Winter Euro-Atlantic blocking activity less sensitive to climate change than previously evaluated

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July 3, 2023

Abstract

Winter Euro-Atlantic atmospheric blocking events have significant socioeconomical impacts as they cause various types of weather extremes in a range of regions. According to current climate projections, fewer of these blocking events will occur as temperatures rise. However, the timing of such a reduction is currently highly uncertain. Meanwhile, recent studies indicate that using climate models with high enough ocean resolutions to simulate mesoscale eddies improve simulated winter Euro-Atlantic blocking events significantly. In this paper, we show from a large ensemble of climate simulations based on the highest emission scenario that largely prominent and coarsely resolved non-eddying climate models project a noticeable significant decline in blocking frequencies from the 2030s-2040s, whereas blocking statistics in eddy-permitting simulations are noticeably decreasing only from years 2060s. Our result suggests with a strong level of confidence that winter blocking activity over the next several decades will keep being dominated by internal variability.



Supplementary Fig. S1: Same as Fig. 1a-e for the hist period (1979-2014) in 46 NE and 15 EP GCM simulations.

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8	Abstract:
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10	impacts as they cause various types of weather extremes in a range of regions. According
11	to current climate projections, fewer of these blocking events will occur as temperatures
12	rise. However, the timing of such a reduction is currently highly uncertain. Meanwhile,
13	recent studies indicate that using climate models with high enough ocean resolutions to
14	simulate mesoscale eddies improve simulated winter Euro-Atlantic blocking
15	events significantly. In this paper, we show from a large ensemble of climate simulations
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20	confidence that winter blocking activity over the next several decades will keep being
21	dominated by internal variability.

22 **1.** Introduction:

Atmospheric blocking is defined by large synoptic-scale pressure anomalies at midlatitudes that alter the midlatitude jet stream's pathways and redistribute precipitation and temperatures across many regions (Woolings et al. 2018). These pressure anomalies can last several days, thereby causing persistent weather extreme conditions at several locations (Kautz et al. 2022). The nature and amplitudes of underlying extremes depend on the timing of the blocking event (Davini and D'Andrea 2020), the location of blocked areas (Brunner et al. 2018), and background climate conditions (De Vries et al. 2013, Kautz et al. 2022, Woolings et al. 2018).

30

Winter blocking events occur frequently in the Euro-Atlantic sector and have a considerable impact on winter regimes across Europe (Cassou et al. 2004, Michel et al. 2023, Terray et al. 2005). These winter blocking conditions result in droughts (Sillman and Croci-Maspoli 2009), heavy precipitation (Kautz et al. 2022), extreme snowfalls (Cattiaux et al. 2011, de Vries et al. 2013), or cold spells (Cattiaux et al. 2010), all of which can have considerable impact on socioeconomic sectors such as energy or agriculture (Woolings et al. 2018).

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38 Upper-level tropical warming and near-surface Arctic warming are expected to result in stronger 39 zonal winds and consequently in fewer Winter Euro-Atlantic Blocking (WEAB) events (Kennedy 40 et al. 2016, Woolings et al. 2018). A wide range of studies showed that most state-of-the-art 41 General Circulation Models (GCMs) of climate of the last three phases of the Coupled Model 42 Intercomparison Project (CMIP3, CMIP5 and CMIP6) simulate this decrease in WEAB 43 frequencies far before the end of the century, both for moderate and large emission scenarios 44 (Davini and D'Andrea 2020, Dunn-Singouin and Sun 2013, Fabiano et al. 2021, Kennedy et al. 45 2016, Masato et al. 2013, Matsueda and Endo 2017, Sillman and Croci-Maspoli 2009). However, 46 Bacer et al. (2022) recently showed no significant decrease in WEAB frequencies for half of the 47 simulations from six CMIP6 GCMs (selected for their relatively good performance in simulating

historical WEAB frequencies observations) by the end of the century and under the highest
emission scenario (SSP5-8.5). Also, CMIP5 GCMs generally do not simulate significant changes
in future structures of main winter Euro-Atlantic patterns (Huguenin et al. 2020). Woolings et al.
(2018) already emphasized that the large disparity in projected WEAB frequencies from GCMs
(Masato et al. 2013) and their generally poor ability to simulate observed WEAB events (Davini
and D'Andrea 2016), both cause large uncertainties on the occurrence, and if so, the timing of a
decline in WEAB frequencies under climate change.

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56 The ability of GCMs to simulate WEAB is notably hampered by extensively documented persistent 57 biases, such as in simulated tropical convection (Gollan et al. 2019), transient eddy forcing 58 (Berckmans et al. 2013, Davini et al. 2017), model orography (Berckmans et al. 2013), or sea 59 surface temperatures (Athanasiadis et al. 2022, Michel et al. 2023, Scaife et al. 2011). Increased 60 atmospheric resolution was shown to improve the simulation of WEAB events notably through 61 reduced biases in simulated transient eddies (Davini et al. 2017, 2021) and improved orographic 62 resolution (Berckmans et al. 2013). Two recent studies illustrated that the bias reduction in 63 simulated sea surface temperature patterns achieved from the presence of mesoscale eddy-64 permitting (EP) ocean models also improves the simulation of WEAB events from GCMs 65 (Athanasiadis et al. 2022, Michel et al. 2023). These results were explained by reduced biases in 66 sea surface temperature gradients (Athanasiadis et al. 2022) shaped by better representation of 67 oceanic fronts (Hewitt et al. 2016, 2020, Michel et al. 2023, Scaife et al. 2011). This improves the 68 realism of simulated low level baroclinicity and vertical heat and moisture fluxes contributing to 69 blocking genesis (Cheung et al. 2023, O'Reilly et al, 2016). Earlier studies on future WEAB 70 behavior were predominantly based on GCM simulation ensembles mostly composed of less 71 accurate non-eddying (NE) GCMs, which are still highly prominent in CMIP6 (Eyring et al. 2016). Therefore, the present study seeks at refining projections of future WEAB activity in the context 72 73 of climate change by comparing two extensive sets of EP and NE GCM future simulations.

75 2. Data and Methods:

76 <u>2.1. Data:</u>

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78 We employ up to 58 future simulations from 21 GCMs that participated to CMIP6 and High 79 resolution MIP (HighresMIP)/PRIMAVERA activities (Supplementary Table S1). We focus our 80 investigation only on simulations for the largest greenhouse gas emission experiments (i.e., 81 SSP5-8.5), as HighResMIP/PRIMAVERA future simulations were conducted only for this future 82 scenario (Haarsma et al. 2016). Therefore, 17 future simulations were taken from the highres-83 future experiment of the HighResMIP/PRIMAVERA activity, and 41 were taken from the SSP5-84 8.5 experiment of the Scenario MIP CMIP6 activity. We choose up to four simulation members per 85 GCM/experiment pair where available to account for the influence of the large internal variability 86 in simulated WEAB events (Supplementary Table S1).

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88 It must be noted that EP GCMs are rare in CMIP6 future simulations but more common in 89 HighResMIP/PRIMAVERA future simulations, where the majority ends in 2050. As a result, future 90 GCM simulations will be studied for different periods: near future (NF, 2015-2050), far future (FF, 2051-2099), and the 21st century (21C,2015-2099), with 21C representing simulations from NF 91 92 that have been conducted until at least 2099 (Supplementary Table S1). In the end, NF is made 93 up of 16 EP simulations and 42 NE simulations (58 in total), while FF and 21C are made up of 7 94 EP simulations and 35 NE simulations (42 in total, Supplementary Table S1). It should be 95 emphasized that only a single NF simulation (HadGEM3-GC31-HH) was produced by a GCM with 96 an ocean resolution higher than 0.1°x0.1°, which is considered eddy-rich. However, because just 97 one of these simulations is being investigated here, it will be handled in the EP GCM group in the 98 section that follows.

Finally, we will compare the results of future simulations for NF and FF periods to those in presentday simulation outputs using 61 additional historical simulations from 1979 to 2014 based on fairly comparable GCM simulations ensembles (15 EP and 46 NE simulations, Supplementary Table S1). Ultimately, the present study is based on a very large ensemble of 119 simulations, both future (58) and historical (61) (Supplementary Table S1).

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In the following, data from the different GCM simulations used (Supplementary Table S1) are
 initially interpolated on regularly separated 1°x1° grids.

