out that the autumn sea ice loss in East Siberian-Chukchi-Beaufort (EsCB) Seas significantly increases the frequency of winter extreme low temperature over westerncentral China. Autumn sea ice loss warms the troposphere and generates anticyclonic anomaly over the Arctic region one month later. Under the effects of synoptic eddy-mean flow interaction and anomalous upward propagated planetary wave 2, the Arctic anticyclonic anomaly strengthens and develops toward Greenland-Northern Europe, accompanied by a weakened stratospheric polar vortex. In winter, following intra-seasonal downward propagation of stratospheric anomalies, the Northern European positive geopotential anomalies enhance and expand downstream within 7 days, favoring Arctic cold air east of Novaya Zemlya southward (hyperpolar path) accumulating in Siberia around Lake of Baikal. In the subsequent 2<sup>-3</sup> days, these cold anomalies rapidly intrude western-central China and induce abrupt sharp cooling, thus more frequent extreme low temperature there.

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19	Key Points:
20 21	Autumn EsCB sea ice loss favors cold air east of Novaya Zemlya invading western- central China, thus more frequent extreme low temperature
22 23	Synoptic eddy-mean flow interaction and anomalous upward planetary wave 2 provide favorable anticyclonic anomaly for extreme events outburst
24 25	Intra-seasonal downward propagated stratospheric anomalies are vital for the Ural anticyclonic anomaly to develop downstream within 7 days
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#### 30 Abstract

Based on reanalysis datasets and sea-ice sensitivity experiments, this study has pointed 31 out that the autumn sea ice loss in East Siberian-Chukchi-Beaufort (EsCB) Seas 32 significantly increases the frequency of winter extreme low temperature over western-33 central China. Autumn sea ice loss warms the troposphere and generates anticyclonic 34 anomaly over the Arctic region one month later. Under the effects of synoptic eddy-35 mean flow interaction and anomalous upward propagated planetary wave 2, the Arctic 36 37 anticyclonic anomaly strengthens and develops toward Greenland-Northern Europe, accompanied by a weakened stratospheric polar vortex. In winter, following intra-38 seasonal downward propagation of stratospheric anomalies, the Northern European 39 positive geopotential anomalies enhance and expand downstream within 7 days, 40 favoring Arctic cold air east of Novaya Zemlya southward (hyperpolar path) 41 accumulating in Siberia around Lake of Baikal. In the subsequent 2~3 days, these cold 42 anomalies rapidly intrude western-central China and induce abrupt sharp cooling, thus 43 more frequent extreme low temperature there. 44

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#### 46 Plain Language Summary

Arctic sea ice change not only regulates the local ecosystem but extends its influences 47 into mid- even low- latitudes through several complicated physical processes. Sea ice 48 variation in EsCB Seas exhibits an increased amplitude and more crucial role in climate 49 change under global warming. The new findings hinted that autumn EsCB sea ice 50 decrease would significantly promote western-central China to experience more 51 frequent winter extreme low temperature. In responses, an Arctic anticyclonic anomaly 52 53 occurs one month later and develops toward Greenland-Northern Europe due to synoptic eddy-mean flow interaction. Enhanced upward propagated planetary wave 2 54 and associated wave-mean flow interaction maintains the tropospheric Arctic 55 anomalies and weakens the stratospheric polar vortex. When entering winter, following 56 intra-seasonal downward propagated stratospheric anomalies, the Northern European 57 anticyclonic anomaly strengthens downstream within 7 days, favoring favors Arctic 58 cold air east of Novaya Zemlya rapidly invading western-central China (hyperpolar 59

path) and sudden sharp cooling. Our results have understood how autumn EsCB sea ice loss contributes to extreme low temperature in China, including possible physical mechanisms and cold air pathways, unlike previous work focusing on Barents-Kara Seas and winter-mean temperature change. It provides a new factor and theoretical foundations for predicting winter extreme low temperature in China.

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#### 66 1. Introduction

In recent decades, following the rapid Arctic warming and continuous Arctic sea 67 ice reduction, the Arctic-midlatitude association and possible mechanisms became one 68 69 of the focus in climate variability research (Cohen et al., 2020). Arctic sea ice change and its interaction with the atmosphere (Cohen et al., 2020; Deng and Dai, 2022; Wu et 70 al., 2022) not only impact the local thermal and dynamic states but also influence mid-71 72 and even low latitudes through complex interactions and feedback processes (Deser et al., 2004; Honda et al., 2009; Petoukhov and Semenov, 2010; Francis and Vavrus, 2012; 73 Vihma, 2014; Wu et al., 2017; Screen et al., 2018; Nakamura et al., 2019; Siew et al., 74 2020; Cohen et al., 2021). Barents-Kara (BK) Seas are the main region of interest in 75 many previous studies that showed autumn and winter sea ice loss favor increased 76 blocking, a strengthened Siberian high, and significant cooling over northern Eurasia 77 during winter (Francis et al., 2009; Honda et al., 2009; Wu et al., 2011; Mori et al., 2015, 78 79 2019; Cohen et al., 2021). The possible mechanisms include weakened high-latitudinal westerly winds due to the decreased meridional temperature gradient (Francis and 80 Vavrus, 2012, 2015) and horizontal or vertical propagation of quasi-stationary planetary 81 82 waves (Honda et al., 2009; Zhang et al., 2018a, b). In addition, enhanced meridional fluctuation of winter atmospheric circulation and stronger quasi-stationary planetary 83 84 waves are a result of reduced autumn Arctic sea ice. This leads to persistent weather patterns and increased frequency of extreme weather events (Francis and Vavrus, 2012, 85 2015; Tang et al., 2013; Wu et al., 2013, 2017; Overland et al., 2021; Zhang et al. 2022). 86

However, uncertainty remains regarding the Arctic-midlatitude association due to insignificant or weak atmospheric responses to sea ice reduction in large ensembles of numerical experiments (Barnes, 2013; Chen et al., 2016a; Sun et al., 2016; Blackport et al., 2019, 2020, 2021; Cohen et al., 2020). The climate effects of sea ice anomalies may be obscured by the chaotic nature of the atmosphere (Overland et al., 2021). Some modeling studies even concluded that winter Eurasian cooling or extreme cold events

are simply due to internal atmospheric variability because of very weak and 93 insignificant simulated atmospheric responses to Arctic sea ice forcing (McCusker et 94 al., 2016; Koenigk et al., 2019). The non-stationary Arctic-midlatitude association due 95 to global warming induces additional uncertainty. For instance, the climatological sea 96 ice northward shift, the impact of autumn Arctic sea ice loss on the winter Siberian high 97 is weakened (Chen et al., 2021). With continuous sea ice loss, the linkage between the 98 99 Arctic and Eurasia exhibits a strong low-frequency fluctuation of warm Arctic-cold Eurasia and warm Arctic-warm Eurasia (Wu et al., 2022). The high sensitivity of mid-100 101 high latitudinal atmospheric responses to the geographical location of Arctic sea ice anomalies and evident differences in climate effects between regional and entire Arctic 102 sea ice change also leads to diversity and uncertainty in previous numerical studies 103 (Chen et al., 2016b; Screen, 2017; Cohen et al., 2020). 104

Recent studies pointed out that autumn sea ice of the East Siberian-Chukchi-105 Beaufort (EsCB) Seas exhibits an increased interannual variability under global 106 warming, and the sea ice loss probably leads to colder northern Eurasia in the 107 108 subsequent early winter and early spring (Ding et al., 2021; Ding and Wu, 2021). A persistent Arctic anticyclonic anomaly, contributed by anomalous upward propagating 109 110 quasi-stationary planetary waves and the associated convergence anomaly in the troposphere and the stratosphere, partly explains the cross-seasonal impacts of autumn 111 regional sea ice. These studies discussed the Arctic-midlatitude association from the 112 perspective of seasonal means, similar as most previous works about BK sea ice change 113 (Wu et al., 2011; Zhang et al., 2018a, b; Cohen et al., 2021). 114

Although the winter mean Eurasian cooling associated with reduced EsCB sea ice 115 is weak and marginally significant (Figure 1a; Ding et al., 2021), especially for East 116 Asia/China there is a possibility of short term high impact events. The related 117 atmospheric anomalies are favorable for rapid and severe cold air outbreaks invading 118 East Asia/China and contributing to extreme low temperatures. Therefore, this study 119 will investigate whether the autumn (September-October) sea ice loss over the EsCB 120 Seas affects the frequency of winter extreme low temperature events over East 121 Asia/China and will explore the mechanisms of such cold air outbreaks. For this 122 purpose, we use statistical diagnosis of observations and sea-ice sensitivity model 123 experiments. 124

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#### 126 **2. Data and Methodology**

127 Atmospheric monthly and daily mean variables are taken from the NCEP-DOE 128 Reanalysis II with  $2.5^{\circ}$  longitude/latitude resolution, including air temperature, 129 geopotential height and horizontal winds (Kanamitsu et al., 2002). The Monthly mean 130 sea ice concentration (SIC) with a horizontal resolution of  $1.0^{\circ} \times 1.0^{\circ}$  comes from the 131 Met Office Hadley Center (Rayner et al., 2003). The study period covers 42 winters 132 from January 1979 to December 2021. All variables are linearly detrended. We obtain 133 similar conclusions using the raw data (not shown).

