No emergence of deep convection in the Arctic Ocean across CMIP6 models

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September 25, 2023

Abstract

As sea ice disappears, the emergence of open ocean deep convection in the Arctic has been suggested. Here, using 36 state-ofthe-art climate models and up to 50 ensemble members per model, we show that Arctic deep convection is rare even under the strongest warming scenario. Only 5 models have somewhat permanent convection by 2100, while 11 have had convection by the middle of the run. For all, the deepest mixed layers are in the Eurasian basin, by St Anna Trough. When the models convect, that region undergoes a salinification and increasing wind speeds; it is freshening otherwise. We discuss the causality and potential reasons for the opposite trends. Given the model's different parameterisations, and given that the ensemble members that convect the deepest, most often, are those with the strongest sensitivity, we conclude that differences in deep convection are most likely linked to the model formulation.

No emergence of deep convection in the Arctic Ocean across CMIP6 models

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Key Points:

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7	•	Oceanic deep convection does not emerge and persist in the Arctic in the major-
8		ity of CMIP6 models, despite a cessation in the Nordic Seas.
9	•	Arctic deep convection occurs only when both surface salinity and winds are in-
10		creasing, year round, yet most models are freshening.
11	•	The models with the strongest sensitivity, especially with an oceanic polar am-
12		plification, have the deepest Arctic mixed layers, most often.

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13 Abstract

As sea ice disappears, the emergence of open ocean deep convection in the Arctic has 14 been suggested. Here, using 36 state-of-the-art climate models and up to 50 ensemble 15 members per model, we show that Arctic deep convection is rare even under the strongest 16 warming scenario. Only 5 models have somewhat permanent convection by 2100, while 17 11 have had convection by the middle of the run. For all, the deepest mixed layers are 18 in the Eurasian basin, by St Anna Trough. When the models convect, that region un-19 dergoes a salinification and increasing wind speeds; it is freshening otherwise. We dis-20 cuss the causality and potential reasons for the opposite trends. Given the model's dif-21 ferent parameterisations, and given that the ensemble members that convect the deep-22 est, most often, are those with the strongest sensitivity, we conclude that differences in 23 deep convection are most likely linked to the model formulation. 24

²⁵ Plain Language Summary

Both observations and modelling simulations suggest that deep vertical mixing (or 26 deep convection) in winter may become the new normal in the Arctic as sea ice disap-27 pears. These simulations are often done using only one model, so here we used all mod-28 els available that participated in the Climate Model Intercomparison Project phase 6, 29 for the strongest warming scenario. We show that after removing those that are already 30 inaccurate in the present, and even with a restrictive threshold, most models have no deep 31 32 convection in the Arctic, or extremely rarely. Only 5 still had deep convection by the time the run finishes in 2100. We investigated the possible links between deep convec-33 tion and surface salinity, surface temperature, sea ice concentration and surface wind speeds, 34 and found that the salinity was most important. Deep convection regions and periods 35 are associated with a saltier, windier surface, while the rest of the Arctic and/or rest of 36 the run freshens. Causality is unclear; we need higher resolution than monthly output. 37 Similar behaviours within model families, a strong link to the model sensitivity, and cited 38 work make us conclude that ultimately, the differences are probably caused primarily by 39 the different model designs. 40

41 **1** Introduction

The Arctic Ocean is changing. The resulting reduction in Arctic sea ice extent and 42 thickness (Mallett et al., 2021; Meier & Stroeve, 2022) could enhance vertical mixing: 43 More brine may be rejected year-round as younger, saltier ice desalinates (Peterson, 2018) 44 or as sea ice reforms in winter over the now seasonal-ice areas (Onarheim et al., 2018), 45 while the ice-freed regions may become more susceptible to wind stirring (Timmermans 46 & Marshall, 2020). In the Eurasian Arctic, the process known as Atlantification (Polyakov 47 et al., 2017), whereby warm water of Atlantic origin penetrates further into the Arctic, 48 may be further weakening the stratification and enhancing sea ice melt. These led Polyakov 49 et al. (2017) to hypothesise that the Arctic may start exhibiting deep convection in win-50 ter, a result found by Lique et al. (2018) in the 4x CO₂ scenario using the model HiGEM. 51

However, this hypothesis is so far not confirmed. Peralta-Ferriz and Woodgate (2015) 52 found that a deepening of mixed layers in the Arctic is unlikely, since stratification greatly 53 dominates over the wind effect. The latest observations in the Eurasian Arctic (Schulz 54 et al., under review) yielded mixed layers no deeper than 130 m, even in winter. Besides, 55 in a recent Arctic study using models that participated in the Climate Model Intercom-56 parison Project phase 6 or CMIP6 (Eyring et al., 2016), Muilwijk et al. (2023) showed 57 that there was no agreement among models regarding future stratification and the ef-58 fect of Atlantification under the strongest warming scenario (SSP5-8.5, O'Neill et al. (2016)). 59 This suggests that HiGEM's deep convection in the Arctic may be a model artefact rather 60 than the future of the Arctic. 61

We here determine whether deep convection emerges in the Arctic in the future scenario SSP5-8.5 using all CMIP6 models and all their ensemble members for which the mixed layer depth output was available, as described in section 2. In section 3, we detail the spatial and temporal patterns of future Arctic mixed layers and discuss possible reasons for these, focusing on the model biases and their trends in surface properties, sea ice and winds. We conclude in section 4.

⁶⁸ 2 Data and Methods

2.1 CMIP6 data

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To investigate deep convection and its potential drivers in the future Arctic, we use all CMIP6 models and all their ensemble members that had monthly mixed layer depth ("mlotst"), ocean surface salinity ("sos") and temperature ("tos"), sea ice concentration ("siconc") and surface wind speed ("sfcWind") available on any of the Earth System Grid Federation (ESGF) nodes for the future scenario SSP5-8.5, for January 2015 to December 2100. The models and their ensemble members are listed in supp. Tables S1 and S2.

For AWI-CM-1-1-MR and GISS-E2-1-G, we used the sea ice thickness ("sivol") be-76 cause sea ice concentration was not available. For CAMS-CSM1-0, we generated the sur-77 face salinity from the full-depth salinity ("so") as the former was not available; we used 78 the salinity of the shallowest level, 5 m depth. Similarly, we used the models' bathymetry 79 ("deptho") when it was available, but had to generate it from the full-depth salinity for 80 11 models, as the last level with salinity data. Finally, for CESM2-WACCM, GFDL-CM4 81 and MRI-ESM2-0, several grid types were available and we chose for simplicity the reg-82 ularised grid ("gr"). For the other models, we took the one grid type available; see supp. 83 Tables S1 and S2. 84

2.2 Methods

The thresholds and choices of this subsection are discussed in supplementary text 86 S1. In agreement with Lique et al. (2018), we consider that there is deep convection in 87 the Arctic if the mixed layer depth (MLD) exceeds 500 m. We here do not quantify over-88 turned volumes. We only perform a binary detection of deep convection, so we select the 89 overall maximum MLD, in space or time depending on the analysis. The ensemble mem-90 bers that exhibit deep convection in the Arctic over the observed part of SSP5-8.5 (2015-91 2023) are shown on the first figure and subsequently removed from the study, as they 92 are already inaccurate at the beginning of the run. Consequently, of the originally 36 mod-93 els, 27 remain for most of the analysis. We do not take biases in Nordic Seas MLD into 94 account for model selection, as all CMIP6 models have spurious deep convection there 95 (Heuzé, 2021). 96

The sea ice edge (supp. Fig S1) is detected as the contour of 15% concentration or 10 cm thickness, averaged over 2040-2060 and 2080-2100. We define the Eurasian basin as the region north of 80°N, longitudes 20°W - 140°E, deeper than 1000 m, and the Nordic Seas as the region of latitudes 66 - 80°N, longitudes 30°W - 20°, deeper than 1000 m see contours on Fig 1. We perform across-model correlations by comparing, for each model and each ensemble member, their maximum MLD and number of year where the MLD exceeds 500 m in the Eurasian basin, to:

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- Average surface temperature and salinity in the Eurasian basin over the first 20 years of SSP5-8.5;
- Tropical and Mid-latitude (60°S 60°N) and Arctic (north of 75°N) warming, as
 difference between the 2080-2100 and 2015-2035 average ocean surface tempera ture, which we use as a proxy for the model sensitivity. We also define the oceanic

polar amplification as the difference between the Arctic and tropical/mid-latitudewarming.