- 108
- 109 <u>2.2. Methods:</u>

110 <u>2.2.1 Blocking detection</u>

We detect blocking events using on a long-established technique based on daily 500 hPa geopotential heights (Z500) fields providing a time-varying 2-dimensionnal blocking index (Sherrer et al. 2006). Other techniques, for instance based on potential vorticity fields or Z500 anomalies (Woolings et al. 2018) are more biased in the context of transient climate variations.

115

116 The method is based on detecting blocking situations for each grid point wherever a reversal in 117 the meridional geopotential heights gradient is observed. Therefore, we consider that a given grid 118 point with longitude λ_0 and latitude ϕ_0 is blocked at time *t* if the three following conditions C1 - C3119 are fulfilled:

120
$$C1: GHGS(\lambda_0, \phi_0) > 0$$

121
$$C2: GHGN(\lambda_0, \phi_0) < -10$$

122 $C3: GHGS_2(\lambda_0, \phi_0) < -5$

123 *GHGS*, *GHGN*, and *GHGS* $_2$ are given by:

124
$$GHGS(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_0) - Z500(\lambda_0, \phi_S)}{\phi_0 - \phi_S}$$

125
$$GHGN(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_N) - Z500(\lambda_0, \phi_0)}{\phi_N - \phi_0}$$

126
$$GHGN(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_S) - Z500(\lambda_0, \phi_{S_2})}{\phi_S - \phi_{S_2}}$$

127 With $\phi_S = \phi_0 - 15^\circ$, $\phi_N = \phi_0 + 15^\circ$, and $\phi_{S_2} = \phi_0 - 30^\circ$. Therefore, for a given spatial field { Λ, Φ } 128 and for *n* daily timesteps, the 2D blocking index defined as:

129
$$M(\lambda, \phi, t) = \begin{cases} 1 \text{ if } C1 \cap C2 \cap C3 \\ 0 \text{ otherwise} \end{cases}, \lambda \in \Lambda, \phi \in \Phi, t = t_1, \dots, t_n \end{cases}$$

130 In addition, all grid points in { Λ , Φ } that are blocked for less than four days are turned to 0 to avoid 131 spurious detection of short-term, not impactful, and isolated blocking conditions across the grid 132 (Davini et al. 2012, Davini et al. 2017, Michel et al. 2023).

133

134 <u>2.2.2 Blocking clustering</u>

135 Clustering atmospheric blocking patterns is a common task that enables to study blocking activity 136 in terms of both frequencies and spatial patterns in an automatic way and has been used 137 frequently (Bacer et al. 2022, Cassou et al. 2004, Fabiano et al. 2021, Michelangeli et al. 1995, Michel et al. 2023, Terray et al. 2005). However, these studies differ with respect to the variable 138 139 that is used for clustering blocking patterns, namely: sea level pressure (Cassou et al. 2004, 140 Terray et al. 2005), geopotential heights at 700 hPa (Michelangeli et al. 1995), principal 141 components of Z500 (Fabiano et al. 2021), and 2D blocking index of blocking conditions longer 142 than 4 days (Michel et al. 2023). The clustering here consists of clustering days (winter ones here, 143 *i.e.* days from December, January, and February) based on their pattern similarities. This 144 approach is useful to study different types of blocking patterns and their respective frequencies, 145 as they were shown to have different and sometimes opposite meteorological fingerprints (Michel 146 et al. 2023). Using such an approach, for instance, Fabiano et al. (2021) evaluated, using a large 147 ensemble of CMIP5 and CMIP6 GCM simulations, that only specific blocking patterns were 148 significantly decreasing in the future as a result of rising temperatures. The clustering used here

149 (Michel et al. 2023) includes much less noisy signals than pressure variables (*i.e.* sea level 150 pressure and geopotential heights) used by other studies (Cassou et al. 2004, Fabiano et al. 151 2021, Michelangeli et al. 1995, Terray et al. 2005). For the k-means to be applied such that days 152 are clustered based on their blocking pattern similarities, the 3D matrix M is turned into a 2D 153 matrix denoted X, where rows are time steps and columns are all the grid points in { Λ , Φ } (Michel et al. 2023). First, k centroids $\mu_1^{(1)}, \dots, \mu_k^{(1)}$ for respective clusters $c_1^{(1)}, \dots, c_k^{(1)}$ are first randomly 154 155 drawn in the { Λ , Φ } space and blocking patterns for each time step are affiliated to the cluster with the closest centroid in terms of Euclidian distance. This procedure is repeated until convergence 156 157 (when centroids stop changing), where the update of centroids at each occurrence is given by:

158
$$\mu_j^{(t+1)} = \frac{1}{\#\{c_j^{(t)}\}} \sum_{i \in c_j^{(t)}} X_i, j = 1, \dots, k$$

159

160 Each k-means model in the study is ran 200 times with different random initializations of centroids. The set of random initial centroids with lowest Euclidian distances between cluster members and 161 162 their respective centroid is retained. The k parameter is generally chosen from reanalysis data 163 using optimization methods (Cassou et al. 2005, Michel et al. 2023). However, former studies did 164 not necessarily found or used similar values: k=4 (Cassou et al. 2005, Fabiano et al. 2020, 2021, 165 Terray et al. 2005), k=5 (Michel et al. 2023) or k=6 (Falkena et al. 2020) WEAB types. Here, as 166 we used the clustering based on the same variable as Michel et al. (2023), we use the parameter 167 k=5 and will refer to the five main patterns identified from ERA5 as: Western Europe (WE), 168 Greenland (Gr.), North Sea (NS), Baltic Sea (BS) and Scandinavia (Sc.).

169

170 In ERA5 data, this clustering for 5 main WEAB pattern is made with a k=6 parameter where only 171 the above five clustered patterns are kept. The 6th cluster constitute a "neutral" or "non-blocking" 172 group and accounts for about 72% of all DJF days in ERA5 between 1979 and 2014 (Michel et al. 2023). For GCMs, we attribute each daily blocking pattern to the closest centroid based on
Euclidian distances. In this procedure, the "neutral" group from the reanalysis clustering is also
used to allow the attribution of GCM blocking patterns to this category when they are not
resembling one of the other five.

177

178 **3.** <u>Results:</u>

179 Spatial distribution of projected WEAB frequencies and trends

180 Mean projected blocking frequencies averaged throughout the NF period (2015-2050) for both 181 NE and EP GCM ensembles (Supplementary Table S1) are shown in Fig. 1a,b, and significant 182 differences at the 90% confidence level between the ensembles are shown in Fig. 1c. It appears 183 that EP GCMs simulate a significantly higher number of blocking events by 2050, where 184 significant differences cover most regions of high blocking occurrence identified by previous studies, such as Greenland, Nordic Sea, or Scandinavia (Fig. 1a-c, Fabiano et al. 2021, Michel 185 186 et al. 2023, Terray et al. 2005). Furthermore, we see a wide area including the Baltic and Nordic 187 seas, British Islands, and the northeastern Atlantic where blocking frequency trends are notably 188 negative in the NE GCM ensemble (Fig. 1d). For EP GCMs, on the other hand, two tiny areas 189 with positive (Iceland) and negative (Greenland) trends are detected, but the fraction of grid points 190 with significant trends is lower than the 10% first species risk of the Student t-test used (7%, Fig. 191 1e). According to this preliminary analysis EP GCMs simulate more blocking events on average 192 by 2050 than NE GCMs (Fig. 1a-c) and also do not project any significant decrease in 193 frequencies until 2050 (Fig. 1d,e).

194

The increased simulated blocking frequencies found for EP GCMs over the historical (Michel et al. 2023) and NF periods (Fig. 1a-c) persist during the *21C* period (Fig. 1f-h). However, the geographical range of significant differences between the two groups is smaller than what we found for NF (Fig. 1c,h). In terms of trends, both model groups show significantly lower blocking frequencies throughout the *21C* period. These significant decreases are geographically consistent in the NE GCM simulation ensemble, but not in the EP ones, although most locations with high blocking prominence experience a significant decline (Fig. 1i-j). Supplementary Fig. S1 shows the same ensemble analyses as Fig. 1 for the historical simulations from Supplementary Table S1, where EP GCM WEAB frequencies are higher, consistent with Athanasiadis et al. 2022 and Michel et al. 2023, and also show significant decrease in NE simulations but not in EP simulations.