The definition of an extreme low temperature day is the daily-mean air 134 temperature (1000 hPa) lower than the 10th percentile of historical records or control 135 experiments, and the sum of extreme days during winter represents the frequency of 136 extreme low temperature. Western-central China experiencing more than one extreme 137 low temperature day is recorded as an extreme event. The interval between two events 138 should be longer than 15 days. There are 12 observed (Table S1) and 185 simulated 139 extreme low temperature events in low SIC years. We mainly focus on the regional sea 140 ice loss in the EsCB Seas (Figure S1a, b) because of its increasing variability (Figure 141 S1c, d), and the area-mean SIC anomalies  $(70.5^{\circ}N-82.5^{\circ}N, 135.5^{\circ}E-119.5^{\circ}W)$ , 142 multiplying by -1.0, are denoted as the EsCB index (Ding et al., 2021). Regression 143 analysis is the primary method to explore the observed association between the 144 frequency of extreme low temperature over central-western China and autumn EsCB 145 sea ice loss. Geopotential height tendency is utilized to portray the feedback of the 146 synoptic-scale eddy to the low-frequency flow (Lau and Holopainen, 1984; Lau, 1988; 147 Lau and Nath, 1991; Cai et al., 2007). Here, the eddy heat flux term associated with 148 baroclinic processes is much smaller (Lau and Holopainen, 1984; Lau and Nath, 1991), 149 150 so we only calculate the eddy vorticity flux term associated with barotropic processes is calculated as follows (Cai et al., 2007): 151

$$F = \nabla^{-2} \left[ -\frac{f}{g} \overline{\nabla \cdot \overrightarrow{V'} \zeta'} \right]$$

153 where  $\overrightarrow{V'}$  and  $\zeta'$  respectively denote the synoptic-scale horizontal winds and relative 154 vorticity, derived from a Butterworth band-pass filter with 2-10 day periods.  $\nabla^{-2}$  is 155 inverse Laplacian,  $\nabla$  is divergence, f is the Coriolis parameter and g is the acceleration

of gravity. We also employ the EP flux  $(F_y = -\rho a \cos \varphi \,\overline{u'v'}, F_z = \rho a \cos \varphi \,\frac{R_f}{HN^2} \,\overline{v'T'})$ 156 and its divergence  $(D_F = \frac{\nabla \cdot \vec{F}}{\rho a \cos \varphi})$  to depict the vertical propagation of quasi-stationary 157 158 planetary wave activity and the wave-mean flow interaction, respectively (Edmon et al., 1980; Plumb, 1985). Here,  $F_z$  ( $F_v$ ) is the vertical (meridional) component,  $\rho$  is the air 159 density, a is the radius of the earth,  $\varphi$  is the latitude, R is the gas constant, f is the 160 Coriolis parameter, H is the scale height, N is buoyancy frequency calculated from the 161 162 temperature data, *u* is zonal wind, *v* is meridional wind and *T* is temperature. The primes and overbars respectively denote zonal deviation and zonal average. The convergence 163 (divergence) of EP flux leads to decelerated (accelerated) westerly winds (Edmon et al., 164 1980; Chen et al., 2002, 2003). 165

166 The Specified Chemistry Whole Atmosphere Community Climate Model version 4.0 (SC-WACCM4; Smith et al., 2014) is employed to investigate the possible role of 167 EsCB sea ice loss for the extreme low temperature events over western-central China. 168 The horizontal resolution is  $1.9^{\circ}$  in latitude and  $2.5^{\circ}$  in longitude, and the vertical 169 resolution has 66 levels extending up to 0.0006 hPa. The control experiments are 170 performed with the climatological monthly SIC and SST averaged over 1982-2001 171 (model-derived data) and other constant external forcings (greenhouse gases, aerosols, 172 solar, etc.) at the year 2000 level. The sensitivity experiments are forced by decreased 173 EsCB sea ice from August to October, calculated from composite detrended Arctic SIC 174 with the detrended EsCB index greater than 0.85 (low sea ice: 1979, 1981, 1990, 2007, 175 2008, 2012; Figure S1b). We only consider three months of sea ice forcing because 176 significant signals with large anomalies mainly occur in August, September and 177 178 October (Figures S2a-c, h-j). Sea ice loss from November to February is very weak with scattered significant regions (Figure S2d-g, k-n). Both, sensitivity and control 179 experiments contain 100 members with different small perturbations added to the initial 180 condition, running from August to February of the next year after the model spins up. 181 The difference between the ensemble mean of the two experiments represents the 182 atmospheric model responses to the prescribed EsCB sea ice loss, whose significance 183 is examined by a two-tailed non-parametric Monte Carlo bootstrap significance test 184





Figure 1 Regressed winter (a) 1000hPa air temperature (shading, unit: °C), (c) extreme low temperature frequency, and (e) 1000 (contour, unit: m) and 500 hPa (shading, unit: m) geopotential height anomalies on normalized autumn (September-October; SO) detrended EsCB index during 1979/80~2020/21. (b), (d), (f) same as (a), (c), (e), but for simulated results. Black dots indicate 90% confidence level. Black rectangles in (c) and (d) represent the range of western-central China (28°N–40°N, 104°E–120°W).

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#### 197 *3.1 Association between winter extreme low temperature and autumn sea ice loss*

The possible contribution of the autumn EsCB sea ice loss to winter temperature 198 change anomalies over East Asia/China and associated Eurasian atmospheric 199 circulation anomalies are displayed in Figure 1, including the regression maps and 200 simulated differences. The autumn EsCB sea ice loss shows a statistically weak link to 201 202 the observed winter-mean cooling in most of China with limited significant range (Figure 1a). However, the reduced autumn EsCB sea ice statistically significantly 203 favors an increase in the frequency of extreme low temperature events over western-204 205 central China (black box in Figure 1c). Compared to the climatological mean (10.71), in six reduced EsCB sea ice years, the area-mean frequency of extreme low temperature 206 (13.83) over western-central China increases by 29.16% (p < 0.1). The sea-ice sensitivity 207 experiments confirm the above observed statistical association. In response to the 208 prescribed autumn EsCB sea ice loss, the model produces relatively weak winter-mean 209 cooling with limited significant range, but simultaneously exhibits a significantly 210 increased frequency of winter extreme low temperature events over western-central 211 212 China (Figures 1b, d). The simulated increased percentage (37.00%) for the frequency of winter extreme low temperature (12.33 versus 9) is close to the observations. Both, 213 214 simulations and observations consistently show that autumn EsCB sea ice decrease can promote more frequent winter extreme low temperature events in western-central China. 215 We hypothesize that atmospheric circulation anomalies associated with regional sea ice 216 melt rapidly invade western-central China in a "pulse-like" manner causing a sudden 217 drop in temperature for a limited time. 218

Figure 2 shows composite maps of the evolution of air temperature and 219 220 geopotential height anomalies at 1000hPa for observed and simulated extreme low temperature events in years with reduced SIC. Day 0 denotes the first day of extreme 221 222 low temperature occurring in western-central China. On the day  $-14 \sim -9$  (initial stage), cold anomalies already appear in northern Siberia (60°N-80°N), east of 60°E, together 223 with positive geopotential height anomalies covering the Arctic region with a 224 southward extension to the Greenland-Northern Europe sector (Figures 2a, g). During 225 the next 6 days (developing stage), positive geopotential height anomalies gradually 226 develop downstream and extend to Lake Baikal around 120°E, conducive to the 227 accumulation and strengthening of cold anomalies over central Siberia (40°N-70°N, 228 30°E-140°E; Figures 2b-d, h-j). Then, following the southward intrusion of the 229