Besides, we compute correlations and trends for each grid cell, after interpolating 111 all the parameters onto each model's mlotst grid. The correlations and trends are com-112 puted for each ensemble member separately, and averaged afterwards. We consider that 113 there is ensemble agreement if more than 50% of the ensemble members have a signif-114 icant correlation/trend (determined using a t-test at 95% significance) of the same sign; 115 we then present the median value and its ensemble-spread. Finally, we present the com-116 posite trends after grouping the models based on their Arctic deep convection behaviour, 117 as described in the result section. 118

¹¹⁹ **3** Results and Discussion

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3.1 Arctic deep convection is rare, restricted both in space and time

The maximum mixed layer depth reached in the Arctic over 2015-2100 varies strongly 121 across models (Fig. 1, note the logarithmic scale). The value does not exceed 100 m for 122 some, such as CAMS-CSM1-0, while others such as CMCC-CM2-SR5 exceed 2000 m in 123 the majority of the Eurasian basin. All models with deep MLD agree that the deepest 124 values are in the Eurasian basin, most commonly by St Anna Trough. Nine models have 125 MLD exceeding the 500 m threshold for deep convection already over 2015 to present 126 for all their ensemble members (red squares on Fig. 1). Interestingly, these are not only 127 by St Anna Trough but also north of Svalbard, suggesting that for these models the Nordic 128 Sea deep convection area extends too far north, most likely following the sea ice edge (shown 129 on supp. Fig. S1). All models presented on Fig. 1 have spuriously deep MLD in the Nordic 130 Seas. 131

After removing the models and ensemble members that are unrealistic in the presentday regarding their Arctic MLD (asterisks in supp Tables S1 and S2), 27 models remain. The temporal evolution of their MLD reveals four groups of models (Fig. 2 and supp Table S3):

136	1.	The first 11 models have no Arctic deep convection during the entire run. The Arc-
137		tic MLD time series (blue lines, Fig. 2) are mostly flat, with no year where the
138		MLD exceeds 500 m, regardless of the ensemble member. The maximum MLD across
139		these models and their ensemble members is often of the order of 100 m, i.e. like
140		currently observed in the Arctic (Schulz et al., under review).
141	2.	Six models have deep convection in the Arctic on rare occasions, in the middle of
142		the run. The maximum number of years with deep convection for this group is 17
143		out of 86 (supp Table S3), but is most often 4 or fewer.
144	3.	Five models also have deep convection in the Arctic in the middle of the run, more
145		often. It starts by 2030, peaks in the first half of the run, and then declines slowly,
146		with the ensemble average back under the 500 m threshold by the end of the run.
147	4.	The last 5 models start convecting in the second half of the run, by 2070, and ap-
148		pear "stably" convecting at the end of the run. Unfortunately, only one ensem-
149		ble member of one of the models (ACCESS-CM2) is available beyond 2100, so we
150		cannot tell whether deep convection in these models would also decline later.

The Nordic Seas MLD (red lines, Fig. 2) falls below 1000 m for all models, and even below 500 m for two thirds of them. The consistent cessation of Nordic Seas deep convection is not related to the models' behaviour in the Arctic. Therefore, unlike suggested by Lique et al. (2018), deep convection does not migrate to the Arctic in response to its cessation further south.



Figure 1. Maximum MLD over January 2015 - December 2100 (shading, logarithmic scale) for SSP5-8.5 at each grid cell. For each model, parentheses indicate the number of ensemble members available (see supp. Tables S1 and S2); when this number is larger than one, the figure shows the ensemble median. Black contours are the 1000 m isobath. Red squares indicate the models for which all ensemble members have deep convection in the Arctic already over 2015-2023, and which are therefore not considered for further analysis. Locations discussed in the manuscript are indicated on the top-right panel: Cyan contour "EB" is the Eurasian Basin, indigo dashed contour "GIN" is the Nordic Seas, and magenta arrow "SAT" is the St Anna Trough.



Figure 2. For the 27 models that do not have deep convection in the Arctic over 2015-2023, time series of their yearly maximum MLD in the Nordic Seas (red) and in the Arctic Ocean (blue). For each model, parentheses indicate the number of ensemble members remaining; when this number is larger than one, the figure shows the range across these ensemble members (shading) and the ensemble median (thick line). Horizontal black lines indicate the 500 and 1000 m MLD thresholds, indicative of deep convection. Vertical black line is the year 2023. Models are ordered based on their Arctic behaviour: First two rows and NorESM2-LM, no Arctic deep convection; ACCESS-ESM1-5 and fourth row, rare convection, by the middle of the run; fifth row, convection peaks by the middle of the run and then declines; bottom row, convection starts late in the run.

What causes these different behaviours then? Ensemble members usually have a consistent behaviour (the shading on Fig. 2 usually agrees with the thick line), and models of the same family tend to belong to the same group. The only exception are the two ACCESS models (Fig. 2), but there are large differences in their designs and implemented schemes of relevance for polar regions and deep convection in particular, as discussed in Mohrmann et al. (2021). In the next section, we investigate in more details what could be causing the different Arctic deep convection behaviours, starting with model designs.

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3.2 Potential causes: High sensitivity, salinification, stronger winds

The models that have deep convection in the Arctic usually have it at the same location (Fig. 1), but the temporal evolution of the mixed layer yields four groups of models (Fig. 2). Models of the same family tend to belong to the same group; these usually have similar biases and sensitivities, so we first investigate these.

The mean temperature in the Eurasian Basin at the beginning of the SSP5-8.5 run 168 and subsequent MLD are positively correlated across models (Fig. 3, left): the warmer 169 at the beginning of the run, the deeper the mixed layers (correlation 0.40), the more of-170 ten (0.33). The same results are obtained when considering the ensemble members sep-171 arately, as on Fig. 3, and the ensemble mean (not shown). There is no such relationship 172 with salinity. The relationship with temperature persists throughout the run, so that the 173 models that warm the most (Fig. 3, centre), and quite strikingly, those for which the ocean 174 surface warms more in the Arctic than in the midlatitudes (Fig. 3, right) are the ones 175 with the deepest MLD, the most often. The models belonging to the E3SM project (cyan 176 contour, Fig. 3) are the exception: They warm strongly but convect rarely or not at all, 177 which could be because their design is very different from that of the other CMIP mod-178 els (Golaz et al., 2019). 179



Figure 3. For each model and each ensemble member that were not removed from the study, scatter plot between the maximum MLD over 2015-2100, over the Eurasian Basin, and: Left, the mean ocean surface temperature over the Eurasian basin, over the first 20 years of the run; Centre, the Arctic warming, i.e. the mean ocean surface over the last 20 years of the run minus that of the first 20 years; Right, that Arctic warming minus that of the tropical to mid-latitudes, which we call "Oceanic Polar Amplification". See Methods for latitude definitions. Colors indicate the number of years where the Arctic MLD exceeds the 500 m threshold, as per supp. Table S3. Cyan contours highlight the E3SM project models (see text).

Although the salinity at the beginning of the run has no relationship with the max-180 imum MLD reached, salinity variations are strongly correlated to MLD variations (Ta-181 ble 1) and the correlation differs depending on the Arctic deep convection behaviour of 182 the models. Most models that have no deep convection in the Arctic or rarely (first two 183 blocks) have a positive correlation between their March MLD and the surface salinity 184 one month before, and a negative correlation with the ocean surface temperature: These 185 models have shallower MLD when they are fresher and warmer (as expected in a chang-186 ing Arctic, e.g. Peralta-Ferriz and Woodgate (2015)). In the two groups of convecting 187 models, especially so in the convective region (lines "DC", Table 1), the correlation is 188 positive with salinity and temperature: deeper MLD are associated with salier and warmer 189 surface waters the month prior. Of the four drivers investigated, the MLD is most strongly 190 correlated with salinity or temperature for the vast majority of models (bold fonts on 191 Table 1); usually it is strongest with the salinity for the non or rarely convecting mod-192 els, and with the temperature for the convecting ones (as expected from the previous para-193 graph). Correlations are similar when considering possible drivers the summer before (not 194 shown, see also supp. Text S1). Unsurprisingly, the possible drivers are not independent 195 (supp. Table S4). The correlations between temperature and salinity, and between salin-196 ity and either sea ice concentration or wind speed are of different signs depending on the 197 convecting behaviour, suggesting that different processes and/or water masses are involved 198 (see next subsection). 199

From the correlation analysis, we suspect that the models with no or rare deep convection may be freshening, while those with convection may become saltier. A trend analysis confirms this hypothesis:

- The models with no deep convection in the Arctic become fresher throughout the Arctic, throughout the run (Fig. 4, first two lines). Their trends are rather weak compared to the other model groups. Their winter sea ice does not retreat far into the Arctic, even by the end of the run (supp. Fig. S1).
- 2. The models with deep convection at the middle of the run, be it rarely (Fig. 4, 207 lines 3 and 4) or peaking and declining (lines 5 and 6) exhibit similar trends. Their 208 ocean surface becomes saltier and winds stronger at the location where mixed lay-209 ers deepen in the first half of the run; they freshen in the rest of the Arctic. In the 210 second half, when mixed layers are shallow again, the ocean surface freshens ev-211 erywhere. The main difference between these two groups is in their sea ice trends, 212 with the models whose convection peaks and slowly declines having no winter sea 213 ice by the end of the run (supp. Fig. S1). 214
- 3. The models that convect at the end of the run (Fig. 4, last two lines) have the opposite salinity trends: first a freshening, then a salinification in the region where mixed layers deepen, along with stronger winds.

Note that the trend patterns are similar in summer (supp. Figure S2), indicating that
the changes occur year round. The conclusion is that Arctic deep convection is associated with both a saltier ocean surface and stronger winds. Stronger winds alone are not
enough (see e.g. the no deep convection group, second half of the run).

3.3 On causality

Deep convection in the Nordic Seas ceases for all CMIP6 models, but rarely emerges as a stable feature in the Arctic. In the Arctic, we find that deep convection is associated with a saltier ocean surface and stronger wind speeds, and is most intense and durable in the models with the strongest sensitivity, especially so if the Arctic warms more than the mid-latitudes. But are the trends we observe the causes or consequences of deep convection?