205 To further test the robustness of the above findings across GCMs, Monte Carlo tests and Student 206 t-tests based on area-averaged blocking frequencies and trends were performed over the study 207 area (50W-50E, 30-72.5N) for three GCM simulation periods: historical (1979-2014, hist), NF, and 208 21C (Supplementary Table S1, Fig. 2). For the three periods, the Monte-Carlo tests determine 209 confidence levels based on proportions of mean blocking frequencies from NE GCM 210 samples of same size as the smaller EP GCM ensembles (Fig. 2a). For hist and NF, it appears 211 that none of the 10,000 randomly drawn NE simulation samples have mean area-averaged 212 blocking frequencies that reach the one of the EP GCM simulations (Fig. 2a), which strongly 213 supports the results shown in Fig. 1. The significance level for 21C, on the other hand, is lower 214 but still higher than 95% (Fig. 2a), also consistent with differences seen between Fig. 1c and Fig. 215 1h.

216 For trends, Student t-tests show that both 95% and 99% confidence intervals for mean area-217 averaged trends from NE GCM simulations do not include 0 for all three periods, including hist 218 (Fig. 2b). On the other hand, EP GCM simulations only indicate significantly decreasing blocking 219 frequencies at the 99% confidence level when the entire 21st century is taken into account, but 220 trends are not significant for NF and hist. This result demonstrates that the response in WEAB 221 frequencies to strong anthropogenic forcing differs significantly between EP and NE simulations, 222 with EP GCMs simulating the same lower blocking frequencies far later in the century than NE 223 GCMs. Furthermore, as observed in historical simulations (Athanasiadis et al. 2022, Michel et al.

2023, Supplementary Fig. S1), EP GCMs simulate substantially higher and likely more realistic
WEAB frequencies for both NF and 21C periods (as they do so for historical, Michel et al. 2023),
implying that such GCMs are better suited to studying future winter blocking in Europe.

227

228 *Future evolution of observed main WEAB patterns*

In addition to the spatial analyses shown in Figs. 1-2, we clustered WEAB events in both historical
and future simulations into five primary blocking patterns, as performed in previous
studies (Cassou et al. 2004, Fabiano et al. 2021, Falkena et al. 2020, Michelangeli et al. 1994,
Michel et al. 2023, Terray et al. 2005).

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We find no significant changes in the mean spatial extent of simulated blocking patterns over time, and neither EP nor NE GCM demonstrate any significant shift in the location of their assigned patterns (Fig. 3a,d,g,j,m). Indeed, for each blocking footprint, the simulated WEAB frequency tends to be more sensitive to NE and EP ocean configurations than to increasing temperatures, although the configuration effect is relatively small.

239

240 In terms of temporal evolution of these blocking patterns, we observe that the WEAB response to 241 anthropogenic forcing differs significantly over time, both between the NE and EP GCM 242 ensembles and between the simulated blocking patterns (Fig. 3b.e.h.k.n). Based on trends 243 evaluated across 36-year intervals (i.e., the same size as hist and NF periods), we find that 244 frequencies from a particular blocking type (i.e., Baltic Sea) do not show a globally consistent 245 decline over time for both NE and EP GCM ensembles (Fig. 3k,I). Greenland blocking declines 246 considerably more consistently in NE GCMs than in EP GCMs, where just a few 36-year slices 247 show significant decreases, although these are sufficient to significantly reduce Greenland 248 blocking occurrence by the end of the century (Fig. 3e,f). The opposite is true for Scandinavian 249 blocking, which has a significant decrease in frequency in EP GCM simulations late in the century 250 (from about 2060), but no globally significant decline in NE GCM simulations (Fig. 3n,o). Both NE 251 and EP GCM ensembles suggest a decline in frequencies after 2050 for the most prevalent 252 Western Europe blocking events (Michel et al. 2023), however the trend appears to be mitigated 253 at decadal to multidecadal time frames for EP GCMs (Fig. 3b,c). Furthermore, EP GCMs show a 254 time frame between 2000 and 2040 where Western Europe's blocking frequencies increase 255 significantly. Finally, WEAB frequencies over the Nordic Sea drop in both GCM groups, but the 256 timing differs considerably, occurring between 2015-2060 for NE GCMs and 2030-2085 for EP 257 GCMs.

258

259 Global view of future WEAB activity

260 Substantial disparities in simulated future WEAB frequencies exist between the NE and EP 261 GCMs, both geographically (Figures 1 and 2) and temporally (Figure 3). In Fig. 4, we present an 262 overall summary of results depicted above, where we show the total WEAB frequencies for EP 263 and NE ensembles (Fig. 4a) estimated as the sums of ensemble-means of timeseries from each 264 simulated blocking pattern for NE and EP GCMs, (Fig. 4b), as well as successive 36-year trends 265 of these timeseries. The magnitude of trends presented in Fig. 4b cannot be directly compared 266 between EP and NE simulation ensembles because they were computed from time series with 267 significantly different averages (Michel et al. 2023).

268

The assumption that EP GCMs have a smaller WEAB response to anthropogenic forcing than NE GCMs derived from spatially distributed trends (Figs. 1-2) and several of the clustered blocking patterns (Fig. 3) is confirmed here for the total frequencies for all blocking patterns (Fig. 4). Indeed, we find that total WEAB frequencies in NE GCMs start decreasing significantly (at the 95% confidence level) for 36-year time frames as early as 2000 (thus centered around 2018 and noticed around 2036) and continue to decrease significantly for following time frames until the end of the century (Fig. 4). The ensemble of EP GCM simulations tells a completely different
story, as 36-year trends in total WEAB frequencies start to be significantly and consistently
negative only for time frames centered around 2040 (thus noticed around 2060), representing a
more than 20-year delay compared to NE GCMS.

279

280 Beyond the significant downward trend in EP GCM WEAB frequencies found in the second half 281 of the 21st century (Fig. 4), the ensemble mean of these GCMs shows alternate decadal to 282 multidecadal periods with either significant or non-significant downward trends. This 283 overall results in an overall decrease in WEAB frequencies for EP GCMs but contrasts sharply 284 with the uninterruptedly significant negative trends observed over the entire 21st century for the 285 NE GCM ensemble. Hence, EP GCMs appear to be responding less to human-caused rising 286 temperatures (Fig. 1-4), and the role of decadal to multidecadal underlying processes such as 287 those associated with the North Atlantic/Arctic Oscillation are more important in shaping future 288 WEAB activity.

289

290 Conclusions and Discussion:

291 Our study was motivated by recent findings demonstrating higher accuracy in simulating WEAB 292 frequencies for EP GCMs compared to much more prevalent NE GCMs in the current CMIP6 293 generation (Athanasiadis et al. 2022, Michel et al. 2023), whereas EP GCMs were simply absent 294 in CMIP3 and CMIP5. Our results support those of most prior studies suggesting that the 295 occurrence of WEAB events will be significantly reduced by the end of the century (Davini and 296 D'Andrea 2020, Dunn-Singouin and Sun 2013, Fabiano et al. 2021, Kennedy et al. 2016, Masato 297 et al. 2013, Matsueda and Endo 2017, Sillman and Croci-Maspoli 2009). However, we discovered 298 significant differences in how NE and EP GCMs simulate the evolution of future WEAB events 299 under high human-induced forcing (Figs. 1-4). Indeed, while NE GCMs suggest that WEAB 300 frequencies may begin to decrease considerably over the next decade (Figs. 1, 4a), EP GCMs

301 suggest a delay of more than 20 years. These findings indicate that no significant decrease in the 302 occurrence of blocking-related winter extremes in Europe is expected in the coming decades. 303 Furthermore, we determined that EP GCMs simulate much more WEAB events in future 304 simulations than NE GCMs as was found for historical simulations (Michel et al. 2023). Given the 305 current need for a more precise picture on the future WEAB projections due to widely disparate 306 projections from NE GCMs (Fig. 2, Woolings et al. 2018), the use of EP GCM simulations in future 307 research is crucial.

308

In addition to the delayed response in decreased WEAB frequencies for EP GCMs, we also found that some 36-year long periods late in the century still experience no significant decrease (Fig. 4) whereas this is almost permanent in NE GCM simulations as early as from the 2030s. This result indicates a stronger role for internal variability in future WEAB occurrences in EP GCMs.