Siberian anticyclonic anomaly, the significant cooling starts shifting southward on the 230 day  $-2 \sim -1$  and finally controls western-central China two days later (outbreak stage; 231 20°N-60°N, 60°E-140°E), replacing the previous warm anomalies and leading to a 232 sharp drop in temperature by 5~6 °C within 2~3 days (Figures 2e-f, k-l). In the middle 233 troposphere, positive geopotential height anomalies cover the Arctic Ocean with 234 southward extension to Northern Europe on the day  $-14 \sim -9$  (Figures S3a, e), gradually 235 strengthen toward the Ural Mountains within the subsequent 4 days (Figures S3b, f) 236 and finally develop downstream to Lake Baikal from day -4 to day 1 (Figure S3c-d, g-237 238 h). To its southeast, significant negative geopotential height anomalies appear around Lake Baikal since day  $-8 \sim -5$  through southeastward horizontal propagation of quasi-239 stationary planetary waves and shift toward the coast of East Asia about 6 days later. 240 This middle tropospheric atmospheric configuration contributes to the Ural blocking 241 high and strengthens the East Asian Trough. This provides favorable conditions for 242 Arctic cold air invading western-central China. The pathway of the above cold 243 anomalies mainly originates east of Novaya Zemlya southward to China (named the 244 hyperpolar path in Bueh et al., 2022). To elucidate the main source and path of extreme 245 low temperature, we further plot the count of cold anomalies  $< -5^{\circ}$ C occurring in each 246 247 Eurasian grid during the evolution of all extreme events based on the evolution characteristic of cold anomalies, including observed and simulated results (Figure 3a, 248 b). Consistent with the former composite maps, when the autumn EsCB sea ice is 249 reduced, the pathways of Arctic cold air inducing western-central China extreme low 250 temperature events are dominated by the hyperpolar path (high frequency in deep red), 251 with approximately 90% (10%) of the pathways deriving from the marginal seas to the 252 253 east (west) of Novaya Zemlya (black dots in Figures 3a, b). Note, that winter air temperature in the climatological mean east of Novaya Zemlya is much colder, which 254 255 explains that the hyperpolar path generally results in strong cooling in western-central China. Consequently, the downstream development of the anticyclonic anomaly over 256 the Arctic Ocean-Northern Europe region is crucial for the rapid southward intrusion of 257 Arctic cold air east of Novaya Zemlya into western-central China. 258 259



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Figure 2 Composite evolution of detrended air temperature anomalies (shading, 261 unit: °C) and detrended geopotential height anomalies (green contour, unit: m) at 1000 262 hPa in (a) day  $-14 \sim -9$ , (b) day  $-8 \sim -7$ , (c) day  $-6 \sim -5$ , (d) day  $-4 \sim -3$ , (e) day  $-2 \sim -1$  and 263 (f) day  $0 \sim 1$  for 12 extreme low temperature events over central-western China during 264 six low EsCB sea ice years. Day  $-14 \sim -9$ , with cold anomalies over northern Siberia, is 265 defined as the initial stage. Day  $-8 \sim -3$ , with cold anomalies over central Siberia, is 266 defined as the developing stage. Day  $-2 \sim 1$ , with cold anomalies over southern Siberia 267 and western-central China, is defined as the outbreak stage. (g) - (l) same as (a) - (f), 268 but for simulated results with 185 extreme low temperature events in 100 members. 269 Black dots and purple contours indicate 90% confidence level. The interval of contour 270 is 20 m. 271 272



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Figure 3 (a) Count (shading) of extreme cold anomalies (<-5 °C) occurring in each 274 Eurasian grid during the evolution of 12 observed extreme cold temperature events. We 275 divide the life cycle of extreme low temperature events into three stages covering the 276 distinct regions according to their evolution characteristics (Figure 2), including initial 277 (day-14 ~ -9; 60°N–80°N, 0°–140°E), developing (day -8 ~ -3; 40°N–70°N, 30°E– 278 140°E) and outbreak (day  $-2 \sim 1$ ; 20°N-60°N, 60°E-140°E) stages. If grids in the 279 selected region appear cold anomalies less than -5°C in either phase, we consider an 280 extreme low temperature event passing through these grids. Black, gray and blue dots 281 represent the mean location of cold anomalies  $< -5^{\circ}$ C in initial, developing, and 282 outbreak phases. The red shading and shift of dots represent the main pathway. Black 283 rectangles in (c) and (d) represent the range of western-central China. (b) same as (a), 284

but for simulated results with 185 extreme low temperature events. Regressed 500 hPa (c) late autumn (October-November-December) geopotential height anomalies (Shading, unit: m), (d) early winter (November-December-January) geopotential height anomalies (Shading, unit: m) and (e) early winter geopotential height tendency anomalies (Shading, unit: m day<sup>-1</sup>) on normalized SO detrended EsCB index. (f) - (h) same as (c) - (e), but for simulated results. Black dots in (c) - (h) indicate 90% confidence level.

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The above anticyclonic anomaly also emerges in the winter-mean anomalous 293 atmospheric circulation shown in Figures 1e, f. As indicated by observations and 294 simulations, the reduction of autumn EsCB sea ice makes the Arctic Ocean and 295 296 Greenland-Northern Europe more prone to positive geopotential height anomalies from the lower to middle troposphere in winter. In the lower troposphere, positive 297 298 geopotential height anomalies cover Northern Europe and expand eastward to East Asia. In the middle troposphere, a northwest-southeastward oriented dipole structure controls 299 300 Eurasia with positive (negative) geopotential height anomalies over the Greenlandnorthern Ural Mountains (Lake Baikal) (Figures 1e, f). Such anomalous atmospheric 301 configuration creates suitable background circulation (decelerated westerlies, reduced 302 meridional potential vorticity gradient and enhanced meridional fluctuation of 303 atmospheric circulation; Francis and Vavrus, 2015; Luo et al. 2018) for the intra-304 seasonal strengthening of the Siberian high (Figure 2), the Ural blocking high and the 305 East Asian Trough (Figure S2), thus leading Arctic cold air rapidly to western-central 306 China (hyperpolar path) and resulting in strong cooling within a few days. 307

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## 309 *3.2 Possible mechanisms: the role of synoptic eddies and planetary waves*

Both, observations and simulations indicate that the increased frequency of 310 extreme low temperature events over western-central China and the anticyclonic 311 anomaly over the Arctic Ocean-Northern Europe are linked to the previous autumn 312 EsCB sea ice loss. To elucidate the possible physical mechanisms, the evolution of 313 observed and simulated 500 hPa geopotential height anomalies from one month after 314 autumn EsCB sea ice minimum is examined. In late autumn (October-November-315 December), significant reduction of autumn regional sea ice increases the upward heat 316 flux from the ocean (Ding et al. 2021; Ding and Wu, 2021) warming the Arctic 317

atmosphere and raising the geopotential height of the entire troposphere. Positive 318 geopotential height anomalies control the Arctic region with southward extension in the 319 North Atlantic sector (Figures 3c, f), then continue to develop and intensify southward 320 in early winter (November-December-January) (Figures 3d, g). Its anomaly center 321 322 finally moves to Greenland-Northern Europe one month later (Figures 1e, f). Previous 323 works discovered that the local synoptic wave-mean flow interaction associated with the weakening (strengthening) of the North Atlantic storm track favors the generation 324 of anticyclonic (cyclonic) forcing to the north (Ding et al., 2017, 2019). Here, we further 325 326 calculated the geopotential height tendency. In early winter, observed and simulated positive geopotential height tendency anomalies dominate the Arctic region with 327 significant and large anomalous signals around Greenland-Northern Europe (Figures 328 3e, h). This indicates that the weakened North Atlantic storm track and associated 329 interaction between synoptic waves and low-frequency flow contributes to the 330 331 southward development and shift of the Arctic anticyclonic anomaly in the North Atlantic sector. In addition, the anomalous upward propagated quasi-stationary 332 planetary wave energy related to the reduced autumn EsCB sea ice also supports the 333 persistent Arctic positive geopotential height anomalies during mid-winter (Zhang et 334 335 al., 2018b; Ding et al., 2021). In early winter, consistent with previous works (Ding and Wu, 2021), using a different model also confirms that planetary wave 2 dominates the 336 increase in the upward propagation of quasi-stationary planetary wave energy, with two 337 anomalous upward EP flux regions in the mid-high latitudes (Figures S4a-c). One 338 upward branch propagates into the lower stratosphere and generates an EP flux 339 convergence anomaly north of 60°N, leading to the decelerated westerly winds and 340 weakened stratospheric polar vortex. The downward propagation of stratospheric 341 anomalies favors the maintenance of the winter Arctic anticyclone anomaly and may 342 provide a potential source for the downstream development of positive geopotential 343 height anomalies around Northern Europe on the intra-seasonal timescale. Another 344 upward branch converges poleward in the upper troposphere around 70°~80°N, directly 345 strengthening the Arctic anticyclonic anomaly by wave-mean flow interaction. 346

As indicated in Figure 2, positive geopotential height anomalies over Northern Europe developing downstream are the vital system for the extreme low temperature outbreaks over western-central China. Consequently, we discuss the possible source of this atmospheric anomaly precursor based on time-height cross sections of observed and simulated area-mean geopotential height anomalies over the Ural Mountains (60°-