Table 1. Correlation coefficient and its standard deviation for each model between the March MLD and the previous February ocean surface salinity (S), ocean surface temperature (T), sea ice concentration (SIC) and surface wind speed (Wind), for the Eurasian basin (EB) and, for the models with Arctic deep convection, where the maximum MLD of Fig. 1 exceeds 500 m (DC). Models are ordered based on their Arctic deep convection behaviour, as per Fig. 2. Only correlations significant at 95% are shown; for models with more than one ensemble member, median of the correlations of the dominating sign. For models with more than one ensemble member, standard deviation is the across-ensemble spread; spatial spread otherwise. Bold fonts highlight the maximum correlation for each model. Correlation between parameters is shown in supp. Table S4.

Model	Region	MLD vs S	MLD vs T	MLD vs SIC	MLD vs Wind
BCC-CSM2-MR	EB	0.47 ± 0.15	-0.41 ± 0.17	-0.28 ± 0.35	0.23 ± 0.24
CAMS-CSM1-0	\mathbf{EB}	0.54 ± 0.02	-0.49 ± 0.00	-0.35 ± 0.03	0.29 ± 0.04
CAS-ESM2-0	\mathbf{EB}	$\textbf{0.59} \pm 0.04$	-0.49 ± 0.06	0.51 ± 0.06	-0.24 ± 0.01
CESM2-FV2	\mathbf{EB}	-0.29 ± 0.37	0.22 ± 0.37	-0.23 ± 0.16	0.31 ± 0.08
CESM2-WACCM	\mathbf{EB}	0.34 ± 0.06	-0.31 ± 0.05	0.25 ± 0.06	0.30 ± 0.02
CESM2	\mathbf{EB}	0.32 ± 0.08	-0.30 ± 0.04	0.30 ± 0.05	0.32 ± 0.02
E3SM-1-1-ECA	\mathbf{EB}	0.43 ± 0.14	-0.45 ± 0.47	$\textbf{0.60} \pm 0.41$	-0.31 ± 0.31
FGOALS-f3-L	\mathbf{EB}	0.51 ± 0.21	-0.51 ± 0.10	0.59 ± 0.13	-0.26 ± 0.16
FGOALS-g3	\mathbf{EB}	$\textbf{0.89} \pm 0.01$	-0.28 ± 0.02	0.46 ± 0.04	0.28 ± 0.04
GFDL-CM4	\mathbf{EB}	$\textbf{0.39} \pm 0.31$	-0.25 ± 0.39	-0.31 ± 0.29	-0.27 ± 0.19
NorESM2-LM	EB	-0.28 ± 0.36	$\textbf{0.36}\pm0.31$	-0.26 ± 0.22	0.26 ± 0.16
ACCESS-ESM1-5	EB	-	-	-	-
	DC	-	-	-	-
E3SM-1-0	EB	0.38 ± 0.01	-0.50 ± 0.12	0.51 ± 0.05	-0.26 ± 0.04
	DC	-	-	-	-
E3SM-1-1	EB	0.39 ± 0.14	-0.41 ± 0.27	0.55 ± 0.28	-0.28 ± 0.27
	DC	0.51 ± 0.01	0.34 ± 0.02	-0.29 ± 0.01	0.43 ± 0.01
GISS-E2-1-G	\mathbf{EB}	0.61 ± 0.03	-0.59 ± 0.04	0.35 ± 0.08	0.27 ± 0.04
	DC	0.48 ± 0.14	0.67 ± 0.10	-0.37 ± 0.08	0.32 ± 0.09
MRI-ESM2-0	\mathbf{EB}	0.62 ± 0.03	-0.40 ± 0.04	0.29 ± 0.05	0.31 ± 0.04
	DC	-	-	-	-
NESM3	\mathbf{EB}	$\textbf{-0.31}\pm0.01$	-0.24 ± 0.01	-0.24 ± 0.01	-0.23 ± 0.02
	DC	-	-	-	-
CMCC-ESM2	\mathbf{EB}	$\textbf{0.59} \pm 0.45$	0.45 ± 0.37	-0.31 ± 0.43	0.28 ± 0.25
	DC	$\textbf{0.65}\pm0.11$	0.49 ± 0.14	-0.40 ± 0.12	0.29 ± 0.08
CanESM5-1	EB	0.50 ± 0.03	0.39 ± 0.11	-0.46 ± 0.08	0.36 ± 0.05
	DC	0.52 ± 0.04	0.39 ± 0.12	-0.48 ± 0.08	0.36 ± 0.05
CanESM5	\mathbf{EB}	-	-	-	-
	DC	-	-	-	-
HadGEM3-GC31-LL	\mathbf{EB}	-	-	-	-
	DC	-	-	-	-
UKESM1-0-LL	\mathbf{EB}	0.56 ± 0.02	0.64 ± 0.01	-0.51 ± 0.02	0.43 ± 0.05
	DC	0.57 ± 0.02	0.65 ± 0.02	-0.54 ± 0.02	0.43 ± 0.06
ACCESS-CM2	\mathbf{EB}	0.40 ± 0.05	0.45 ± 0.19	-0.28 \pm 0.08	0.28 ± 0.03
	DC	0.44 ± 0.05	$\textbf{0.76}\pm0.11$	-0.51 \pm 0.08	0.34 ± 0.04
CNRM-CM6-1	\mathbf{EB}	0.40 ± 0.07	0.73 ± 0.05	-0.62 ± 0.05	0.38 ± 0.05
	DC	0.46 ± 0.05	0.77 ± 0.08	-0.68 ± 0.05	0.47 ± 0.04
CNRM-ESM2-1	\mathbf{EB}	0.44 ± 0.15	$\textbf{0.77} \pm 0.48$	$\textbf{-}0.53\pm0.31$	0.43 ± 0.13
	DC	0.46 ± 0.09	$\textbf{0.86}\pm0.10$	$\textbf{-}0.57\pm0.16$	0.47 ± 0.12
EC-Earth3-Veg-LR	\mathbf{EB}	0.44 ± 0.04	0.85 ± 0.05	-0.55 \pm 0.12	0.35 ± 0.05
	DC	0.45 ± 0.03	0.90 ± 0.05	-0.66 ± 0.04	0.39 ± 0.03
EC-Earth3-Veg	EB	0.47 ± 0.13	$\textbf{0.85} \pm 0.43$	$\textbf{-}0.49\pm0.17$	0.32 ± 0.09
	DC	0.52 ± 0.10	$\textbf{0.89} \pm 0.07$	-0.53 ± 0.12	0.34 ± 0.10



Figure 4. Composite trends based on the models' Arctic deep convection behaviours of Fig. 2, for the first half of the 21st century run (top rows) and the second half (bottom rows), in March MLD (first column), and February ocean surface salinity (S, second), ocean surface temperature (T, third), sea ice concentration (fourth) and surface wind speed (last). Behaviours are described to the left of the figure, and number of models for each behaviour is given in parentheses. For each panel, stippling indicates non-significant trends and/or model disagreement regarding the trend's sign. Straight stippled lines across the North Pole are an artefact of the necessary interpolation. -10-

The Arctic Ocean surface warming and sea ice loss trends, year-round, are to be 229 expected in a warming world (IPCC, 2019). Besides, they are consistent across-models, 230 regardless of their convecting behaviour in the Arctic. Similarly, the models' sensitiv-231 ities have been attributed to their representation of cloud cover and cloud albedo (Zelinka 232 et al., 2020), which locally can be affected by the heat and moisture fluxes from deep con-233 vection (Monroe et al., 2021), but in coarse models is more likely due to each model's 234 cloud parameterisation (Zelinka et al., 2020). Finally, changes in both "normal" winds 235 (Screen et al., 2018) and cyclones (Rinke et al., 2017) are most commonly attributed to 236 the large scale atmospheric circulation, although local changes at the boundary layer be-237 cause of sea ice loss may accelerate winds (DuVivier et al., accepted). 238

As for the salinity, Lique et al. (2018) attributed the freshening - salinification dipoles 239 to changes in the large scale oceanic circulation, whereas Davis et al. (2016) attributed 240 the increased salinity to increased vertical mixing. That is, for the former it drives Arc-241 tic deep convection; for the latter, it is a consequence of it. A local change in salinity can 242 be also caused by enhanced sea ice formation, but we find a negative correlation between 243 winter salinity and sea ice concentration among the convecting models (supp. Table S4), 244 which makes this causality unlikely given that the sea ice maps do not seem to exhibit 245 polynyas (supp. Fig. S1). Salinification taking place year-round, at the same location 246 as winds are increasing, would instead suggest an enhanced sea ice drift, but the detailed 247 sea ice mass budget of Keen et al. (2021) shows the opposite. 248