313

314 Due to the current low number of EP GCM simulations available in CMIP6 and HighResMIP/PRIMAVERA ensembles, we were unable to investigate other future simulations 315 316 such as those under intermediate (e.g., SSP2-4.5 and SSP3-7.0) or low (e.g., SSP1-2.6) 317 greenhouse gas emission scenarios. In this respect, Hausfather and Peters (2020) recalled that 318 the most widely used and widely communicated future scenario (i.e. SSP5-8.5) that we study here 319 is extremely useful for scientists attempting to understand the climate response to strong CO_2 320 forcings. However, its forcing is so large that it remains largely unlikely and unrealistic to 321 effectively happen in reality given current socioeconomic trends (Hausfather and Peters 2020). 322 As a result, the delayed decrease in WEAB frequencies found for EP GCMs over NE GCMs may 323 be even larger in the real-world, because future CO₂ emissions may be lower than those 324 prescribed for the GCM simulations investigated here (Hausfather and Peters 2020). Therefore, 325 it is likely that no significantly decreasing WEAB frequencies are expected to be noticed over the 326 next several decades, and internal variability may continue to primarily influence WEAB activity throughout this time (Woolings et al. 2018). However, a thorough examination of the response of
WEAB frequencies under more realistic emission scenarios in a large enough EP GCM simulation
ensemble has yet to be performed to corroborate the aforementioned statement. Current CMIP6
simulations do not yet allow for such an important analysis, and modeling centers handling EP
GCMs are encouraged to include simulations for intermediate emission scenarios in future CMIP
updates.

333 If we had limited our research to the NF simulation period (2015-2050), for which most EP GCM 334 simulations were done (Supplementary Table S1), we would have reached somewhat similar 335 findings as we do now, but with a much more incomplete picture. Indeed, downward trends in 336 WEAB frequencies in EP GCM simulations begin later than 2050, which would not have been 337 noticed if restricting the analysis to NF. This also reinforces the necessity for longer future 338 simulations using EP GCMs in future CMIP protocols, as these GCMs have been demonstrated 339 to simulate several climate processes far more correctly thans the currently widely used NE GCMs 340 (Athanasiadis et al. 2022, Delworth et al. 2012, Hewitt et al. 2016, 2020, Michel et al. 2023, 341 Moreno-Chamarro et al. 2021, Roberts et al. 2018).

342

345 **Open Research section**

346 *Data availability statement:*

347 Original CMIP6 (Eyring et al. 2016) and HighResMIP/PRIMAVERA (Haarsma et al.

348 2016) data are available through CEDA Earth System Grid Federation portals (https://esgf-

- 349 index1.ceda.ac.uk/search/cmip6-ceda/, https://esgf-index1.ceda.ac.uk/search/primavera-ceda/).
- 350 ERA5 reanalysis data (Hersbach et al. 2020) are available on the Copernicus Climate Change
- 351 Service (C3S) Climate Data Store (https://cds.climate.copernicus.eu/cdsapp#!/home). Data
- 352 produced through the study are available on a Zenodo repository with following DOI:
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Figure 1: (a,b) Mean Winter Euro-Atlantic Blocking (WEAB) frequencies averaged over the near
future (NF) period (2015-2050) in 42 non-eddying (NE) and 16 eddy-permitting (EP) General
Circulation Model (GCM) future simulations under the highest greenhouse gas emissions scenario.
(c) Difference in mean WEAB frequencies between EP (b) and NE (a) GCM simulations. Green
colors indicates where WEAB frequencies are significantly higher in EP (resp. NE) GCM simulations
at the 90% confidence level based on a two-tailed Student t-test. White colors indicate where there
is no significant difference in WEAB frequencies from NE and EP GCMs based on the same test.

- 556 (d,e) Trends in percentage points per year in WEAB over the period under the highest greenhouse
- 557 gas emission scenario, for NE and EP GCMs respectively. White colors indicate where trends are
- 558 not significant at the 90% confidence level based on two-tailed regression slop Student t test. (f-j)
- 559 Same as (a-e) for the 21C period (2015-2099) in 35 NE and 7 EP GCM simulations.



562 Figure 2: Area-averaged WEAB frequencies (a) and trends (b) over the study area: 50W-50E, 30N-563 72.5N. Light, normal, and dark blue (resp. green) boxplots give WEAB frequencies EP (resp. NE) GCM 564 simulations for hist (1979-2014), NF (2015-2050) and 21C (2015-2100). Purple circles (resp. squares) 565 indicate averages for each boxplot EP (resp. NE) GCM simulations group. (a) Yellow and red squares 566 indicate 95% and 99% Monte-Carlo significance levels for EP GCM area averages. These levels are 567 determined for each period (hist, NF, 21C) by sampling 10,000 groups of NE GCMs with the same 568 number of GCM ensembles. (b) Yellow and red dashed lines indicate 95% and 99% Student 569 confidence interval for each boxplot. The horizontal black dashed line indicates the 0 value.



572 Figure 3: (a) ERA5 in-cluster blocking frequencies (colors) together with 20% contours of EP 573 (resp. NE) GCM ensembles given as light, normal, and dark green (resp. blue) lines for hist (1979-574 2014), NF (2015-2050) and FF (2051-2100) periods, respectively. For the Western Europe 575 blocking type. (b) Blue (resp. green) line indicates the composite time series of blocking frequency 576 available EP (NE) GCM simulations over hist, NF, and FF periods. Vertical red lines separate hist, 577 NF, and FF periods. For the Western Europe blocking type. (c) Trends in blocking frequencies 578 over 36-year time frames calculated as linear regression slopes of blocking frequencies against 579 time. Diamonds indicate regressions slopes significant at the 95% confidence level based on a

580	Student t-test. Results for EP (resp. NE) GCM ensembles are indicated with blue (resp. green)
581	colors. Vertical red lines separate hist, NF, and FF periods. The horizontal black dashed line
582	indicates the 0 value. For the Western Europe blocking type. (d-f) Same as (a-c) for the Greenland
583	blocking type. (g-i) Same as (a-c) for the North Sea blocking type. (j-l) Same as (a-c) for the Baltic
584	Sea blocking type. (m-o) Same as (a-c) for the Scandinavia blocking type.
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Figure 4: (a) Composite time series of total frequencies of clustered blocking patterns for EP (blue)
and NE (GCMs). Vertical red lines separate hist, NF, and FF. (b) Trends in composite total
blocking frequencies over 36-year time frames calculated as linear regression slopes of blocking

595 frequencies against time for EP (blue) and NE (green) GCM simulations. Diamonds indicate 596 regressions slopes significant at the 95% confidence level based on a Student t-test. Vertical red 597 lines separate hist, NF, and FF periods. The horizontal black dashed line indicates the 0 value. 598

1	Winter Euro-Atlantic blocking activity less sensitive to climate change
2	than previously evaluated
3	
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7	
8	Abstract:
9	Winter Euro-Atlantic atmospheric blocking events have significant socioeconomical
10	impacts as they cause various types of weather extremes in a range of regions. According
11	to current climate projections, fewer of these blocking events will occur as temperatures
12	rise. However, the timing of such a reduction is currently highly uncertain. Meanwhile,
13	recent studies indicate that using climate models with high enough ocean resolutions to
14	simulate mesoscale eddies improve simulated winter Euro-Atlantic blocking
15	events significantly. In this paper, we show from a large ensemble of climate simulations
16	based on the highest emission scenario that largely prominent and coarsely resolved non-
17	eddying climate models project a noticeable significant decline in blocking frequencies
18	from the 2030s-2040s, whereas blocking statistics in eddy-permitting simulations are
19	noticeably decreasing only from years 2060s. Our result suggests with a strong level of
20	confidence that winter blocking activity over the next several decades will keep being
21	dominated by internal variability.

22 **1.** Introduction:

Atmospheric blocking is defined by large synoptic-scale pressure anomalies at midlatitudes that alter the midlatitude jet stream's pathways and redistribute precipitation and temperatures across many regions (Woolings et al. 2018). These pressure anomalies can last several days, thereby causing persistent weather extreme conditions at several locations (Kautz et al. 2022). The nature and amplitudes of underlying extremes depend on the timing of the blocking event (Davini and D'Andrea 2020), the location of blocked areas (Brunner et al. 2018), and background climate conditions (De Vries et al. 2013, Kautz et al. 2022, Woolings et al. 2018).

30

Winter blocking events occur frequently in the Euro-Atlantic sector and have a considerable impact on winter regimes across Europe (Cassou et al. 2004, Michel et al. 2023, Terray et al. 2005). These winter blocking conditions result in droughts (Sillman and Croci-Maspoli 2009), heavy precipitation (Kautz et al. 2022), extreme snowfalls (Cattiaux et al. 2011, de Vries et al. 2013), or cold spells (Cattiaux et al. 2010), all of which can have considerable impact on socioeconomic sectors such as energy or agriculture (Woolings et al. 2018).