90°N, 40-80°E). Positive geopotential height anomalies, indicating the weakened polar 352 vortex, control the lower-middle troposphere throughout the whole outburst of extreme 353 low temperature air masses. Around day  $-10 \sim -9$ , the stratospheric anomalies begin to 354 propagate downward and reach the lower-middle troposphere in the subsequent six days 355 near the Ural Mountains (Figures S5a, b). This favors positive geopotential height 356 anomalies in the troposphere expanding toward Lake Baikal. The timing of downward 357 propagation is highly consistent with the timing (day -8~-3) of the enhancement and 358 downstream development of anticyclonic anomaly around the Ural Mountains (Figures 359 360 S3b-c, f-g). Therefore, the intra-seasonal downward propagation of planetary wave energy contributes to the Northern European anticyclonic anomaly downstream 361 developing and supporting the Arctic cold air east of Novaya Zemlya to rapidly 362 southward intrude western-central China (hyperpolar path). 363

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## 365 *3.3 Different impacts in comparison to Barents-Kara Sea ice*

Autumn sea ice variability in BK Seas, as the second EOF mode (Ding et al. 2021), 366 367 is widely concerned and its climate role shows different feature compared with the EsCB sea ice loss. The sea ice loss exerts much stronger and more extensive impacts 368 369 on winter-mean climate change, inducing significant cooling in most regions of China (Figures S6a, b) with stronger Siberian High and deeper East Asian Trough (Figures 370 S6e, f; Wu et al., 2011). For the extreme low temperature, the significantly increased 371 frequency mainly occurs in northwestern, northeastern and eastern China rather than in 372 the western-central part (Figures S6c, d). 373

The Arctic-midlatitude association is non-stationary and varies with time under 374 global warming (Chen et al., 2021; Wu et al., 2022), and similar results are obtained in 375 our study (Figure S7a). The 19-year sliding correlation coefficients between the autumn 376 EsCB index and the winter frequency of extreme low temperature over western-central 377 China display significant positive correlations (r>0.4, p<0.05) since the late 1990s, with 378 the highest value exceeding 0.7, indicating that the climate importance of EsCB sea ice 379 change is increasingly evident. In contrast, the climate effects of BK sea ice anomalies 380 381 on northwestern China are significant before the late 1990s (r>0.4, p<0.05) and have turned almost insignificant in recent two decades. The rapidly reduced climatological 382 mean sea ice may modulate the above relationship (Chen et al., 2021). 383

In addition, the internal atmospheric variability can also interfere with the Arcticmidlatitude association. In the simulations, we further calculate the signal-to-noise ratio, defined as the relative contributions induced by sea ice forcing to the model internal variability. The increased frequency of winter extreme low temperature forced by EsCB (BK) sea ice loss is about 30~40% of the model internal variability over western-central (northeastern and eastern) China (Figure S7b, c). Therefore, the Arctic-midlatitudes linkage is complex and non-stationary even if only considering the sea ice forcing.



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Figure 4 Schematic diagram of the possible physical pathway for the reduced autumn 393 EsCB sea ice affecting the extreme low temperature events over western-central China. 394 AC and PV respectively indicate anticyclonic anomalies and polar vortex. Shadings in 395 the polar projection map, pressure-latitude map and latitude-longitude map represent 396 397 sea ice anomalies, EP flux divergence anomalies and 1000hPa air temperature anomalies, respectively. Contours in all maps represent the geopotential height 398 anomalies. Purple (green) contours in the latitude-longitude map indicate the location 399 of Eurasian AC when (10 days before) the extreme cold outburst. Dots represent the 400 increased frequency of extreme low temperature. Blue dashed line arrow represents the 401 cold air outbreak path. Blue solid line arrow represents intra-seasonal downward 402 propagation of stratospheric anomalies. Vectors represent the EP flux. Curved arrow 403 represents the synoptic eddy - mean flow interaction and "+" with counterclockwise 404 arrow indicates positive geopotential height tendency anomalies. 405

#### 407 **4. Summary and Discussion**

By analyzing the reanalysis datasets and performing sea-ice sensitivity 408 experiments, this study emphasizes the significant impact of autumn EsCB sea ice loss 409 on climate change over western-central China. It is mainly reflected in the significantly 410 increased frequency of winter extreme low temperature events rather than by winter-411 412 mean cooling. The specific physical processes are summarized in Figure 4. When the previous autumn EsCB sea ice is reduced, enhanced heat flux upward from the open 413 water results in significant local warming and elevation of geopotential height levels 414 415 controlling the Arctic troposphere one month later. Then, under the influence of two possible mechanisms, the positive geopotential potential height anomalies persist into 416 winter. One possible mechanism is the local synoptic eddy-mean flow interaction 417 associated with the weakened North Atlantic storm track that facilitates the Arctic 418 anticyclonic anomaly developing southward with the anomalous center shifting to 419 420 Greenland-Northern Europe. The other possible mechanism is associated with the anomalous upward propagation of quasi-stationary planetary waves and its generated 421 422 wave-mean flow interaction (EP flux convergence anomaly), dominated by planetary wave 2. The tropospheric branch directly strengthens the Arctic positive geopotential 423 424 height anomalies in early winter. The stratospheric branch attenuates the winter polar vortex, favoring the intra-seasonal downward propagation of stratospheric anomalies 425 and contributing to the persistent Arctic anticyclonic anomaly throughout the entire 426 troposphere. Therefore, Greenland-Northern Europe is generally dominated by positive 427 geopotential anomalies in the wintertime. When the stratospheric anomalies propagate 428 429 downward on the intra-seasonal timescale, the Northern European positive geopotential anomalies strengthen and develop downstream within 7 days, which favors severe 430 Arctic cold air east of Novaya Zemlya shifting southward (hyperpolar path) and 431 432 accumulating in Siberia around Lake of Baikal. In the subsequent 2~3 days, these cold anomalies rapidly invade western-central China, bringing a sudden sharp drop in 433 temperature and explaining the increased frequency of extreme low temperature there. 434 435 The climate role of autumn EsCB sea ice change in western-central China has become increasingly significant in recent two decades. In contrast, autumn sea ice loss in BK 436 Seas favors more frequent extreme low temperature events over northeastern and 437 eastern China, whose dominant effects occur before the late 1990s. 438

Although the simulations capture the main features of the observations, certain
 simulated deviations still exist. For example, the location of the simulated winter-mean

Arctic anticyclonic anomaly is farther north than the observed. On intra-seasonal 441 timescale, compared to observations, the simulated Siberian cold anomalies from day -442 14 to day -5 shift eastward and are more significant, accompanied by stronger positive 443 geopotential height anomalies over northern Eurasian. These simulated deviations may 444 be related to the inaccurate model descriptions and lack of other external forcings such 445 as Eurasian snow cover and sea surface temperature in Pacific or Atlantic Ocean. 446 Besides downward propagating stratospheric signals, other factors such as the intra-447 seasonal oscillation may also trigger the downstream development of Northern 448 449 European anticyclonic anomaly. These phenomena are out of scope of our discussion. In addition, favorable initial atmospheric conditions or internal atmospheric variability 450 for autumn EsCB sea ice affecting extreme low temperature in China needs further 451 452 investigation.

453

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460

### 461 **Open Research**

The Monthly mean sea ice concentration (SIC) from the Met Office Hadley Center are 462 available via https://www.metoffice.gov.uk/hadobs/hadisst/data/download.html. The 463 NCEP-NCAR (National Center for Environmental Prediction) - DOE (Department of 464 Reanalysis Π dataset is available 465 Energy) at https://psl.noaa.gov/data/gridded/data.ncep.reanalysis2.html. 466

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1	Possible linkage between winter extreme low temperature over
2	central-western China and autumn sea ice loss
3	
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18	
19	Key Points:
20 21	Autumn EsCB sea ice loss favors cold air east of Novaya Zemlya invading western- central China, thus more frequent extreme low temperature
22 23	Synoptic eddy-mean flow interaction and anomalous upward planetary wave 2 provide favorable anticyclonic anomaly for extreme events outburst
24 25	Intra-seasonal downward propagated stratospheric anomalies are vital for the Ural anticyclonic anomaly to develop downstream within 7 days
26 27 28	
29	

#### 30 Abstract

Based on reanalysis datasets and sea-ice sensitivity experiments, this study has pointed 31 out that the autumn sea ice loss in East Siberian-Chukchi-Beaufort (EsCB) Seas 32 significantly increases the frequency of winter extreme low temperature over western-33 central China. Autumn sea ice loss warms the troposphere and generates anticyclonic 34 anomaly over the Arctic region one month later. Under the effects of synoptic eddy-35 mean flow interaction and anomalous upward propagated planetary wave 2, the Arctic 36 37 anticyclonic anomaly strengthens and develops toward Greenland-Northern Europe, accompanied by a weakened stratospheric polar vortex. In winter, following intra-38 seasonal downward propagation of stratospheric anomalies, the Northern European 39 positive geopotential anomalies enhance and expand downstream within 7 days, 40 favoring Arctic cold air east of Novaya Zemlya southward (hyperpolar path) 41 accumulating in Siberia around Lake of Baikal. In the subsequent 2~3 days, these cold 42 anomalies rapidly intrude western-central China and induce abrupt sharp cooling, thus 43 more frequent extreme low temperature there. 44