Alternatively, the impact of deeper MLD on the Arctic surface salinity will depend 249 on each model's Atlantic layer. Heuzé et al. (2023) showed strong biases in that layer, 250 with most models having it too deep. Khosravi et al. (2022) further showed that there 251 is no consistency across CMIP6 models regarding the changes of that Atlantic layer dur-252 ing SSP5-8.5, which Muilwijk et al. (2023) linked to the models' lack of consensus re-253 garding future changes in stratification in the Eurasian basin and in Atlantification. As 254 stratification and deep convection are intimately linked, this is another feedback that makes 255 the causality uncertain. One can unfortunately not study the onset of deep convection 256 in more details without higher temporal resolution output. Most likely, the causality will 257 be different for each model, based on their choices of parameterisation in the atmosphere 258 (Zelinka et al., 2020), sea ice (Keen et al., 2021) and ocean (Heuzé et al., 2023), and even 259 their definition of the mixed layer (Griffies et al., 2016). But individual model studies 260 require access to each model's code, and are way beyond the scope of this paper. 261

²⁶² 4 Conclusions

We used all CMIP6 models and all their ensemble members for which the mixed 263 layer depth and its potential drivers the surface salinity, temperature, sea ice concentra-264 tion, and wind speed, were available for the strongest warming scenario SSP5-8.5. Af-265 ter removing the ensemble members that had spurious Arctic deep convection (defined, 266 as in Lique et al. (2018), as MLD deeper than 500 m) over 2015-2023, we were left with 267 27 models, of which 11 had no deep convection in the Arctic over 2015 - 2100; 6 that had 268 it extremely rarely (usually 4 years or fewer), by the middle of the run; 5 for which deep 269 convection peaked in the first half of the run and then declined and disappeared; and 270 5 in which deep convection emerged in the Arctic in the second half of the run and still 271 convected in 2100. All models exhibit a cessation of deep convection in the Nordic Seas, 272 showing that deep convection in the Arctic is not simply a northward migration of the 273 Nordic Seas ventilation. The Arctic MLD was most strongly correlated with the surface 274 salinity, and the sign of this correlation depended on whether the model convected or 275 not. Similarly, when and where the models are not convecting, their surface salinity fresh-276 ens; at the location where they do, when they do, it becomes saltier and surface winds 277 are increasing. Neither the exact mechanism triggering deep convection nor the direc-278 tion of the causality between deep convection and that compound salinification and wind 279 event can be investigated in more details with CMIP6 monthly output. The fact that 280

models of the same family have the same convection behaviour; that the depth and fre-281 quency of the maximum MLD is strongly correlated to early-run biases and sensitivity; 282 and that the other processes involved have been linked to individual model parameter-283 isations (Zelinka et al., 2020; Keen et al., 2021; Muilwijk et al., 2023) suggest that the trigger for Arctic deep convection is model-specific, and its determination requires in-285 depth sensitivity studies for each model. Such in-depth investigation could also lead to 286 model improvement. CMIP6 models consistently exaggerate deep convection both in the 287 North Atlantic and in the Southern Ocean (Heuzé, 2021). Understanding why they have 288 no such consensus in the Arctic could hold the key to a more realistic representation of 289 mixing, globally. 290

²⁹¹ 5 Open Research

All CMIP6 data are freely available via the Earth Grid System Federation. For this paper, we primarily used the German Climate Computing Centre (DKRZ) node https:// esgf-data.dkrz.de/search/cmip6-dkrz/.

295 Acknowledgments

²⁹⁶ CH is funded by the Swedish Research Council (dnr 2018-03859).

297 **References**

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No emergence of deep convection in the Arctic Ocean across CMIP6 models

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Key Points:

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7	•	Oceanic deep convection does not emerge and persist in the Arctic in the major-
8		ity of CMIP6 models, despite a cessation in the Nordic Seas.
9	•	Arctic deep convection occurs only when both surface salinity and winds are in-
10		creasing, year round, yet most models are freshening.
11	•	The models with the strongest sensitivity, especially with an oceanic polar am-
12		plification, have the deepest Arctic mixed layers, most often.

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13 Abstract

As sea ice disappears, the emergence of open ocean deep convection in the Arctic has 14 been suggested. Here, using 36 state-of-the-art climate models and up to 50 ensemble 15 members per model, we show that Arctic deep convection is rare even under the strongest 16 warming scenario. Only 5 models have somewhat permanent convection by 2100, while 17 11 have had convection by the middle of the run. For all, the deepest mixed layers are 18 in the Eurasian basin, by St Anna Trough. When the models convect, that region un-19 dergoes a salinification and increasing wind speeds; it is freshening otherwise. We dis-20 cuss the causality and potential reasons for the opposite trends. Given the model's dif-21 ferent parameterisations, and given that the ensemble members that convect the deep-22 est, most often, are those with the strongest sensitivity, we conclude that differences in 23 deep convection are most likely linked to the model formulation. 24

²⁵ Plain Language Summary

Both observations and modelling simulations suggest that deep vertical mixing (or 26 deep convection) in winter may become the new normal in the Arctic as sea ice disap-27 pears. These simulations are often done using only one model, so here we used all mod-28 els available that participated in the Climate Model Intercomparison Project phase 6, 29 for the strongest warming scenario. We show that after removing those that are already 30 inaccurate in the present, and even with a restrictive threshold, most models have no deep 31 32 convection in the Arctic, or extremely rarely. Only 5 still had deep convection by the time the run finishes in 2100. We investigated the possible links between deep convec-33 tion and surface salinity, surface temperature, sea ice concentration and surface wind speeds, 34 and found that the salinity was most important. Deep convection regions and periods 35 are associated with a saltier, windier surface, while the rest of the Arctic and/or rest of 36 the run freshens. Causality is unclear; we need higher resolution than monthly output. 37 Similar behaviours within model families, a strong link to the model sensitivity, and cited 38 work make us conclude that ultimately, the differences are probably caused primarily by 39 the different model designs. 40

41 **1** Introduction

The Arctic Ocean is changing. The resulting reduction in Arctic sea ice extent and 42 thickness (Mallett et al., 2021; Meier & Stroeve, 2022) could enhance vertical mixing: 43 More brine may be rejected year-round as younger, saltier ice desalinates (Peterson, 2018) 44 or as sea ice reforms in winter over the now seasonal-ice areas (Onarheim et al., 2018), 45 while the ice-freed regions may become more susceptible to wind stirring (Timmermans 46 & Marshall, 2020). In the Eurasian Arctic, the process known as Atlantification (Polyakov 47 et al., 2017), whereby warm water of Atlantic origin penetrates further into the Arctic, 48 may be further weakening the stratification and enhancing sea ice melt. These led Polyakov 49 et al. (2017) to hypothesise that the Arctic may start exhibiting deep convection in win-50 ter, a result found by Lique et al. (2018) in the 4x CO₂ scenario using the model HiGEM. 51

However, this hypothesis is so far not confirmed. Peralta-Ferriz and Woodgate (2015) 52 found that a deepening of mixed layers in the Arctic is unlikely, since stratification greatly 53 dominates over the wind effect. The latest observations in the Eurasian Arctic (Schulz 54 et al., under review) yielded mixed layers no deeper than 130 m, even in winter. Besides, 55 in a recent Arctic study using models that participated in the Climate Model Intercom-56 parison Project phase 6 or CMIP6 (Eyring et al., 2016), Muilwijk et al. (2023) showed 57 that there was no agreement among models regarding future stratification and the ef-58 fect of Atlantification under the strongest warming scenario (SSP5-8.5, O'Neill et al. (2016)). 59 This suggests that HiGEM's deep convection in the Arctic may be a model artefact rather 60 than the future of the Arctic. 61

We here determine whether deep convection emerges in the Arctic in the future scenario SSP5-8.5 using all CMIP6 models and all their ensemble members for which the mixed layer depth output was available, as described in section 2. In section 3, we detail the spatial and temporal patterns of future Arctic mixed layers and discuss possible reasons for these, focusing on the model biases and their trends in surface properties, sea ice and winds. We conclude in section 4.

⁶⁸ 2 Data and Methods

2.1 CMIP6 data

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To investigate deep convection and its potential drivers in the future Arctic, we use all CMIP6 models and all their ensemble members that had monthly mixed layer depth ("mlotst"), ocean surface salinity ("sos") and temperature ("tos"), sea ice concentration ("siconc") and surface wind speed ("sfcWind") available on any of the Earth System Grid Federation (ESGF) nodes for the future scenario SSP5-8.5, for January 2015 to December 2100. The models and their ensemble members are listed in supp. Tables S1 and S2.