37

38 Upper-level tropical warming and near-surface Arctic warming are expected to result in stronger 39 zonal winds and consequently in fewer Winter Euro-Atlantic Blocking (WEAB) events (Kennedy 40 et al. 2016, Woolings et al. 2018). A wide range of studies showed that most state-of-the-art 41 General Circulation Models (GCMs) of climate of the last three phases of the Coupled Model 42 Intercomparison Project (CMIP3, CMIP5 and CMIP6) simulate this decrease in WEAB 43 frequencies far before the end of the century, both for moderate and large emission scenarios 44 (Davini and D'Andrea 2020, Dunn-Singouin and Sun 2013, Fabiano et al. 2021, Kennedy et al. 45 2016, Masato et al. 2013, Matsueda and Endo 2017, Sillman and Croci-Maspoli 2009). However, 46 Bacer et al. (2022) recently showed no significant decrease in WEAB frequencies for half of the 47 simulations from six CMIP6 GCMs (selected for their relatively good performance in simulating

historical WEAB frequencies observations) by the end of the century and under the highest
emission scenario (SSP5-8.5). Also, CMIP5 GCMs generally do not simulate significant changes
in future structures of main winter Euro-Atlantic patterns (Huguenin et al. 2020). Woolings et al.
(2018) already emphasized that the large disparity in projected WEAB frequencies from GCMs
(Masato et al. 2013) and their generally poor ability to simulate observed WEAB events (Davini
and D'Andrea 2016), both cause large uncertainties on the occurrence, and if so, the timing of a
decline in WEAB frequencies under climate change.

55

56 The ability of GCMs to simulate WEAB is notably hampered by extensively documented persistent 57 biases, such as in simulated tropical convection (Gollan et al. 2019), transient eddy forcing 58 (Berckmans et al. 2013, Davini et al. 2017), model orography (Berckmans et al. 2013), or sea 59 surface temperatures (Athanasiadis et al. 2022, Michel et al. 2023, Scaife et al. 2011). Increased 60 atmospheric resolution was shown to improve the simulation of WEAB events notably through 61 reduced biases in simulated transient eddies (Davini et al. 2017, 2021) and improved orographic 62 resolution (Berckmans et al. 2013). Two recent studies illustrated that the bias reduction in 63 simulated sea surface temperature patterns achieved from the presence of mesoscale eddy-64 permitting (EP) ocean models also improves the simulation of WEAB events from GCMs 65 (Athanasiadis et al. 2022, Michel et al. 2023). These results were explained by reduced biases in 66 sea surface temperature gradients (Athanasiadis et al. 2022) shaped by better representation of 67 oceanic fronts (Hewitt et al. 2016, 2020, Michel et al. 2023, Scaife et al. 2011). This improves the 68 realism of simulated low level baroclinicity and vertical heat and moisture fluxes contributing to 69 blocking genesis (Cheung et al. 2023, O'Reilly et al, 2016). Earlier studies on future WEAB 70 behavior were predominantly based on GCM simulation ensembles mostly composed of less 71 accurate non-eddying (NE) GCMs, which are still highly prominent in CMIP6 (Eyring et al. 2016). Therefore, the present study seeks at refining projections of future WEAB activity in the context 72 73 of climate change by comparing two extensive sets of EP and NE GCM future simulations.

75 2. Data and Methods:

76 <u>2.1. Data:</u>

77

78 We employ up to 58 future simulations from 21 GCMs that participated to CMIP6 and High 79 resolution MIP (HighresMIP)/PRIMAVERA activities (Supplementary Table S1). We focus our 80 investigation only on simulations for the largest greenhouse gas emission experiments (i.e., 81 SSP5-8.5), as HighResMIP/PRIMAVERA future simulations were conducted only for this future 82 scenario (Haarsma et al. 2016). Therefore, 17 future simulations were taken from the highres-83 future experiment of the HighResMIP/PRIMAVERA activity, and 41 were taken from the SSP5-84 8.5 experiment of the Scenario MIP CMIP6 activity. We choose up to four simulation members per 85 GCM/experiment pair where available to account for the influence of the large internal variability 86 in simulated WEAB events (Supplementary Table S1).

87

88 It must be noted that EP GCMs are rare in CMIP6 future simulations but more common in 89 HighResMIP/PRIMAVERA future simulations, where the majority ends in 2050. As a result, future 90 GCM simulations will be studied for different periods: near future (NF, 2015-2050), far future (FF, 2051-2099), and the 21st century (21C,2015-2099), with 21C representing simulations from NF 91 92 that have been conducted until at least 2099 (Supplementary Table S1). In the end, NF is made 93 up of 16 EP simulations and 42 NE simulations (58 in total), while FF and 21C are made up of 7 94 EP simulations and 35 NE simulations (42 in total, Supplementary Table S1). It should be 95 emphasized that only a single NF simulation (HadGEM3-GC31-HH) was produced by a GCM with 96 an ocean resolution higher than 0.1°x0.1°, which is considered eddy-rich. However, because just 97 one of these simulations is being investigated here, it will be handled in the EP GCM group in the 98 section that follows.

Finally, we will compare the results of future simulations for NF and FF periods to those in presentday simulation outputs using 61 additional historical simulations from 1979 to 2014 based on fairly comparable GCM simulations ensembles (15 EP and 46 NE simulations, Supplementary Table S1). Ultimately, the present study is based on a very large ensemble of 119 simulations, both future (58) and historical (61) (Supplementary Table S1).

105

In the following, data from the different GCM simulations used (Supplementary Table S1) are
 initially interpolated on regularly separated 1°x1° grids.

- 108
- 109 <u>2.2. Methods:</u>

110 <u>2.2.1 Blocking detection</u>

We detect blocking events using on a long-established technique based on daily 500 hPa geopotential heights (Z500) fields providing a time-varying 2-dimensionnal blocking index (Sherrer et al. 2006). Other techniques, for instance based on potential vorticity fields or Z500 anomalies (Woolings et al. 2018) are more biased in the context of transient climate variations.

115

116 The method is based on detecting blocking situations for each grid point wherever a reversal in 117 the meridional geopotential heights gradient is observed. Therefore, we consider that a given grid 118 point with longitude λ_0 and latitude ϕ_0 is blocked at time *t* if the three following conditions C1 - C3119 are fulfilled:

120
$$C1: GHGS(\lambda_0, \phi_0) > 0$$

121
$$C2: GHGN(\lambda_0, \phi_0) < -10$$

122 $C3: GHGS_2(\lambda_0, \phi_0) < -5$

123 *GHGS*, *GHGN*, and *GHGS* $_2$ are given by:

124
$$GHGS(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_0) - Z500(\lambda_0, \phi_S)}{\phi_0 - \phi_S}$$

125
$$GHGN(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_N) - Z500(\lambda_0, \phi_0)}{\phi_N - \phi_0}$$

126
$$GHGN(\lambda_0, \phi_0) = \frac{Z500(\lambda_0, \phi_S) - Z500(\lambda_0, \phi_{S_2})}{\phi_S - \phi_{S_2}}$$

127 With $\phi_S = \phi_0 - 15^\circ$, $\phi_N = \phi_0 + 15^\circ$, and $\phi_{S_2} = \phi_0 - 30^\circ$. Therefore, for a given spatial field { Λ, Φ } 128 and for *n* daily timesteps, the 2D blocking index defined as:

129
$$M(\lambda, \phi, t) = \begin{cases} 1 \text{ if } C1 \cap C2 \cap C3 \\ 0 \text{ otherwise} \end{cases}, \lambda \in \Lambda, \phi \in \Phi, t = t_1, \dots, t_n \end{cases}$$

130 In addition, all grid points in { Λ , Φ } that are blocked for less than four days are turned to 0 to avoid 131 spurious detection of short-term, not impactful, and isolated blocking conditions across the grid 132 (Davini et al. 2012, Davini et al. 2017, Michel et al. 2023).