45

#### 46 Plain Language Summary

Arctic sea ice change not only regulates the local ecosystem but extends its influences 47 into mid- even low- latitudes through several complicated physical processes. Sea ice 48 variation in EsCB Seas exhibits an increased amplitude and more crucial role in climate 49 change under global warming. The new findings hinted that autumn EsCB sea ice 50 decrease would significantly promote western-central China to experience more 51 frequent winter extreme low temperature. In responses, an Arctic anticyclonic anomaly 52 53 occurs one month later and develops toward Greenland-Northern Europe due to synoptic eddy-mean flow interaction. Enhanced upward propagated planetary wave 2 54 and associated wave-mean flow interaction maintains the tropospheric Arctic 55 anomalies and weakens the stratospheric polar vortex. When entering winter, following 56 intra-seasonal downward propagated stratospheric anomalies, the Northern European 57 anticyclonic anomaly strengthens downstream within 7 days, favoring favors Arctic 58 cold air east of Novaya Zemlya rapidly invading western-central China (hyperpolar 59

path) and sudden sharp cooling. Our results have understood how autumn EsCB sea ice loss contributes to extreme low temperature in China, including possible physical mechanisms and cold air pathways, unlike previous work focusing on Barents-Kara Seas and winter-mean temperature change. It provides a new factor and theoretical foundations for predicting winter extreme low temperature in China.

65

#### 66 1. Introduction

In recent decades, following the rapid Arctic warming and continuous Arctic sea 67 ice reduction, the Arctic-midlatitude association and possible mechanisms became one 68 69 of the focus in climate variability research (Cohen et al., 2020). Arctic sea ice change and its interaction with the atmosphere (Cohen et al., 2020; Deng and Dai, 2022; Wu et 70 al., 2022) not only impact the local thermal and dynamic states but also influence mid-71 72 and even low latitudes through complex interactions and feedback processes (Deser et al., 2004; Honda et al., 2009; Petoukhov and Semenov, 2010; Francis and Vavrus, 2012; 73 Vihma, 2014; Wu et al., 2017; Screen et al., 2018; Nakamura et al., 2019; Siew et al., 74 2020; Cohen et al., 2021). Barents-Kara (BK) Seas are the main region of interest in 75 many previous studies that showed autumn and winter sea ice loss favor increased 76 blocking, a strengthened Siberian high, and significant cooling over northern Eurasia 77 during winter (Francis et al., 2009; Honda et al., 2009; Wu et al., 2011; Mori et al., 2015, 78 79 2019; Cohen et al., 2021). The possible mechanisms include weakened high-latitudinal westerly winds due to the decreased meridional temperature gradient (Francis and 80 Vavrus, 2012, 2015) and horizontal or vertical propagation of quasi-stationary planetary 81 82 waves (Honda et al., 2009; Zhang et al., 2018a, b). In addition, enhanced meridional fluctuation of winter atmospheric circulation and stronger quasi-stationary planetary 83 84 waves are a result of reduced autumn Arctic sea ice. This leads to persistent weather patterns and increased frequency of extreme weather events (Francis and Vavrus, 2012, 85 2015; Tang et al., 2013; Wu et al., 2013, 2017; Overland et al., 2021; Zhang et al. 2022). 86

However, uncertainty remains regarding the Arctic-midlatitude association due to insignificant or weak atmospheric responses to sea ice reduction in large ensembles of numerical experiments (Barnes, 2013; Chen et al., 2016a; Sun et al., 2016; Blackport et al., 2019, 2020, 2021; Cohen et al., 2020). The climate effects of sea ice anomalies may be obscured by the chaotic nature of the atmosphere (Overland et al., 2021). Some modeling studies even concluded that winter Eurasian cooling or extreme cold events

are simply due to internal atmospheric variability because of very weak and 93 insignificant simulated atmospheric responses to Arctic sea ice forcing (McCusker et 94 al., 2016; Koenigk et al., 2019). The non-stationary Arctic-midlatitude association due 95 to global warming induces additional uncertainty. For instance, the climatological sea 96 ice northward shift, the impact of autumn Arctic sea ice loss on the winter Siberian high 97 is weakened (Chen et al., 2021). With continuous sea ice loss, the linkage between the 98 99 Arctic and Eurasia exhibits a strong low-frequency fluctuation of warm Arctic-cold Eurasia and warm Arctic-warm Eurasia (Wu et al., 2022). The high sensitivity of mid-100 101 high latitudinal atmospheric responses to the geographical location of Arctic sea ice anomalies and evident differences in climate effects between regional and entire Arctic 102 sea ice change also leads to diversity and uncertainty in previous numerical studies 103 (Chen et al., 2016b; Screen, 2017; Cohen et al., 2020). 104

Recent studies pointed out that autumn sea ice of the East Siberian-Chukchi-105 Beaufort (EsCB) Seas exhibits an increased interannual variability under global 106 warming, and the sea ice loss probably leads to colder northern Eurasia in the 107 108 subsequent early winter and early spring (Ding et al., 2021; Ding and Wu, 2021). A persistent Arctic anticyclonic anomaly, contributed by anomalous upward propagating 109 110 quasi-stationary planetary waves and the associated convergence anomaly in the troposphere and the stratosphere, partly explains the cross-seasonal impacts of autumn 111 regional sea ice. These studies discussed the Arctic-midlatitude association from the 112 perspective of seasonal means, similar as most previous works about BK sea ice change 113 (Wu et al., 2011; Zhang et al., 2018a, b; Cohen et al., 2021). 114

Although the winter mean Eurasian cooling associated with reduced EsCB sea ice 115 is weak and marginally significant (Figure 1a; Ding et al., 2021), especially for East 116 Asia/China there is a possibility of short term high impact events. The related 117 atmospheric anomalies are favorable for rapid and severe cold air outbreaks invading 118 East Asia/China and contributing to extreme low temperatures. Therefore, this study 119 will investigate whether the autumn (September-October) sea ice loss over the EsCB 120 Seas affects the frequency of winter extreme low temperature events over East 121 Asia/China and will explore the mechanisms of such cold air outbreaks. For this 122 purpose, we use statistical diagnosis of observations and sea-ice sensitivity model 123 experiments. 124

125

#### 126 **2. Data and Methodology**

127 Atmospheric monthly and daily mean variables are taken from the NCEP-DOE 128 Reanalysis II with  $2.5^{\circ}$  longitude/latitude resolution, including air temperature, 129 geopotential height and horizontal winds (Kanamitsu et al., 2002). The Monthly mean 130 sea ice concentration (SIC) with a horizontal resolution of  $1.0^{\circ} \times 1.0^{\circ}$  comes from the 131 Met Office Hadley Center (Rayner et al., 2003). The study period covers 42 winters 132 from January 1979 to December 2021. All variables are linearly detrended. We obtain 133 similar conclusions using the raw data (not shown).

The definition of an extreme low temperature day is the daily-mean air 134 temperature (1000 hPa) lower than the 10th percentile of historical records or control 135 experiments, and the sum of extreme days during winter represents the frequency of 136 extreme low temperature. Western-central China experiencing more than one extreme 137 low temperature day is recorded as an extreme event. The interval between two events 138 should be longer than 15 days. There are 12 observed (Table S1) and 185 simulated 139 extreme low temperature events in low SIC years. We mainly focus on the regional sea 140 ice loss in the EsCB Seas (Figure S1a, b) because of its increasing variability (Figure 141 S1c, d), and the area-mean SIC anomalies  $(70.5^{\circ}N-82.5^{\circ}N, 135.5^{\circ}E-119.5^{\circ}W)$ , 142 multiplying by -1.0, are denoted as the EsCB index (Ding et al., 2021). Regression 143 analysis is the primary method to explore the observed association between the 144 frequency of extreme low temperature over central-western China and autumn EsCB 145 sea ice loss. Geopotential height tendency is utilized to portray the feedback of the 146 synoptic-scale eddy to the low-frequency flow (Lau and Holopainen, 1984; Lau, 1988; 147 Lau and Nath, 1991; Cai et al., 2007). Here, the eddy heat flux term associated with 148 baroclinic processes is much smaller (Lau and Holopainen, 1984; Lau and Nath, 1991), 149 150 so we only calculate the eddy vorticity flux term associated with barotropic processes is calculated as follows (Cai et al., 2007): 151

$$F = \nabla^{-2} \left[ -\frac{f}{g} \overline{\nabla \cdot \overrightarrow{V'} \zeta'} \right]$$