For AWI-CM-1-1-MR and GISS-E2-1-G, we used the sea ice thickness ("sivol") be-76 cause sea ice concentration was not available. For CAMS-CSM1-0, we generated the sur-77 face salinity from the full-depth salinity ("so") as the former was not available; we used 78 the salinity of the shallowest level, 5 m depth. Similarly, we used the models' bathymetry 79 ("deptho") when it was available, but had to generate it from the full-depth salinity for 80 11 models, as the last level with salinity data. Finally, for CESM2-WACCM, GFDL-CM4 81 and MRI-ESM2-0, several grid types were available and we chose for simplicity the reg-82 ularised grid ("gr"). For the other models, we took the one grid type available; see supp. 83 Tables S1 and S2. 84

2.2 Methods

The thresholds and choices of this subsection are discussed in supplementary text 86 S1. In agreement with Lique et al. (2018), we consider that there is deep convection in 87 the Arctic if the mixed layer depth (MLD) exceeds 500 m. We here do not quantify over-88 turned volumes. We only perform a binary detection of deep convection, so we select the 89 overall maximum MLD, in space or time depending on the analysis. The ensemble mem-90 bers that exhibit deep convection in the Arctic over the observed part of SSP5-8.5 (2015-91 2023) are shown on the first figure and subsequently removed from the study, as they 92 are already inaccurate at the beginning of the run. Consequently, of the originally 36 mod-93 els, 27 remain for most of the analysis. We do not take biases in Nordic Seas MLD into 94 account for model selection, as all CMIP6 models have spurious deep convection there 95 (Heuzé, 2021). 96

The sea ice edge (supp. Fig S1) is detected as the contour of 15% concentration or 10 cm thickness, averaged over 2040-2060 and 2080-2100. We define the Eurasian basin as the region north of 80°N, longitudes 20°W - 140°E, deeper than 1000 m, and the Nordic Seas as the region of latitudes 66 - 80°N, longitudes 30°W - 20°, deeper than 1000 m see contours on Fig 1. We perform across-model correlations by comparing, for each model and each ensemble member, their maximum MLD and number of year where the MLD exceeds 500 m in the Eurasian basin, to:

- 104 105
- Average surface temperature and salinity in the Eurasian basin over the first 20 years of SSP5-8.5;
- Tropical and Mid-latitude (60°S 60°N) and Arctic (north of 75°N) warming, as
 difference between the 2080-2100 and 2015-2035 average ocean surface tempera ture, which we use as a proxy for the model sensitivity. We also define the oceanic

polar amplification as the difference between the Arctic and tropical/mid-latitudewarming.

Besides, we compute correlations and trends for each grid cell, after interpolating 111 all the parameters onto each model's mlotst grid. The correlations and trends are com-112 puted for each ensemble member separately, and averaged afterwards. We consider that 113 there is ensemble agreement if more than 50% of the ensemble members have a signif-114 icant correlation/trend (determined using a t-test at 95% significance) of the same sign; 115 we then present the median value and its ensemble-spread. Finally, we present the com-116 posite trends after grouping the models based on their Arctic deep convection behaviour, 117 as described in the result section. 118

¹¹⁹ **3** Results and Discussion

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3.1 Arctic deep convection is rare, restricted both in space and time

The maximum mixed layer depth reached in the Arctic over 2015-2100 varies strongly 121 across models (Fig. 1, note the logarithmic scale). The value does not exceed 100 m for 122 some, such as CAMS-CSM1-0, while others such as CMCC-CM2-SR5 exceed 2000 m in 123 the majority of the Eurasian basin. All models with deep MLD agree that the deepest 124 values are in the Eurasian basin, most commonly by St Anna Trough. Nine models have 125 MLD exceeding the 500 m threshold for deep convection already over 2015 to present 126 for all their ensemble members (red squares on Fig. 1). Interestingly, these are not only 127 by St Anna Trough but also north of Svalbard, suggesting that for these models the Nordic 128 Sea deep convection area extends too far north, most likely following the sea ice edge (shown 129 on supp. Fig. S1). All models presented on Fig. 1 have spuriously deep MLD in the Nordic 130 Seas. 131

After removing the models and ensemble members that are unrealistic in the presentday regarding their Arctic MLD (asterisks in supp Tables S1 and S2), 27 models remain. The temporal evolution of their MLD reveals four groups of models (Fig. 2 and supp Table S3):

136	1.	The first 11 models have no Arctic deep convection during the entire run. The Arc-
137		tic MLD time series (blue lines, Fig. 2) are mostly flat, with no year where the
138		MLD exceeds 500 m, regardless of the ensemble member. The maximum MLD across
139		these models and their ensemble members is often of the order of 100 m, i.e. like
140		currently observed in the Arctic (Schulz et al., under review).
141	2.	Six models have deep convection in the Arctic on rare occasions, in the middle of
142		the run. The maximum number of years with deep convection for this group is 17
143		out of 86 (supp Table S3), but is most often 4 or fewer.
144	3.	Five models also have deep convection in the Arctic in the middle of the run, more
145		often. It starts by 2030, peaks in the first half of the run, and then declines slowly,
146		with the ensemble average back under the 500 m threshold by the end of the run.
147	4.	The last 5 models start convecting in the second half of the run, by 2070, and ap-
148		pear "stably" convecting at the end of the run. Unfortunately, only one ensem-
149		ble member of one of the models (ACCESS-CM2) is available beyond 2100, so we
150		cannot tell whether deep convection in these models would also decline later.

The Nordic Seas MLD (red lines, Fig. 2) falls below 1000 m for all models, and even below 500 m for two thirds of them. The consistent cessation of Nordic Seas deep convection is not related to the models' behaviour in the Arctic. Therefore, unlike suggested by Lique et al. (2018), deep convection does not migrate to the Arctic in response to its cessation further south.



Figure 1. Maximum MLD over January 2015 - December 2100 (shading, logarithmic scale) for SSP5-8.5 at each grid cell. For each model, parentheses indicate the number of ensemble members available (see supp. Tables S1 and S2); when this number is larger than one, the figure shows the ensemble median. Black contours are the 1000 m isobath. Red squares indicate the models for which all ensemble members have deep convection in the Arctic already over 2015-2023, and which are therefore not considered for further analysis. Locations discussed in the manuscript are indicated on the top-right panel: Cyan contour "EB" is the Eurasian Basin, indigo dashed contour "GIN" is the Nordic Seas, and magenta arrow "SAT" is the St Anna Trough.



Figure 2. For the 27 models that do not have deep convection in the Arctic over 2015-2023, time series of their yearly maximum MLD in the Nordic Seas (red) and in the Arctic Ocean (blue). For each model, parentheses indicate the number of ensemble members remaining; when this number is larger than one, the figure shows the range across these ensemble members (shading) and the ensemble median (thick line). Horizontal black lines indicate the 500 and 1000 m MLD thresholds, indicative of deep convection. Vertical black line is the year 2023. Models are ordered based on their Arctic behaviour: First two rows and NorESM2-LM, no Arctic deep convection; ACCESS-ESM1-5 and fourth row, rare convection, by the middle of the run; fifth row, convection peaks by the middle of the run and then declines; bottom row, convection starts late in the run.

What causes these different behaviours then? Ensemble members usually have a consistent behaviour (the shading on Fig. 2 usually agrees with the thick line), and models of the same family tend to belong to the same group. The only exception are the two ACCESS models (Fig. 2), but there are large differences in their designs and implemented schemes of relevance for polar regions and deep convection in particular, as discussed in Mohrmann et al. (2021). In the next section, we investigate in more details what could be causing the different Arctic deep convection behaviours, starting with model designs.

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3.2 Potential causes: High sensitivity, salinification, stronger winds

The models that have deep convection in the Arctic usually have it at the same location (Fig. 1), but the temporal evolution of the mixed layer yields four groups of models (Fig. 2). Models of the same family tend to belong to the same group; these usually have similar biases and sensitivities, so we first investigate these.

The mean temperature in the Eurasian Basin at the beginning of the SSP5-8.5 run 168 and subsequent MLD are positively correlated across models (Fig. 3, left): the warmer 169 at the beginning of the run, the deeper the mixed layers (correlation 0.40), the more of-170 ten (0.33). The same results are obtained when considering the ensemble members sep-171 arately, as on Fig. 3, and the ensemble mean (not shown). There is no such relationship 172 with salinity. The relationship with temperature persists throughout the run, so that the 173 models that warm the most (Fig. 3, centre), and quite strikingly, those for which the ocean 174 surface warms more in the Arctic than in the midlatitudes (Fig. 3, right) are the ones 175 with the deepest MLD, the most often. The models belonging to the E3SM project (cyan 176 contour, Fig. 3) are the exception: They warm strongly but convect rarely or not at all, 177 which could be because their design is very different from that of the other CMIP mod-178 els (Golaz et al., 2019). 179



Figure 3. For each model and each ensemble member that were not removed from the study, scatter plot between the maximum MLD over 2015-2100, over the Eurasian Basin, and: Left, the mean ocean surface temperature over the Eurasian basin, over the first 20 years of the run; Centre, the Arctic warming, i.e. the mean ocean surface over the last 20 years of the run minus that of the first 20 years; Right, that Arctic warming minus that of the tropical to mid-latitudes, which we call "Oceanic Polar Amplification". See Methods for latitude definitions. Colors indicate the number of years where the Arctic MLD exceeds the 500 m threshold, as per supp. Table S3. Cyan contours highlight the E3SM project models (see text).