133

134 <u>2.2.2 Blocking clustering</u>

135 Clustering atmospheric blocking patterns is a common task that enables to study blocking activity 136 in terms of both frequencies and spatial patterns in an automatic way and has been used 137 frequently (Bacer et al. 2022, Cassou et al. 2004, Fabiano et al. 2021, Michelangeli et al. 1995, Michel et al. 2023, Terray et al. 2005). However, these studies differ with respect to the variable 138 139 that is used for clustering blocking patterns, namely: sea level pressure (Cassou et al. 2004, 140 Terray et al. 2005), geopotential heights at 700 hPa (Michelangeli et al. 1995), principal 141 components of Z500 (Fabiano et al. 2021), and 2D blocking index of blocking conditions longer 142 than 4 days (Michel et al. 2023). The clustering here consists of clustering days (winter ones here, 143 *i.e.* days from December, January, and February) based on their pattern similarities. This 144 approach is useful to study different types of blocking patterns and their respective frequencies, 145 as they were shown to have different and sometimes opposite meteorological fingerprints (Michel 146 et al. 2023). Using such an approach, for instance, Fabiano et al. (2021) evaluated, using a large 147 ensemble of CMIP5 and CMIP6 GCM simulations, that only specific blocking patterns were 148 significantly decreasing in the future as a result of rising temperatures. The clustering used here

149 (Michel et al. 2023) includes much less noisy signals than pressure variables (*i.e.* sea level 150 pressure and geopotential heights) used by other studies (Cassou et al. 2004, Fabiano et al. 151 2021, Michelangeli et al. 1995, Terray et al. 2005). For the k-means to be applied such that days 152 are clustered based on their blocking pattern similarities, the 3D matrix M is turned into a 2D 153 matrix denoted X, where rows are time steps and columns are all the grid points in { Λ , Φ } (Michel et al. 2023). First, k centroids $\mu_1^{(1)}, \dots, \mu_k^{(1)}$ for respective clusters $c_1^{(1)}, \dots, c_k^{(1)}$ are first randomly 154 155 drawn in the { Λ , Φ } space and blocking patterns for each time step are affiliated to the cluster with the closest centroid in terms of Euclidian distance. This procedure is repeated until convergence 156 157 (when centroids stop changing), where the update of centroids at each occurrence is given by:

158
$$\mu_j^{(t+1)} = \frac{1}{\#\{c_j^{(t)}\}} \sum_{i \in c_j^{(t)}} X_i, j = 1, \dots, k$$

159

160 Each k-means model in the study is ran 200 times with different random initializations of centroids. The set of random initial centroids with lowest Euclidian distances between cluster members and 161 162 their respective centroid is retained. The k parameter is generally chosen from reanalysis data 163 using optimization methods (Cassou et al. 2005, Michel et al. 2023). However, former studies did 164 not necessarily found or used similar values: k=4 (Cassou et al. 2005, Fabiano et al. 2020, 2021, 165 Terray et al. 2005), k=5 (Michel et al. 2023) or k=6 (Falkena et al. 2020) WEAB types. Here, as 166 we used the clustering based on the same variable as Michel et al. (2023), we use the parameter 167 k=5 and will refer to the five main patterns identified from ERA5 as: Western Europe (WE), 168 Greenland (Gr.), North Sea (NS), Baltic Sea (BS) and Scandinavia (Sc.).

169

170 In ERA5 data, this clustering for 5 main WEAB pattern is made with a k=6 parameter where only 171 the above five clustered patterns are kept. The 6th cluster constitute a "neutral" or "non-blocking" 172 group and accounts for about 72% of all DJF days in ERA5 between 1979 and 2014 (Michel et al. 2023). For GCMs, we attribute each daily blocking pattern to the closest centroid based on
Euclidian distances. In this procedure, the "neutral" group from the reanalysis clustering is also
used to allow the attribution of GCM blocking patterns to this category when they are not
resembling one of the other five.

177

178 **3.** <u>Results:</u>

179 Spatial distribution of projected WEAB frequencies and trends

180 Mean projected blocking frequencies averaged throughout the NF period (2015-2050) for both 181 NE and EP GCM ensembles (Supplementary Table S1) are shown in Fig. 1a,b, and significant 182 differences at the 90% confidence level between the ensembles are shown in Fig. 1c. It appears 183 that EP GCMs simulate a significantly higher number of blocking events by 2050, where 184 significant differences cover most regions of high blocking occurrence identified by previous studies, such as Greenland, Nordic Sea, or Scandinavia (Fig. 1a-c, Fabiano et al. 2021, Michel 185 186 et al. 2023, Terray et al. 2005). Furthermore, we see a wide area including the Baltic and Nordic 187 seas, British Islands, and the northeastern Atlantic where blocking frequency trends are notably 188 negative in the NE GCM ensemble (Fig. 1d). For EP GCMs, on the other hand, two tiny areas 189 with positive (Iceland) and negative (Greenland) trends are detected, but the fraction of grid points 190 with significant trends is lower than the 10% first species risk of the Student t-test used (7%, Fig. 191 1e). According to this preliminary analysis EP GCMs simulate more blocking events on average 192 by 2050 than NE GCMs (Fig. 1a-c) and also do not project any significant decrease in 193 frequencies until 2050 (Fig. 1d,e).

194

The increased simulated blocking frequencies found for EP GCMs over the historical (Michel et al. 2023) and NF periods (Fig. 1a-c) persist during the *21C* period (Fig. 1f-h). However, the geographical range of significant differences between the two groups is smaller than what we found for NF (Fig. 1c,h). In terms of trends, both model groups show significantly lower blocking frequencies throughout the *21C* period. These significant decreases are geographically consistent in the NE GCM simulation ensemble, but not in the EP ones, although most locations with high blocking prominence experience a significant decline (Fig. 1i-j). Supplementary Fig. S1 shows the same ensemble analyses as Fig. 1 for the historical simulations from Supplementary Table S1, where EP GCM WEAB frequencies are higher, consistent with Athanasiadis et al. 2022 and Michel et al. 2023, and also show significant decrease in NE simulations but not in EP simulations.

205 To further test the robustness of the above findings across GCMs, Monte Carlo tests and Student 206 t-tests based on area-averaged blocking frequencies and trends were performed over the study 207 area (50W-50E, 30-72.5N) for three GCM simulation periods: historical (1979-2014, hist), NF, and 208 21C (Supplementary Table S1, Fig. 2). For the three periods, the Monte-Carlo tests determine 209 confidence levels based on proportions of mean blocking frequencies from NE GCM 210 samples of same size as the smaller EP GCM ensembles (Fig. 2a). For hist and NF, it appears 211 that none of the 10,000 randomly drawn NE simulation samples have mean area-averaged 212 blocking frequencies that reach the one of the EP GCM simulations (Fig. 2a), which strongly 213 supports the results shown in Fig. 1. The significance level for 21C, on the other hand, is lower 214 but still higher than 95% (Fig. 2a), also consistent with differences seen between Fig. 1c and Fig. 215 1h.

216 For trends, Student t-tests show that both 95% and 99% confidence intervals for mean area-217 averaged trends from NE GCM simulations do not include 0 for all three periods, including hist 218 (Fig. 2b). On the other hand, EP GCM simulations only indicate significantly decreasing blocking 219 frequencies at the 99% confidence level when the entire 21st century is taken into account, but 220 trends are not significant for NF and hist. This result demonstrates that the response in WEAB 221 frequencies to strong anthropogenic forcing differs significantly between EP and NE simulations, 222 with EP GCMs simulating the same lower blocking frequencies far later in the century than NE 223 GCMs. Furthermore, as observed in historical simulations (Athanasiadis et al. 2022, Michel et al.

2023, Supplementary Fig. S1), EP GCMs simulate substantially higher and likely more realistic
WEAB frequencies for both NF and 21C periods (as they do so for historical, Michel et al. 2023),
implying that such GCMs are better suited to studying future winter blocking in Europe.

227

228 *Future evolution of observed main WEAB patterns*

In addition to the spatial analyses shown in Figs. 1-2, we clustered WEAB events in both historical
and future simulations into five primary blocking patterns, as performed in previous
studies (Cassou et al. 2004, Fabiano et al. 2021, Falkena et al. 2020, Michelangeli et al. 1994,
Michel et al. 2023, Terray et al. 2005).

233

We find no significant changes in the mean spatial extent of simulated blocking patterns over time, and neither EP nor NE GCM demonstrate any significant shift in the location of their assigned patterns (Fig. 3a,d,g,j,m). Indeed, for each blocking footprint, the simulated WEAB frequency tends to be more sensitive to NE and EP ocean configurations than to increasing temperatures, although the configuration effect is relatively small.