153 where  $\overrightarrow{V'}$  and  $\zeta'$  respectively denote the synoptic-scale horizontal winds and relative 154 vorticity, derived from a Butterworth band-pass filter with 2-10 day periods.  $\nabla^{-2}$  is 155 inverse Laplacian,  $\nabla$  is divergence, f is the Coriolis parameter and g is the acceleration

of gravity. We also employ the EP flux  $(F_y = -\rho a \cos \varphi \,\overline{u'v'}, F_z = \rho a \cos \varphi \,\frac{R_f}{HN^2} \,\overline{v'T'})$ 156 and its divergence  $(D_F = \frac{\nabla \cdot \vec{F}}{\rho a \cos \varphi})$  to depict the vertical propagation of quasi-stationary 157 158 planetary wave activity and the wave-mean flow interaction, respectively (Edmon et al., 1980; Plumb, 1985). Here,  $F_z$  ( $F_v$ ) is the vertical (meridional) component,  $\rho$  is the air 159 density, a is the radius of the earth,  $\varphi$  is the latitude, R is the gas constant, f is the 160 Coriolis parameter, H is the scale height, N is buoyancy frequency calculated from the 161 162 temperature data, u is zonal wind, v is meridional wind and T is temperature. The primes and overbars respectively denote zonal deviation and zonal average. The convergence 163 (divergence) of EP flux leads to decelerated (accelerated) westerly winds (Edmon et al., 164 1980; Chen et al., 2002, 2003). 165

166 The Specified Chemistry Whole Atmosphere Community Climate Model version 4.0 (SC-WACCM4; Smith et al., 2014) is employed to investigate the possible role of 167 EsCB sea ice loss for the extreme low temperature events over western-central China. 168 The horizontal resolution is  $1.9^{\circ}$  in latitude and  $2.5^{\circ}$  in longitude, and the vertical 169 resolution has 66 levels extending up to 0.0006 hPa. The control experiments are 170 performed with the climatological monthly SIC and SST averaged over 1982-2001 171 (model-derived data) and other constant external forcings (greenhouse gases, aerosols, 172 solar, etc.) at the year 2000 level. The sensitivity experiments are forced by decreased 173 EsCB sea ice from August to October, calculated from composite detrended Arctic SIC 174 with the detrended EsCB index greater than 0.8o (low sea ice: 1979, 1981, 1990, 2007, 175 2008, 2012; Figure S1b). We only consider three months of sea ice forcing because 176 significant signals with large anomalies mainly occur in August, September and 177 178 October (Figures S2a-c, h-j). Sea ice loss from November to February is very weak with scattered significant regions (Figure S2d-g, k-n). Both, sensitivity and control 179 experiments contain 100 members with different small perturbations added to the initial 180 condition, running from August to February of the next year after the model spins up. 181 The difference between the ensemble mean of the two experiments represents the 182 atmospheric model responses to the prescribed EsCB sea ice loss, whose significance 183 is examined by a two-tailed non-parametric Monte Carlo bootstrap significance test 184





Figure 1 Regressed winter (a) 1000hPa air temperature (shading, unit: °C), (c) extreme low temperature frequency, and (e) 1000 (contour, unit: m) and 500 hPa (shading, unit: m) geopotential height anomalies on normalized autumn (September-October; SO) detrended EsCB index during 1979/80~2020/21. (b), (d), (f) same as (a), (c), (e), but for simulated results. Black dots indicate 90% confidence level. Black rectangles in (c) and (d) represent the range of western-central China (28°N–40°N, 104°E–120°W).

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#### 197 *3.1 Association between winter extreme low temperature and autumn sea ice loss*

The possible contribution of the autumn EsCB sea ice loss to winter temperature 198 change anomalies over East Asia/China and associated Eurasian atmospheric 199 circulation anomalies are displayed in Figure 1, including the regression maps and 200 simulated differences. The autumn EsCB sea ice loss shows a statistically weak link to 201 202 the observed winter-mean cooling in most of China with limited significant range (Figure 1a). However, the reduced autumn EsCB sea ice statistically significantly 203 favors an increase in the frequency of extreme low temperature events over western-204 205 central China (black box in Figure 1c). Compared to the climatological mean (10.71), in six reduced EsCB sea ice years, the area-mean frequency of extreme low temperature 206 (13.83) over western-central China increases by 29.16% (p < 0.1). The sea-ice sensitivity 207 experiments confirm the above observed statistical association. In response to the 208 prescribed autumn EsCB sea ice loss, the model produces relatively weak winter-mean 209 cooling with limited significant range, but simultaneously exhibits a significantly 210 increased frequency of winter extreme low temperature events over western-central 211 212 China (Figures 1b, d). The simulated increased percentage (37.00%) for the frequency of winter extreme low temperature (12.33 versus 9) is close to the observations. Both, 213 214 simulations and observations consistently show that autumn EsCB sea ice decrease can promote more frequent winter extreme low temperature events in western-central China. 215 We hypothesize that atmospheric circulation anomalies associated with regional sea ice 216 melt rapidly invade western-central China in a "pulse-like" manner causing a sudden 217 drop in temperature for a limited time. 218

Figure 2 shows composite maps of the evolution of air temperature and 219 220 geopotential height anomalies at 1000hPa for observed and simulated extreme low temperature events in years with reduced SIC. Day 0 denotes the first day of extreme 221 222 low temperature occurring in western-central China. On the day  $-14 \sim -9$  (initial stage), cold anomalies already appear in northern Siberia (60°N-80°N), east of 60°E, together 223 with positive geopotential height anomalies covering the Arctic region with a 224 southward extension to the Greenland-Northern Europe sector (Figures 2a, g). During 225 the next 6 days (developing stage), positive geopotential height anomalies gradually 226 develop downstream and extend to Lake Baikal around 120°E, conducive to the 227 accumulation and strengthening of cold anomalies over central Siberia (40°N-70°N, 228 30°E-140°E; Figures 2b-d, h-j). Then, following the southward intrusion of the 229

Siberian anticyclonic anomaly, the significant cooling starts shifting southward on the 230 day  $-2 \sim -1$  and finally controls western-central China two days later (outbreak stage; 231 20°N-60°N, 60°E-140°E), replacing the previous warm anomalies and leading to a 232 sharp drop in temperature by 5~6 °C within 2~3 days (Figures 2e-f, k-l). In the middle 233 troposphere, positive geopotential height anomalies cover the Arctic Ocean with 234 southward extension to Northern Europe on the day  $-14 \sim -9$  (Figures S3a, e), gradually 235 strengthen toward the Ural Mountains within the subsequent 4 days (Figures S3b, f) 236 and finally develop downstream to Lake Baikal from day -4 to day 1 (Figure S3c-d, g-237 238 h). To its southeast, significant negative geopotential height anomalies appear around Lake Baikal since day  $-8 \sim -5$  through southeastward horizontal propagation of quasi-239 stationary planetary waves and shift toward the coast of East Asia about 6 days later. 240 This middle tropospheric atmospheric configuration contributes to the Ural blocking 241 high and strengthens the East Asian Trough. This provides favorable conditions for 242 Arctic cold air invading western-central China. The pathway of the above cold 243 anomalies mainly originates east of Novaya Zemlya southward to China (named the 244 hyperpolar path in Bueh et al., 2022). To elucidate the main source and path of extreme 245 low temperature, we further plot the count of cold anomalies  $< -5^{\circ}$ C occurring in each 246 247 Eurasian grid during the evolution of all extreme events based on the evolution characteristic of cold anomalies, including observed and simulated results (Figure 3a, 248 b). Consistent with the former composite maps, when the autumn EsCB sea ice is 249 reduced, the pathways of Arctic cold air inducing western-central China extreme low 250 temperature events are dominated by the hyperpolar path (high frequency in deep red), 251 with approximately 90% (10%) of the pathways deriving from the marginal seas to the 252 253 east (west) of Novaya Zemlya (black dots in Figures 3a, b). Note, that winter air temperature in the climatological mean east of Novaya Zemlya is much colder, which 254 255 explains that the hyperpolar path generally results in strong cooling in western-central China. Consequently, the downstream development of the anticyclonic anomaly over 256 the Arctic Ocean-Northern Europe region is crucial for the rapid southward intrusion of 257 Arctic cold air east of Novaya Zemlya into western-central China. 258 259



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Figure 2 Composite evolution of detrended air temperature anomalies (shading, 261 unit: °C) and detrended geopotential height anomalies (green contour, unit: m) at 1000 262 hPa in (a) day  $-14 \sim -9$ , (b) day  $-8 \sim -7$ , (c) day  $-6 \sim -5$ , (d) day  $-4 \sim -3$ , (e) day  $-2 \sim -1$  and 263 (f) day  $0 \sim 1$  for 12 extreme low temperature events over central-western China during 264 six low EsCB sea ice years. Day  $-14 \sim -9$ , with cold anomalies over northern Siberia, is 265 defined as the initial stage. Day  $-8 \sim -3$ , with cold anomalies over central Siberia, is 266 defined as the developing stage. Day  $-2 \sim 1$ , with cold anomalies over southern Siberia 267 and western-central China, is defined as the outbreak stage. (g) - (l) same as (a) - (f), 268 but for simulated results with 185 extreme low temperature events in 100 members. 269 Black dots and purple contours indicate 90% confidence level. The interval of contour 270 is 20 m. 271 272