Although the salinity at the beginning of the run has no relationship with the max-180 imum MLD reached, salinity variations are strongly correlated to MLD variations (Ta-181 ble 1) and the correlation differs depending on the Arctic deep convection behaviour of 182 the models. Most models that have no deep convection in the Arctic or rarely (first two 183 blocks) have a positive correlation between their March MLD and the surface salinity 184 one month before, and a negative correlation with the ocean surface temperature: These 185 models have shallower MLD when they are fresher and warmer (as expected in a chang-186 ing Arctic, e.g. Peralta-Ferriz and Woodgate (2015)). In the two groups of convecting 187 models, especially so in the convective region (lines "DC", Table 1), the correlation is 188 positive with salinity and temperature: deeper MLD are associated with salier and warmer 189 surface waters the month prior. Of the four drivers investigated, the MLD is most strongly 190 correlated with salinity or temperature for the vast majority of models (bold fonts on 191 Table 1); usually it is strongest with the salinity for the non or rarely convecting mod-192 els, and with the temperature for the convecting ones (as expected from the previous para-193 graph). Correlations are similar when considering possible drivers the summer before (not 194 shown, see also supp. Text S1). Unsurprisingly, the possible drivers are not independent 195 (supp. Table S4). The correlations between temperature and salinity, and between salin-196 ity and either sea ice concentration or wind speed are of different signs depending on the 197 convecting behaviour, suggesting that different processes and/or water masses are involved 198 (see next subsection). 199

From the correlation analysis, we suspect that the models with no or rare deep convection may be freshening, while those with convection may become saltier. A trend analysis confirms this hypothesis:

- The models with no deep convection in the Arctic become fresher throughout the Arctic, throughout the run (Fig. 4, first two lines). Their trends are rather weak compared to the other model groups. Their winter sea ice does not retreat far into the Arctic, even by the end of the run (supp. Fig. S1).
- 2. The models with deep convection at the middle of the run, be it rarely (Fig. 4, 207 lines 3 and 4) or peaking and declining (lines 5 and 6) exhibit similar trends. Their 208 ocean surface becomes saltier and winds stronger at the location where mixed lay-209 ers deepen in the first half of the run; they freshen in the rest of the Arctic. In the 210 second half, when mixed layers are shallow again, the ocean surface freshens ev-211 erywhere. The main difference between these two groups is in their sea ice trends, 212 with the models whose convection peaks and slowly declines having no winter sea 213 ice by the end of the run (supp. Fig. S1). 214
- 3. The models that convect at the end of the run (Fig. 4, last two lines) have the opposite salinity trends: first a freshening, then a salinification in the region where mixed layers deepen, along with stronger winds.

Note that the trend patterns are similar in summer (supp. Figure S2), indicating that
the changes occur year round. The conclusion is that Arctic deep convection is associated with both a saltier ocean surface and stronger winds. Stronger winds alone are not
enough (see e.g. the no deep convection group, second half of the run).

3.3 On causality

Deep convection in the Nordic Seas ceases for all CMIP6 models, but rarely emerges as a stable feature in the Arctic. In the Arctic, we find that deep convection is associated with a saltier ocean surface and stronger wind speeds, and is most intense and durable in the models with the strongest sensitivity, especially so if the Arctic warms more than the mid-latitudes. But are the trends we observe the causes or consequences of deep convection?

Table 1. Correlation coefficient and its standard deviation for each model between the March MLD and the previous February ocean surface salinity (S), ocean surface temperature (T), sea ice concentration (SIC) and surface wind speed (Wind), for the Eurasian basin (EB) and, for the models with Arctic deep convection, where the maximum MLD of Fig. 1 exceeds 500 m (DC). Models are ordered based on their Arctic deep convection behaviour, as per Fig. 2. Only correlations significant at 95% are shown; for models with more than one ensemble member, median of the correlations of the dominating sign. For models with more than one ensemble member, standard deviation is the across-ensemble spread; spatial spread otherwise. Bold fonts highlight the maximum correlation for each model. Correlation between parameters is shown in supp. Table S4.

Model	Region	MLD vs S	MLD vs T	MLD vs SIC	MLD vs Wind
BCC-CSM2-MR	EB	0.47 ± 0.15	-0.41 ± 0.17	-0.28 ± 0.35	0.23 ± 0.24
CAMS-CSM1-0	\mathbf{EB}	0.54 ± 0.02	-0.49 ± 0.00	-0.35 ± 0.03	0.29 ± 0.04
CAS-ESM2-0	\mathbf{EB}	$\textbf{0.59}\pm0.04$	-0.49 ± 0.06	0.51 ± 0.06	-0.24 ± 0.01
CESM2-FV2	\mathbf{EB}	-0.29 ± 0.37	0.22 ± 0.37	-0.23 ± 0.16	0.31 ± 0.08
CESM2-WACCM	\mathbf{EB}	0.34 ± 0.06	-0.31 ± 0.05	0.25 ± 0.06	0.30 ± 0.02
CESM2	\mathbf{EB}	0.32 ± 0.08	-0.30 ± 0.04	0.30 ± 0.05	0.32 ± 0.02
E3SM-1-1-ECA	\mathbf{EB}	0.43 ± 0.14	-0.45 ± 0.47	$\textbf{0.60} \pm 0.41$	-0.31 ± 0.31
FGOALS-f3-L	\mathbf{EB}	0.51 ± 0.21	-0.51 ± 0.10	0.59 ± 0.13	-0.26 ± 0.16
FGOALS-g3	\mathbf{EB}	$\textbf{0.89} \pm 0.01$	-0.28 ± 0.02	0.46 ± 0.04	0.28 ± 0.04
GFDL-CM4	\mathbf{EB}	$\textbf{0.39} \pm 0.31$	-0.25 ± 0.39	-0.31 ± 0.29	-0.27 ± 0.19
NorESM2-LM	EB	-0.28 ± 0.36	$\textbf{0.36}\pm0.31$	-0.26 ± 0.22	0.26 ± 0.16
ACCESS-ESM1-5	EB	-	-	-	-
	DC	-	-	-	-
E3SM-1-0	EB	0.38 ± 0.01	-0.50 ± 0.12	0.51 ± 0.05	-0.26 ± 0.04
	DC	-	-	-	-
E3SM-1-1	EB	0.39 ± 0.14	-0.41 ± 0.27	0.55 ± 0.28	-0.28 ± 0.27
	DC	0.51 ± 0.01	0.34 ± 0.02	-0.29 ± 0.01	0.43 ± 0.01
GISS-E2-1-G	\mathbf{EB}	0.61 ± 0.03	-0.59 ± 0.04	0.35 ± 0.08	0.27 ± 0.04
	DC	0.48 ± 0.14	0.67 ± 0.10	-0.37 ± 0.08	0.32 ± 0.09
MRI-ESM2-0	\mathbf{EB}	0.62 ± 0.03	-0.40 ± 0.04	0.29 ± 0.05	0.31 ± 0.04
	DC	-	-	-	-
NESM3	\mathbf{EB}	$\textbf{-0.31}\pm0.01$	-0.24 ± 0.01	-0.24 ± 0.01	-0.23 ± 0.02
	DC	-	-	-	-
CMCC-ESM2	\mathbf{EB}	$\textbf{0.59} \pm 0.45$	0.45 ± 0.37	-0.31 ± 0.43	0.28 ± 0.25
	DC	0.65 ± 0.11	0.49 ± 0.14	-0.40 ± 0.12	0.29 ± 0.08
CanESM5-1	EB	0.50 ± 0.03	0.39 ± 0.11	-0.46 ± 0.08	0.36 ± 0.05
	DC	0.52 ± 0.04	0.39 ± 0.12	-0.48 ± 0.08	0.36 ± 0.05
CanESM5	\mathbf{EB}	-	-	-	-
	DC	-	-	-	-
HadGEM3-GC31-LL	\mathbf{EB}	-	-	-	-
	DC	-	-	-	-
UKESM1-0-LL	\mathbf{EB}	0.56 ± 0.02	0.64 ± 0.01	-0.51 ± 0.02	0.43 ± 0.05
	DC	0.57 ± 0.02	0.65 ± 0.02	-0.54 ± 0.02	0.43 ± 0.06
ACCESS-CM2	\mathbf{EB}	0.40 ± 0.05	0.45 ± 0.19	-0.28 \pm 0.08	0.28 ± 0.03
	DC	0.44 ± 0.05	$\textbf{0.76}\pm0.11$	-0.51 \pm 0.08	0.34 ± 0.04
CNRM-CM6-1	\mathbf{EB}	0.40 ± 0.07	0.73 ± 0.05	-0.62 ± 0.05	0.38 ± 0.05
	DC	0.46 ± 0.05	0.77 ± 0.08	-0.68 ± 0.05	0.47 ± 0.04
CNRM-ESM2-1	\mathbf{EB}	0.44 ± 0.15	$\textbf{0.77} \pm 0.48$	$\textbf{-}0.53\pm0.31$	0.43 ± 0.13
	DC	0.46 ± 0.09	$\textbf{0.86}\pm0.10$	$\textbf{-}0.57\pm0.16$	0.47 ± 0.12
EC-Earth3-Veg-LR	\mathbf{EB}	0.44 ± 0.04	0.85 ± 0.05	-0.55 \pm 0.12	0.35 ± 0.05
	DC	0.45 ± 0.03	0.90 ± 0.05	-0.66 ± 0.04	0.39 ± 0.03
EC-Earth3-Veg	EB	0.47 ± 0.13	$\textbf{0.85} \pm 0.43$	$\textbf{-}0.49\pm0.17$	0.32 ± 0.09
	DC	0.52 ± 0.10	$\textbf{0.89} \pm 0.07$	-0.53 ± 0.12	0.34 ± 0.10



Figure 4. Composite trends based on the models' Arctic deep convection behaviours of Fig. 2, for the first half of the 21st century run (top rows) and the second half (bottom rows), in March MLD (first column), and February ocean surface salinity (S, second), ocean surface temperature (T, third), sea ice concentration (fourth) and surface wind speed (last). Behaviours are described to the left of the figure, and number of models for each behaviour is given in parentheses. For each panel, stippling indicates non-significant trends and/or model disagreement regarding the trend's sign. Straight stippled lines across the North Pole are an artefact of the necessary interpolation. -10-