239

240 In terms of temporal evolution of these blocking patterns, we observe that the WEAB response to 241 anthropogenic forcing differs significantly over time, both between the NE and EP GCM 242 ensembles and between the simulated blocking patterns (Fig. 3b.e.h.k.n). Based on trends 243 evaluated across 36-year intervals (i.e., the same size as hist and NF periods), we find that 244 frequencies from a particular blocking type (i.e., Baltic Sea) do not show a globally consistent 245 decline over time for both NE and EP GCM ensembles (Fig. 3k,I). Greenland blocking declines 246 considerably more consistently in NE GCMs than in EP GCMs, where just a few 36-year slices 247 show significant decreases, although these are sufficient to significantly reduce Greenland 248 blocking occurrence by the end of the century (Fig. 3e,f). The opposite is true for Scandinavian 249 blocking, which has a significant decrease in frequency in EP GCM simulations late in the century 250 (from about 2060), but no globally significant decline in NE GCM simulations (Fig. 3n,o). Both NE 251 and EP GCM ensembles suggest a decline in frequencies after 2050 for the most prevalent 252 Western Europe blocking events (Michel et al. 2023), however the trend appears to be mitigated 253 at decadal to multidecadal time frames for EP GCMs (Fig. 3b,c). Furthermore, EP GCMs show a 254 time frame between 2000 and 2040 where Western Europe's blocking frequencies increase 255 significantly. Finally, WEAB frequencies over the Nordic Sea drop in both GCM groups, but the 256 timing differs considerably, occurring between 2015-2060 for NE GCMs and 2030-2085 for EP 257 GCMs.

258

259 Global view of future WEAB activity

260 Substantial disparities in simulated future WEAB frequencies exist between the NE and EP 261 GCMs, both geographically (Figures 1 and 2) and temporally (Figure 3). In Fig. 4, we present an 262 overall summary of results depicted above, where we show the total WEAB frequencies for EP 263 and NE ensembles (Fig. 4a) estimated as the sums of ensemble-means of timeseries from each 264 simulated blocking pattern for NE and EP GCMs, (Fig. 4b), as well as successive 36-year trends 265 of these timeseries. The magnitude of trends presented in Fig. 4b cannot be directly compared 266 between EP and NE simulation ensembles because they were computed from time series with 267 significantly different averages (Michel et al. 2023).

268

The assumption that EP GCMs have a smaller WEAB response to anthropogenic forcing than NE GCMs derived from spatially distributed trends (Figs. 1-2) and several of the clustered blocking patterns (Fig. 3) is confirmed here for the total frequencies for all blocking patterns (Fig. 4). Indeed, we find that total WEAB frequencies in NE GCMs start decreasing significantly (at the 95% confidence level) for 36-year time frames as early as 2000 (thus centered around 2018 and noticed around 2036) and continue to decrease significantly for following time frames until the end of the century (Fig. 4). The ensemble of EP GCM simulations tells a completely different
story, as 36-year trends in total WEAB frequencies start to be significantly and consistently
negative only for time frames centered around 2040 (thus noticed around 2060), representing a
more than 20-year delay compared to NE GCMS.

279

280 Beyond the significant downward trend in EP GCM WEAB frequencies found in the second half 281 of the 21st century (Fig. 4), the ensemble mean of these GCMs shows alternate decadal to 282 multidecadal periods with either significant or non-significant downward trends. This 283 overall results in an overall decrease in WEAB frequencies for EP GCMs but contrasts sharply 284 with the uninterruptedly significant negative trends observed over the entire 21st century for the 285 NE GCM ensemble. Hence, EP GCMs appear to be responding less to human-caused rising 286 temperatures (Fig. 1-4), and the role of decadal to multidecadal underlying processes such as 287 those associated with the North Atlantic/Arctic Oscillation are more important in shaping future 288 WEAB activity.

289

290 Conclusions and Discussion:

291 Our study was motivated by recent findings demonstrating higher accuracy in simulating WEAB 292 frequencies for EP GCMs compared to much more prevalent NE GCMs in the current CMIP6 293 generation (Athanasiadis et al. 2022, Michel et al. 2023), whereas EP GCMs were simply absent 294 in CMIP3 and CMIP5. Our results support those of most prior studies suggesting that the 295 occurrence of WEAB events will be significantly reduced by the end of the century (Davini and 296 D'Andrea 2020, Dunn-Singouin and Sun 2013, Fabiano et al. 2021, Kennedy et al. 2016, Masato 297 et al. 2013, Matsueda and Endo 2017, Sillman and Croci-Maspoli 2009). However, we discovered 298 significant differences in how NE and EP GCMs simulate the evolution of future WEAB events 299 under high human-induced forcing (Figs. 1-4). Indeed, while NE GCMs suggest that WEAB 300 frequencies may begin to decrease considerably over the next decade (Figs. 1, 4a), EP GCMs

301 suggest a delay of more than 20 years. These findings indicate that no significant decrease in the 302 occurrence of blocking-related winter extremes in Europe is expected in the coming decades. 303 Furthermore, we determined that EP GCMs simulate much more WEAB events in future 304 simulations than NE GCMs as was found for historical simulations (Michel et al. 2023). Given the 305 current need for a more precise picture on the future WEAB projections due to widely disparate 306 projections from NE GCMs (Fig. 2, Woolings et al. 2018), the use of EP GCM simulations in future 307 research is crucial.

308

In addition to the delayed response in decreased WEAB frequencies for EP GCMs, we also found that some 36-year long periods late in the century still experience no significant decrease (Fig. 4) whereas this is almost permanent in NE GCM simulations as early as from the 2030s. This result indicates a stronger role for internal variability in future WEAB occurrences in EP GCMs.

313

314 Due to the current low number of EP GCM simulations available in CMIP6 and HighResMIP/PRIMAVERA ensembles, we were unable to investigate other future simulations 315 316 such as those under intermediate (e.g., SSP2-4.5 and SSP3-7.0) or low (e.g., SSP1-2.6) 317 greenhouse gas emission scenarios. In this respect, Hausfather and Peters (2020) recalled that 318 the most widely used and widely communicated future scenario (i.e. SSP5-8.5) that we study here 319 is extremely useful for scientists attempting to understand the climate response to strong CO_2 320 forcings. However, its forcing is so large that it remains largely unlikely and unrealistic to 321 effectively happen in reality given current socioeconomic trends (Hausfather and Peters 2020). 322 As a result, the delayed decrease in WEAB frequencies found for EP GCMs over NE GCMs may 323 be even larger in the real-world, because future CO₂ emissions may be lower than those 324 prescribed for the GCM simulations investigated here (Hausfather and Peters 2020). Therefore, 325 it is likely that no significantly decreasing WEAB frequencies are expected to be noticed over the 326 next several decades, and internal variability may continue to primarily influence WEAB activity throughout this time (Woolings et al. 2018). However, a thorough examination of the response of
WEAB frequencies under more realistic emission scenarios in a large enough EP GCM simulation
ensemble has yet to be performed to corroborate the aforementioned statement. Current CMIP6
simulations do not yet allow for such an important analysis, and modeling centers handling EP
GCMs are encouraged to include simulations for intermediate emission scenarios in future CMIP
updates.

333 If we had limited our research to the NF simulation period (2015-2050), for which most EP GCM 334 simulations were done (Supplementary Table S1), we would have reached somewhat similar 335 findings as we do now, but with a much more incomplete picture. Indeed, downward trends in 336 WEAB frequencies in EP GCM simulations begin later than 2050, which would not have been 337 noticed if restricting the analysis to NF. This also reinforces the necessity for longer future 338 simulations using EP GCMs in future CMIP protocols, as these GCMs have been demonstrated 339 to simulate several climate processes far more correctly thans the currently widely used NE GCMs 340 (Athanasiadis et al. 2022, Delworth et al. 2012, Hewitt et al. 2016, 2020, Michel et al. 2023, 341 Moreno-Chamarro et al. 2021, Roberts et al. 2018).

342

345 **Open Research section**

346 *Data availability statement:*

347 Original CMIP6 (Eyring et al. 2016) and HighResMIP/PRIMAVERA (Haarsma et al.

348 2016) data are available through CEDA Earth System Grid Federation portals (https://esgf-

- 349 index1.ceda.ac.uk/search/cmip6-ceda/, https://esgf-index1.ceda.ac.uk/search/primavera-ceda/).
- 350 ERA5 reanalysis data (Hersbach et al. 2020) are available on the Copernicus Climate Change
- 351 Service (C3S) Climate Data Store (https://cds.climate.copernicus.eu/cdsapp#!/home). Data
- 352 produced through the study are available on a Zenodo repository with following DOI:
- 353 10.5281/zenodo.8090836.