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Figure 3 (a) Count (shading) of extreme cold anomalies (<-5 °C) occurring in each 274 Eurasian grid during the evolution of 12 observed extreme cold temperature events. We 275 divide the life cycle of extreme low temperature events into three stages covering the 276 distinct regions according to their evolution characteristics (Figure 2), including initial 277 (day-14 ~ -9; 60°N–80°N, 0°–140°E), developing (day -8 ~ -3; 40°N–70°N, 30°E– 278 140°E) and outbreak (day  $-2 \sim 1$ ; 20°N-60°N, 60°E-140°E) stages. If grids in the 279 selected region appear cold anomalies less than -5°C in either phase, we consider an 280 extreme low temperature event passing through these grids. Black, gray and blue dots 281 represent the mean location of cold anomalies  $< -5^{\circ}$ C in initial, developing, and 282 outbreak phases. The red shading and shift of dots represent the main pathway. Black 283 rectangles in (c) and (d) represent the range of western-central China. (b) same as (a), 284

but for simulated results with 185 extreme low temperature events. Regressed 500 hPa (c) late autumn (October-November-December) geopotential height anomalies (Shading, unit: m), (d) early winter (November-December-January) geopotential height anomalies (Shading, unit: m) and (e) early winter geopotential height tendency anomalies (Shading, unit: m day<sup>-1</sup>) on normalized SO detrended EsCB index. (f) - (h) same as (c) - (e), but for simulated results. Black dots in (c) - (h) indicate 90% confidence level.

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The above anticyclonic anomaly also emerges in the winter-mean anomalous 293 atmospheric circulation shown in Figures 1e, f. As indicated by observations and 294 simulations, the reduction of autumn EsCB sea ice makes the Arctic Ocean and 295 296 Greenland-Northern Europe more prone to positive geopotential height anomalies from the lower to middle troposphere in winter. In the lower troposphere, positive 297 298 geopotential height anomalies cover Northern Europe and expand eastward to East Asia. In the middle troposphere, a northwest-southeastward oriented dipole structure controls 299 300 Eurasia with positive (negative) geopotential height anomalies over the Greenlandnorthern Ural Mountains (Lake Baikal) (Figures 1e, f). Such anomalous atmospheric 301 configuration creates suitable background circulation (decelerated westerlies, reduced 302 meridional potential vorticity gradient and enhanced meridional fluctuation of 303 atmospheric circulation; Francis and Vavrus, 2015; Luo et al. 2018) for the intra-304 seasonal strengthening of the Siberian high (Figure 2), the Ural blocking high and the 305 East Asian Trough (Figure S2), thus leading Arctic cold air rapidly to western-central 306 China (hyperpolar path) and resulting in strong cooling within a few days. 307

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## 309 *3.2 Possible mechanisms: the role of synoptic eddies and planetary waves*

Both, observations and simulations indicate that the increased frequency of 310 extreme low temperature events over western-central China and the anticyclonic 311 anomaly over the Arctic Ocean-Northern Europe are linked to the previous autumn 312 EsCB sea ice loss. To elucidate the possible physical mechanisms, the evolution of 313 observed and simulated 500 hPa geopotential height anomalies from one month after 314 autumn EsCB sea ice minimum is examined. In late autumn (October-November-315 December), significant reduction of autumn regional sea ice increases the upward heat 316 flux from the ocean (Ding et al. 2021; Ding and Wu, 2021) warming the Arctic 317

atmosphere and raising the geopotential height of the entire troposphere. Positive 318 geopotential height anomalies control the Arctic region with southward extension in the 319 North Atlantic sector (Figures 3c, f), then continue to develop and intensify southward 320 in early winter (November-December-January) (Figures 3d, g). Its anomaly center 321 322 finally moves to Greenland-Northern Europe one month later (Figures 1e, f). Previous 323 works discovered that the local synoptic wave-mean flow interaction associated with the weakening (strengthening) of the North Atlantic storm track favors the generation 324 of anticyclonic (cyclonic) forcing to the north (Ding et al., 2017, 2019). Here, we further 325 326 calculated the geopotential height tendency. In early winter, observed and simulated positive geopotential height tendency anomalies dominate the Arctic region with 327 significant and large anomalous signals around Greenland-Northern Europe (Figures 328 3e, h). This indicates that the weakened North Atlantic storm track and associated 329 interaction between synoptic waves and low-frequency flow contributes to the 330 331 southward development and shift of the Arctic anticyclonic anomaly in the North Atlantic sector. In addition, the anomalous upward propagated quasi-stationary 332 planetary wave energy related to the reduced autumn EsCB sea ice also supports the 333 persistent Arctic positive geopotential height anomalies during mid-winter (Zhang et 334 335 al., 2018b; Ding et al., 2021). In early winter, consistent with previous works (Ding and Wu, 2021), using a different model also confirms that planetary wave 2 dominates the 336 increase in the upward propagation of quasi-stationary planetary wave energy, with two 337 anomalous upward EP flux regions in the mid-high latitudes (Figures S4a-c). One 338 upward branch propagates into the lower stratosphere and generates an EP flux 339 convergence anomaly north of 60°N, leading to the decelerated westerly winds and 340 weakened stratospheric polar vortex. The downward propagation of stratospheric 341 anomalies favors the maintenance of the winter Arctic anticyclone anomaly and may 342 provide a potential source for the downstream development of positive geopotential 343 height anomalies around Northern Europe on the intra-seasonal timescale. Another 344 upward branch converges poleward in the upper troposphere around 70°~80°N, directly 345 strengthening the Arctic anticyclonic anomaly by wave-mean flow interaction. 346

As indicated in Figure 2, positive geopotential height anomalies over Northern Europe developing downstream are the vital system for the extreme low temperature outbreaks over western-central China. Consequently, we discuss the possible source of this atmospheric anomaly precursor based on time-height cross sections of observed and simulated area-mean geopotential height anomalies over the Ural Mountains (60°-

90°N, 40-80°E). Positive geopotential height anomalies, indicating the weakened polar 352 vortex, control the lower-middle troposphere throughout the whole outburst of extreme 353 low temperature air masses. Around day  $-10 \sim -9$ , the stratospheric anomalies begin to 354 propagate downward and reach the lower-middle troposphere in the subsequent six days 355 near the Ural Mountains (Figures S5a, b). This favors positive geopotential height 356 anomalies in the troposphere expanding toward Lake Baikal. The timing of downward 357 propagation is highly consistent with the timing (day -8~-3) of the enhancement and 358 downstream development of anticyclonic anomaly around the Ural Mountains (Figures 359 360 S3b-c, f-g). Therefore, the intra-seasonal downward propagation of planetary wave energy contributes to the Northern European anticyclonic anomaly downstream 361 developing and supporting the Arctic cold air east of Novaya Zemlya to rapidly 362 southward intrude western-central China (hyperpolar path). 363

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## 365 *3.3 Different impacts in comparison to Barents-Kara Sea ice*

Autumn sea ice variability in BK Seas, as the second EOF mode (Ding et al. 2021), 366 367 is widely concerned and its climate role shows different feature compared with the EsCB sea ice loss. The sea ice loss exerts much stronger and more extensive impacts 368 369 on winter-mean climate change, inducing significant cooling in most regions of China (Figures S6a, b) with stronger Siberian High and deeper East Asian Trough (Figures 370 S6e, f; Wu et al., 2011). For the extreme low temperature, the significantly increased 371 frequency mainly occurs in northwestern, northeastern and eastern China rather than in 372 the western-central part (Figures S6c, d). 373

The Arctic-midlatitude association is non-stationary and varies with time under 374 global warming (Chen et al., 2021; Wu et al., 2022), and similar results are obtained in 375 our study (Figure S7a). The 19-year sliding correlation coefficients between the autumn 376 EsCB index and the winter frequency of extreme low temperature over western-central 377 China display significant positive correlations (r>0.4, p<0.05) since the late 1990s, with 378 the highest value exceeding 0.7, indicating that the climate importance of EsCB sea ice 379 change is increasingly evident. In contrast, the climate effects of BK sea ice anomalies 380 381 on northwestern China are significant before the late 1990s (r>0.4, p<0.05) and have turned almost insignificant in recent two decades. The rapidly reduced climatological 382 mean sea ice may modulate the above relationship (Chen et al., 2021). 383

In addition, the internal atmospheric variability can also interfere with the Arcticmidlatitude association. In the simulations, we further calculate the signal-to-noise ratio, defined as the relative contributions induced by sea ice forcing to the model internal variability. The increased frequency of winter extreme low temperature forced by EsCB (BK) sea ice loss is about 30~40% of the model internal variability over western-central (northeastern and eastern) China (Figure S7b, c). Therefore, the Arctic-midlatitudes linkage is complex and non-stationary even if only considering the sea ice forcing.