The Arctic Ocean surface warming and sea ice loss trends, year-round, are to be 229 expected in a warming world (IPCC, 2019). Besides, they are consistent across-models, 230 regardless of their convecting behaviour in the Arctic. Similarly, the models' sensitiv-231 ities have been attributed to their representation of cloud cover and cloud albedo (Zelinka 232 et al., 2020), which locally can be affected by the heat and moisture fluxes from deep con-233 vection (Monroe et al., 2021), but in coarse models is more likely due to each model's 234 cloud parameterisation (Zelinka et al., 2020). Finally, changes in both "normal" winds 235 (Screen et al., 2018) and cyclones (Rinke et al., 2017) are most commonly attributed to 236 the large scale atmospheric circulation, although local changes at the boundary layer be-237 cause of sea ice loss may accelerate winds (DuVivier et al., accepted). 238

As for the salinity, Lique et al. (2018) attributed the freshening - salinification dipoles 239 to changes in the large scale oceanic circulation, whereas Davis et al. (2016) attributed 240 the increased salinity to increased vertical mixing. That is, for the former it drives Arc-241 tic deep convection; for the latter, it is a consequence of it. A local change in salinity can 242 be also caused by enhanced sea ice formation, but we find a negative correlation between 243 winter salinity and sea ice concentration among the convecting models (supp. Table S4), 244 which makes this causality unlikely given that the sea ice maps do not seem to exhibit 245 polynyas (supp. Fig. S1). Salinification taking place year-round, at the same location 246 as winds are increasing, would instead suggest an enhanced sea ice drift, but the detailed 247 sea ice mass budget of Keen et al. (2021) shows the opposite. 248

Alternatively, the impact of deeper MLD on the Arctic surface salinity will depend 249 on each model's Atlantic layer. Heuzé et al. (2023) showed strong biases in that layer, 250 with most models having it too deep. Khosravi et al. (2022) further showed that there 251 is no consistency across CMIP6 models regarding the changes of that Atlantic layer dur-252 ing SSP5-8.5, which Muilwijk et al. (2023) linked to the models' lack of consensus re-253 garding future changes in stratification in the Eurasian basin and in Atlantification. As 254 stratification and deep convection are intimately linked, this is another feedback that makes 255 the causality uncertain. One can unfortunately not study the onset of deep convection 256 in more details without higher temporal resolution output. Most likely, the causality will 257 be different for each model, based on their choices of parameterisation in the atmosphere 258 (Zelinka et al., 2020), sea ice (Keen et al., 2021) and ocean (Heuzé et al., 2023), and even 259 their definition of the mixed layer (Griffies et al., 2016). But individual model studies 260 require access to each model's code, and are way beyond the scope of this paper. 261

²⁶² 4 Conclusions

We used all CMIP6 models and all their ensemble members for which the mixed 263 layer depth and its potential drivers the surface salinity, temperature, sea ice concentra-264 tion, and wind speed, were available for the strongest warming scenario SSP5-8.5. Af-265 ter removing the ensemble members that had spurious Arctic deep convection (defined, 266 as in Lique et al. (2018), as MLD deeper than 500 m) over 2015-2023, we were left with 267 27 models, of which 11 had no deep convection in the Arctic over 2015 - 2100; 6 that had 268 it extremely rarely (usually 4 years or fewer), by the middle of the run; 5 for which deep 269 convection peaked in the first half of the run and then declined and disappeared; and 270 5 in which deep convection emerged in the Arctic in the second half of the run and still 271 convected in 2100. All models exhibit a cessation of deep convection in the Nordic Seas, 272 showing that deep convection in the Arctic is not simply a northward migration of the 273 Nordic Seas ventilation. The Arctic MLD was most strongly correlated with the surface 274 salinity, and the sign of this correlation depended on whether the model convected or 275 not. Similarly, when and where the models are not convecting, their surface salinity fresh-276 ens; at the location where they do, when they do, it becomes saltier and surface winds 277 are increasing. Neither the exact mechanism triggering deep convection nor the direc-278 tion of the causality between deep convection and that compound salinification and wind 279 event can be investigated in more details with CMIP6 monthly output. The fact that 280

models of the same family have the same convection behaviour; that the depth and fre-281 quency of the maximum MLD is strongly correlated to early-run biases and sensitivity; 282 and that the other processes involved have been linked to individual model parameter-283 isations (Zelinka et al., 2020; Keen et al., 2021; Muilwijk et al., 2023) suggest that the trigger for Arctic deep convection is model-specific, and its determination requires in-285 depth sensitivity studies for each model. Such in-depth investigation could also lead to 286 model improvement. CMIP6 models consistently exaggerate deep convection both in the 287 North Atlantic and in the Southern Ocean (Heuzé, 2021). Understanding why they have 288 no such consensus in the Arctic could hold the key to a more realistic representation of 289 mixing, globally. 290

²⁹¹ 5 Open Research

All CMIP6 data are freely available via the Earth Grid System Federation. For this paper, we primarily used the German Climate Computing Centre (DKRZ) node https:// esgf-data.dkrz.de/search/cmip6-dkrz/.

295 Acknowledgments

²⁹⁶ CH is funded by the Swedish Research Council (dnr 2018-03859).

297 **References**

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Supporting Information for "No emergence of deep convection in the Arctic Ocean across CMIP6 models"

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- 1. Text S1
- 2. Figures S1 to S2 $\,$
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Introduction

This document provides one supplementary text, two supplementary figures and four supplementary tables, referred to in the main text as Figures S1-s2 and Tables S1-S4, as well as the corresponding references.

Text S1. Choices of thresholds and robustness Globally, a deeper threshold than 500 m is commonly used to determine deep convection. Using a threshold of 1000 m instead of that of 500 m (chosen by consistency with Lique, Johnson, and Plancherel (2018)) does not significantly change the results. It only increased the number of models that are not convecting in the Arctic from 11 to 18, as can be seen on the main text Fig. 2. Time series of the yearly maximum mixed layer depth have been produced with the actual maximum regardless of time; April, month of deepest mixed layers for the few convecting models (not shown); and March, the month traditionally used and where the non-convecting models have their deepest mixed layers. The differences between the results were not significant. We therefore perform correlations and show results using the March mixed layers, for consistency with other studies.

The sensitivies and oceanic polar amplification values are not significantly affected by the choice of latitude bands. That is, the model order remains the same, only the individual models' magnitudes are changed. We chose to present the results that exclude 60°N to 75°N, as these are dominated by the biased Nordic Seas.

Finally, we tested different lags for the correlations and trends ranging from 1 to 6 months, as we aim to determine which process could be driving changes in the mixed layer depth. We show primarily those with one month lag (i.e. February salinity, temperature, sea ice and winds), that had the strongest signal-to-noise ratio. We also show as supplementary material the 6 months (September) values, although the reader must bear in mind that most models lose their September sea ice early (see supp. Fig. S1), therefore the statistics involving sea ice become less robust. We also tested different thresholds for

the ensemble and model agreement. Because of the low number of ensemble members for most models (especially so after removing the ones that have Arctic deep convection over 2015-2023), 66% was too restrictive: When rounded, it often meant "all ensemble members/models". Our choice of 50% threshold does not change the results significantly. In particular, it does not change the sign or relative magnitude of the correlations/trends. The trends are presented over exactly half of the run, but we tested with different number of years, with the entire run, and visually verified them for every model and ensemble member. We chose to present them separately over the two halves of the run as this matches the models' Arctic deep convection behaviour, as will be discussed.

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Figure S1. Median 2040-2060 (blue) and 2080-2100 (magenta overlaying the blue) March sea ice extent for all 36 models. Threshold is 15% sea ice concentration or 10 cm sea ice thickness. Black + indicate where the 2015-2100 maximum mixed layer depth (MLD) of Fig. 1 exceeds 500 m. Red boxes as per Fig. 1. September 20, 2023, 3:11pm