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Figure 1: (a,b) Mean Winter Euro-Atlantic Blocking (WEAB) frequencies averaged over the near
future (NF) period (2015-2050) in 42 non-eddying (NE) and 16 eddy-permitting (EP) General
Circulation Model (GCM) future simulations under the highest greenhouse gas emissions scenario.
(c) Difference in mean WEAB frequencies between EP (b) and NE (a) GCM simulations. Green
colors indicates where WEAB frequencies are significantly higher in EP (resp. NE) GCM simulations
at the 90% confidence level based on a two-tailed Student t-test. White colors indicate where there
is no significant difference in WEAB frequencies from NE and EP GCMs based on the same test.

- 556 (d,e) Trends in percentage points per year in WEAB over the period under the highest greenhouse
- 557 gas emission scenario, for NE and EP GCMs respectively. White colors indicate where trends are
- 558 not significant at the 90% confidence level based on two-tailed regression slop Student t test. (f-j)
- 559 Same as (a-e) for the 21C period (2015-2099) in 35 NE and 7 EP GCM simulations.



562 Figure 2: Area-averaged WEAB frequencies (a) and trends (b) over the study area: 50W-50E, 30N-563 72.5N. Light, normal, and dark blue (resp. green) boxplots give WEAB frequencies EP (resp. NE) GCM 564 simulations for hist (1979-2014), NF (2015-2050) and 21C (2015-2100). Purple circles (resp. squares) 565 indicate averages for each boxplot EP (resp. NE) GCM simulations group. (a) Yellow and red squares 566 indicate 95% and 99% Monte-Carlo significance levels for EP GCM area averages. These levels are 567 determined for each period (hist, NF, 21C) by sampling 10,000 groups of NE GCMs with the same 568 number of GCM ensembles. (b) Yellow and red dashed lines indicate 95% and 99% Student 569 confidence interval for each boxplot. The horizontal black dashed line indicates the 0 value.



572 Figure 3: (a) ERA5 in-cluster blocking frequencies (colors) together with 20% contours of EP 573 (resp. NE) GCM ensembles given as light, normal, and dark green (resp. blue) lines for hist (1979-574 2014), NF (2015-2050) and FF (2051-2100) periods, respectively. For the Western Europe 575 blocking type. (b) Blue (resp. green) line indicates the composite time series of blocking frequency 576 available EP (NE) GCM simulations over hist, NF, and FF periods. Vertical red lines separate hist, 577 NF, and FF periods. For the Western Europe blocking type. (c) Trends in blocking frequencies 578 over 36-year time frames calculated as linear regression slopes of blocking frequencies against 579 time. Diamonds indicate regressions slopes significant at the 95% confidence level based on a

580	Student t-test. Results for EP (resp. NE) GCM ensembles are indicated with blue (resp. green)
581	colors. Vertical red lines separate hist, NF, and FF periods. The horizontal black dashed line
582	indicates the 0 value. For the Western Europe blocking type. (d-f) Same as (a-c) for the Greenland
583	blocking type. (g-i) Same as (a-c) for the North Sea blocking type. (j-l) Same as (a-c) for the Baltic
584	Sea blocking type. (m-o) Same as (a-c) for the Scandinavia blocking type.
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Figure 4: (a) Composite time series of total frequencies of clustered blocking patterns for EP (blue)
and NE (GCMs). Vertical red lines separate hist, NF, and FF. (b) Trends in composite total
blocking frequencies over 36-year time frames calculated as linear regression slopes of blocking

595 frequencies against time for EP (blue) and NE (green) GCM simulations. Diamonds indicate 596 regressions slopes significant at the 95% confidence level based on a Student t-test. Vertical red 597 lines separate hist, NF, and FF periods. The horizontal black dashed line indicates the 0 value. 598

Model name	EP/NE	historical	hist-1950	ssp585	highres-
	ensemble	(CMIP6)	(H/P)	(CMIP6)	future
		. ,			(H/P)
ACCESS-CM2	NE	r1i1p1f1*	Х	r1i1p1f1*	X
		r4i1p1f1*		r4i1p1f1*	
		r5i1p1f1*		r5i1p1f1*	
CanESMM5	NE	r1i1p1f1	Х	r1i1p2f1*	Х
		r2i1p2f1		r2i1p1f1*	
		r3i1p2f1		r3i1p2f1*	
		r4i1p2f1		r4i1p2f1*	
CESM2	NE	r4i1p1f1	Х	r4i1p1f1	Х
		r10i1p1f1		r10i1p1f1	
		r11i1n1f1		r11i1p1f1	
CNRM-CM6-1	NE	r1i1n1f2	X	r1i1n1f2*	r1i1n1f2
		r2i1n1f2	~	r2i1n1f2*	mpnz
		r3i1n1f2		r3i1p1f2*	
		r/i1n1f2		15110112	
	ED	r1i1p112	Y	r1i1n1f2*	r1i1n1f2
	LF	mpnz	~	THIPHZ	impiiz
EC Earth 3P		Y	r1i1n2f1	v	r1i1n2f1
EC-Earti-SF		~	r2i1p2f1	~	r2i1p2f1
			r2i1p2f1		r2i1p2f1
EC Earth 2D UD	ED		r1i1p2f1		r1i1p2f1
EC-Eartin-SF-FIK	EF		r0:1=0f1		r111p211
			1211p211		1211µ211 #2:1#261*
		V	r311p211	V	
ECIVIVYF-IF5-FIR	EP	~	r111p111	×	~
			1211p111 r2:1 p 1f1		
		V	1311p111 #4:1#4f4	V	V
ECMVVF-IFS-LR	INE	X	r111p1f1	×	X
			r211p111		
		V	1311p111 #1:1 p 1f1	V	V
ECIVIVE-IES-IVIR	EP	~	r111p111	×	~
			r211p111		
					V
GFDL-CM4	EP	riipiii	~	riipiii	~
		v	v	v	r1;1p1f1
HadGEMI3-GC31-HH	EP	~	~	×	гнрнг
		V	V	V	r1;1,0,1f1
	EP	^	^	^	r110-111
					r1i2p111
	NE	r1;1n1f2	v	r1;1n1f2*	v Thispitt
		r);10140	^	r2i101f2*	^
		12110113		12110113	
		13110113		1311p113"	
		1411p113	v	14110113 r1:1r 1f0*	r1;1 r 1£1
		1111p113	· ·	1111p113"	r1:0=1+1
		1211p113		1211p113"	1112p111
		r3/1p113		r311p1f3^	гнэртт
		r4i1p1t3		r4i1p1f3*	

INM-CM5-0	NE	r1i1p1f1	X	r1i1p1f1*	Х
IPSL-CM6A-LR	NE	r1i1p1f1 r2i1p1f1 r31p1f1 r4i1p1f1	X	r1i1p1f1 r2i1p1f1 r31p1f1 r4i1p1f1	Х
MIROC6	NE	r1i1p1f1 r2i1p1f1 r3i1p1f1	X	r1i1p1f1 r2i1p1f1 r3i1p1f1	Х
MPI-ESM1-2-HR	NE	r1i1p1f1 r2i1p1f1 r3i1p1f1	X	r1i1p1f1 r2i1p1f1	r1i1p1f1
MPI-ESM1-2-LR	NE	r1i1p1f1 r2i1p1f1 r3i1p1f1 r4i1p1f1	X	r1i1p1f1 r2i1p1f1 r3i1p1f1 r4i1p1f1	Х
MPI-ESM1-2-XR	NE	X	r1i1p1f1	X	r1i1p1f1
NorESM2-MM	NE	r1i1p1f1	X	r1i1p1f1	Х
TaiESM1	NE	r1i1p1f1	Х	r1i1p1f1	Х
UKESM1-0-LL	NE	r1i1p1f2 r2i1p1f2 r3i1p1f2 r4i1p1f2	X	r1i1p1f2 r2i1p1f2 r3i1p1f2 r4i1p1f2	Х

*: present for *21C* period (2015-2099); EP: Eddy-Permitting; NE:Non-Eddying; H/P: HighResMIP/PRIMAVERA

Supplementary Table S1: Table summarizing the ensemble of 119 simulations (61 for historical, and 58 for future experiments). For ocean grid configurations, "NE" indicates climate models with a non-eddying ocean component's resolution, and "EP" indicates climate models with an eddy-permitting ocean component's resolution. For each model/experiment, the name of members used in the study are indicated. Stars indicate climate model simulations that were investigated over the 21st century (*21C*) period (2015-2099).