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Figure 4 Schematic diagram of the possible physical pathway for the reduced autumn 393 EsCB sea ice affecting the extreme low temperature events over western-central China. 394 AC and PV respectively indicate anticyclonic anomalies and polar vortex. Shadings in 395 the polar projection map, pressure-latitude map and latitude-longitude map represent 396 397 sea ice anomalies, EP flux divergence anomalies and 1000hPa air temperature anomalies, respectively. Contours in all maps represent the geopotential height 398 anomalies. Purple (green) contours in the latitude-longitude map indicate the location 399 of Eurasian AC when (10 days before) the extreme cold outburst. Dots represent the 400 increased frequency of extreme low temperature. Blue dashed line arrow represents the 401 cold air outbreak path. Blue solid line arrow represents intra-seasonal downward 402 propagation of stratospheric anomalies. Vectors represent the EP flux. Curved arrow 403 represents the synoptic eddy - mean flow interaction and "+" with counterclockwise 404 arrow indicates positive geopotential height tendency anomalies. 405

#### 407 **4. Summary and Discussion**

By analyzing the reanalysis datasets and performing sea-ice sensitivity 408 experiments, this study emphasizes the significant impact of autumn EsCB sea ice loss 409 on climate change over western-central China. It is mainly reflected in the significantly 410 increased frequency of winter extreme low temperature events rather than by winter-411 412 mean cooling. The specific physical processes are summarized in Figure 4. When the previous autumn EsCB sea ice is reduced, enhanced heat flux upward from the open 413 water results in significant local warming and elevation of geopotential height levels 414 415 controlling the Arctic troposphere one month later. Then, under the influence of two possible mechanisms, the positive geopotential potential height anomalies persist into 416 winter. One possible mechanism is the local synoptic eddy-mean flow interaction 417 associated with the weakened North Atlantic storm track that facilitates the Arctic 418 anticyclonic anomaly developing southward with the anomalous center shifting to 419 420 Greenland-Northern Europe. The other possible mechanism is associated with the anomalous upward propagation of quasi-stationary planetary waves and its generated 421 422 wave-mean flow interaction (EP flux convergence anomaly), dominated by planetary wave 2. The tropospheric branch directly strengthens the Arctic positive geopotential 423 424 height anomalies in early winter. The stratospheric branch attenuates the winter polar vortex, favoring the intra-seasonal downward propagation of stratospheric anomalies 425 and contributing to the persistent Arctic anticyclonic anomaly throughout the entire 426 troposphere. Therefore, Greenland-Northern Europe is generally dominated by positive 427 geopotential anomalies in the wintertime. When the stratospheric anomalies propagate 428 429 downward on the intra-seasonal timescale, the Northern European positive geopotential anomalies strengthen and develop downstream within 7 days, which favors severe 430 Arctic cold air east of Novaya Zemlya shifting southward (hyperpolar path) and 431 432 accumulating in Siberia around Lake of Baikal. In the subsequent 2~3 days, these cold anomalies rapidly invade western-central China, bringing a sudden sharp drop in 433 temperature and explaining the increased frequency of extreme low temperature there. 434 435 The climate role of autumn EsCB sea ice change in western-central China has become increasingly significant in recent two decades. In contrast, autumn sea ice loss in BK 436 Seas favors more frequent extreme low temperature events over northeastern and 437 eastern China, whose dominant effects occur before the late 1990s. 438

Although the simulations capture the main features of the observations, certain
 simulated deviations still exist. For example, the location of the simulated winter-mean

Arctic anticyclonic anomaly is farther north than the observed. On intra-seasonal 441 timescale, compared to observations, the simulated Siberian cold anomalies from day -442 14 to day -5 shift eastward and are more significant, accompanied by stronger positive 443 geopotential height anomalies over northern Eurasian. These simulated deviations may 444 be related to the inaccurate model descriptions and lack of other external forcings such 445 as Eurasian snow cover and sea surface temperature in Pacific or Atlantic Ocean. 446 Besides downward propagating stratospheric signals, other factors such as the intra-447 seasonal oscillation may also trigger the downstream development of Northern 448 449 European anticyclonic anomaly. These phenomena are out of scope of our discussion. In addition, favorable initial atmospheric conditions or internal atmospheric variability 450 for autumn EsCB sea ice affecting extreme low temperature in China needs further 451 452 investigation.

453

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460

### 461 **Open Research**

The Monthly mean sea ice concentration (SIC) from the Met Office Hadley Center are 462 available via https://www.metoffice.gov.uk/hadobs/hadisst/data/download.html. The 463 NCEP-NCAR (National Center for Environmental Prediction) - DOE (Department of 464 Reanalysis Π dataset is available 465 Energy) at https://psl.noaa.gov/data/gridded/data.ncep.reanalysis2.html. 466

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## **Figure Supplement**



Figure S1 (a) Regressed SO sea ice concentration (SIC) anomalies (Shading, unit: %) on normalized SO detrended EsCB index, (b) Composite detrended SIC anomalies (Shading, unit: %) for six low EsCB sea ice years (1979, 1981, 1990, 2007, 2008, 2012), (c) the original EsCB index (black line) and its detrended component (bar chart), and (d) 15-year sliding standard deviation (red dot line) of autumn EsCB index and associated Mann-Kendall (M-K) test (dashed and solid green line). Black dots in (a) and (b) indicates 90% confidence level. Gray double (single) dotted line in (c) indicates 95% (90%) confidence level.



Figure S2 (a) Regressed sea ice concentration (SIC) anomalies (Shading, unit: %) on normalized autumn detrended EsCB index from (a) August to (g) February. (h) - (n) same as (a) - (g), but for composite detrended SIC anomalies (Shading, unit: %) for six low EsCB sea ice years (1979, 1981, 1990, 2007, 2008, 2012). Black dots in (a) and (b) indicates 90% confidence level.



Figure S3 Composite evolution of detrended geopotential height anomalies (shading, unit: m) at 500 hPa in (a) day  $-14 \sim -9$ , (b) day  $-8 \sim -5$ , (c) day  $-4 \sim -1$  and (d) day  $0 \sim 1$  for 12 extreme low temperature events over central-western China during six low EsCB sea ice years. (g) – (l) same as (a) - (f), but for simulated results with 185 extreme low temperature events in 100 members. Black dots contours indicate 90% confidence level. The interval of contour is 20 m. Green line arrow indicates the downstream development of the Northern European anticyclonic anomaly.



Figure S4 (a) Simulated early winter zonal mean EP flux (vector, unit:  $m^2 s^{-2}$ ), EP flux divergence (shading, unit:  $m s^{-1} day^{-1}$ ) and geopotential height anomalies (green contour; interval: -10, 10, 30, 50 m). (b), (c) same as (a), but for the zonal mean EP flux and EP flux divergence anomaly of planetary wave 1 and 2, respectively. Dots for shadings indicate the 90% confidence level. Vectors only depict the part exceeding  $10^5 m^2 s^{-2}$ .



Figure S5 Time-height cross sections from day -14 to day 1 of (a) area-mean detrended geopotential height anomalies (shading; unit: m) over the Ural Mountains (60°-90°N, 40-80°E) during six low EsCB sea ice years. (b) same as (a), but for simulated results. Dots for shadings indicate the 90% confidence level.



Figure S6 Regressed (a) 1000hPa air temperature (shading, unit: °C), (c) extreme low temperature frequency, and (e) 1000 (contour, unit: m) and 500 hPa (shading, unit: m) geopotential height anomalies on normalized detrended autumn (September-October) BK (70.5°N–80.5°N, 40.5°E–134.5°E) index (Ding et al. 2021) during 1979/80-1999/00. (b), (d), (f) same as (a), (c), (e), but for simulated results with 100 members forced by composite detrended Arctic SIC with the detrended BK index greater than 0.8σ (low sea ice: 1983, 1984, 1985, 1995, 2009, 2011, 2018). Black dots indicate 90% confidence level.



Figure S7 (a) Sliding correlation coefficients between autumn detrended EsCB (BKL) index and winter detrended extreme low temperature frequency over western-central (northwestern) China with a 19-year window. Dots indicate 95% confidence level. (b) Signal-to-noise ratio (SNR) is defined as the frequency change of extreme low temperature induced by EsCB sea ice forcing divided by the model internal variability (standard deviation in control experiments). (c) same as (b), but for BK sea ice loss.

Year	Beginning date for each event	Days
1980	January 10, January 29	1, 12
1981	December 1	5
1982	January 16	2
1990	December 1	3
2008	January 12, December 4, December 21	36, 2, 3
2009	January 11, February 25	3, 3
2012	December 22	10
2013	February 7	3

Table S1 The beginning dates and accumulative days for 12 extreme low temperature events in the observation