Figure S2. Composite trends based on the models' Arctic deep convection behaviours, for the first half of the 21st century run (top rows) and the second half (bottom rows), in March mixed layer depth (MLD, first column), and September ocean surface salinity, ocean surface temperature, sea ice concentration <u>gpdtsurface</u> <u>wind</u> <u>spece</u>. <u>3.9449</u> June 2023, <u>3.9449</u> June 2023, <u>3.9449</u> June 2023, <u>3.9449</u> June 2023, June 2023,

with the model's refe	erence ^a		
Model	Ensemble members	Grid	Reference
	r1i1p1f1; r2i1p1f1; r3i1p1f1*; r4i1p1f1;		
ACCESS-CM2	r5i1p1f1; r6i1p1f1; r7i1p1f1; r8i1p1f1;	gn	Mackallah et al. (2022)
	r9i1p1f1*; r10i1p1f1*		
	r1i1p1f1; r2i1p1f1; r3i1p1f1*; r4i1p1f1*;		
	r5i1p1f1; r6i1p1f1; r7i1p1f1*; r8i1p1f1;		
	r9i1p1f1; r10i1p1f1; r11i1p1f1*; r12i1p1f1;		
	r13i1p1f1; r14i1p1f1; r15i1p1f1*;r16i1p1f1;		
ACCESS-ESM1-5	r17i1p1f1; r18i1p1f1*; r19i1p1f1; r20i1p1f1;	gn	Mackallah at al. (2022)
NOOLDO LOMII O	r21i1p1f1; r22i1p1f1*; r23i1p1f1*; r24i1p1f1;		
	r25i1p1f1; r26i1p1f1; r27i1p1f1; r28i1p1f1;		
	r29i1p1f1; r30i1p1f1; r31i1p1f1*; r32i1p1f1;		
	r33i1p1f1; r34i1p1f1; r35i1p1f1; r36i1p1f1;		
	r37i1p1f1; r38i1p1f1*; r39i1p1f1; r40i1p1f1		
AWI-CM-1-1-MR*	rlilplf1*	gn	Semmler et al. (2020)
BCC-CSM2-MR	rlilplfl	gn	Wu et al. (2019)
CAMS-CSM1-0	r1i1p1f1; r2i1p1f1	gn	Rong, Li, and Chen (2019)
CAS-ESM2-0	r1i1p1f1; r3i1p1f1	gn	Dong et al. (2021)
CESM2-FV2	r1i2p2f1	gn	Danabasoglu et al. (2020)
CESM2-WACCM	rlilp1f1; r2i1p1f1; r3i1p1f1; r4i1p1f1;	gr	Danabasoglu et al. (2020)
	r5i1p1f1	0-	
CESM2	r4ilplf1; r10ilplf1; r11ilplf1	gn	Danabasoglu et al. (2020)
CMCC-CM2-SR5*	rlilplf1*	gn	Cherchi et al. (2019)
CMCC-ESM2	rlilplfl	gn	Lovato et al. (2022)
CNRM-CM6-1-HR*	rlilplf2*	gn	Voldoire et al. (2019)
CNRM-CM6-1	r111p1f2; r211p1f2; r311p1f2; r411p1f2;	gn	Voldoire et al. (2019)
CNDM ECM9 1	r511p1f2; r611p1f2		C_{ij}
UNRM-ESM2-1	r111p112	gn	Seleman et al. (2019)
	riiipiii; riiip2ii; r2iipiii; r2iip2ii; 		
ComEQME 1	r511p111'; r511p211'; r411p111'; r411p211'; r5:1p1f1*, r5:1p2f1, r6:1p1f1*, r6:1p2f1*,	c:10	Circulated at al. (2022)
CanESM5-1	r511p111'; r511p211; r611p111'; r611p211'; r7:1p1f1, r7:1p9f1, r6:1p1f1*,r6:1p9f1.	gn	Sigmond et al. (2023)
	1/11p111; 1/11p211; fol1p111; fol1p211; n0;1p1f1*, n0;1p2f1*, n10;1p1f1*, n10;1p2f1*		
CanFSM5 CanOF*	1911p111, 1911p211, 11011p111, 11011p211 r1j1p9f1*, r9j1p9f1*, r2j1p9f1*	an	Swart of al. (2010)
	1111p211,1211p211,1311p211	gn	Swart et al. (2019)

Table S1. Models, ensemble members and corresponding grid type used in this study, along

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^a Ensemble members marked with an asterisk (*) have deep convection in the Arctic over 2015

- 2023 already and were excluded from the correlation and trend analyses. Models marked with an asterisk (*) have deep convection in the Arctic over 2015 - 2023 already for all their ensemble members.

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 Table S2.
 Table S1 continued

Model	Ensemble members	Grid	Reference
	r1i1p1f1*; r1i1p2f1*; r2i1p1f1*; r2i1p2f1*;		
	r3i1p1f1*; r3i1p2f1*; r4i1p1f1*; r4i1p2f1;		
	r5i1p1f1*; r5i1p2f1*; r6i1p1f1*; r6i1p2f1*;		
	r7i1p1f1*; r7i1p2f1*; r8i1p1f1*;r8i1p2f1;		
CanESM5	r9i1p1f1*; r9i1p2f1*; r10i1p1f1*; r10i1p2f1*;		
	r11i1p1f1*; r11i1p2f1*; r12i1p1f1*; r12i1p2f1*;		
	r13i1p1f1*;r13i1p2f1*; r14i1p1f1*; r14i1p2f1;	gn	Swart et al. (2019)
	r15i1p1f1*; r15i1p2f1*; r16i1p1f1*; r16i1p2f1;	0	· · · · · ·
	r17i1p1f1*: r17i1p2f1: r18i1p1f1*: r18i1p2f1*:		
	r19i1p1f1*: r19i1p2f1*: r20i1p1f1*: r20i1p2f1*:		
	r21i1p1f1*: r21i1p2f1*: r22i1p1f1*: r22i1p2f1*:		
	r23i1p1f1*: r23i1p2f1*: r24i1p1f1*: r24i1p2f1*:		
	r25i1p1f1*: r25i1p2f1*		
	r1i1p1f1: r2i1p1f1: r3i1p1f1: r4i1p1f1:		
E3SM-1-0	r5i1p1f1	gr	Golaz et al. (2019)
E3SM-1-1-ECA	rlilplfl	er	Golaz et al. (2019)
E3SM-1-1	r1i1p1f1	gr	Golaz et al. (2019)
EC-Earth3-Veg-LR	r1i1p1f1: r2i1p1f1: r3i1p1f1	gn	Döscher et al. (2021)
EC-Earth3-Veg	r1i1p1f1*: r2i1p1f1	gn	Döscher et al. (2021)
FGOALS-f3-L	r1i1p1f1	gn	He et al. (2019)
FGOALS-g3	r1i1p1f1: r2i1p1f1: r3i1p1f1: r4i1p1f1	gn	Li et al. (2020)
GFDL-CM4	r1i1p1f1	gr	Adcroft et al. (2019)
	r1i1p1f2: r1i1p3f1: r1i1p5f1*: r2i1p1f2:	0	
GISS-E2-1-G	r2i1p3f1: r2i1p5f1: r3i1p1f2*: r3i1p3f1:	gn	Kellev et al. (2020)
	r4i1p1f2: r4i1p3f1	0	
HadGEM3-GC31-LL	r1i1p1f3; r2i1p1f3; r3i1p1f3; r4i1p1f3	gn	Kuhlbrodt et al. (2018)
HadGEM3-GC31-MM*	r1i1p1f3*; r2i1p1f3*; r3i1p1f3*; r4i1p1f3*	gn	Kuhlbrodt et al. (2018)
MIROC-ES2H*	r1i1p4f2*; r2i1p4f2*; r3i1p4f2	gn	Hajima et al. (2020)
MIROC6*	r1i1p1f1*	gn	Tatebe et al. (2019)
MPI-ESM1-2-HR*	r2i1p1f1*	gn	Müller et al. (2018)
	r1i1p1f1*; r2i1p1f1*; r3i1p1f1*; r4i1p1f1*;	0	
	r5i1p1f1*; r6i1p1f1*; r7i1p1f1*; r8i1p1f1*;		
	r9i1p1f1*; r10i1p1f1*; r11i1p1f1*; r12i1p1f1*;		
	r13i1p1f1*; r14i1p1f1*; r15i1p1f1*; r16i1p1f1*;		
MPI-ESM1-2-LR*	r17i1p1f1*; r18i1p1f1*; r19i1p1f1*; r20i1p1f1*;	gn	Mauritsen et al. (2019)
	r21i1p1f1*; r22i1p1f1*; r23i1p1f1*; r24i1p1f1*;		
	r25i1p1f1*; r26i1p1f1*; r27i1p1f1*; r28i1p1f1*;		
	r29i1p1f1*; r30i1p1f1*		
MDI EGMO O	r1i1p1f1; r1i2p1f1; r2i1p1f1; r3i1p1f1;		$V_{\rm el}$
MRI-ESM2-0	r4i1p1f1; r5i1p1f1*	gr	Yukimoto et al. (2019)
NESM3	r1i1p1f1; r2i1p1f1	gn	Cao et al. (2018)
NorESM2-LM	r1i1p1f1	gn	Seland et al. (2020)
IIKESM1 O I I	r1i1p1f2; r2i1p1f2*; r3i1p1f2; r4i1p1f2;	-	Sollar at al (2010)
UKESWII-U-LL	r8i1p1f2*	gn	Senar et al. (2019)

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Table S3.Table version of Figure 2: Models; number of ensemble members used, across-
ensemble minimum - median - maximum of the yearly maximum mixed layer depth (in meters);
across-ensemble minimum - mode - maximum number of years with deep convection in the Arctic;

Model	Nb.	Across-ensemble max MLD	Across-ensemble years	Type
	members	min - median - max	min - mode - max	
ACCESS-CM2	7	883 - 1041 - 1134	10 - 23 - 25	Late
ACCESS-ESM1-5	30	661 - 1019 - 1356	2 - 4 - 17	MidR
BCC-CSM2-MR	1	104	0	NoDC
CAMS-CSM1-0	2	198 - 224 - 251	0 - 0 - 0	NoDC
CAS-ESM2-0	2	157 - 165 - 173	0 - 0 - 0	NoDC
CESM2-FV2	1	97	0	NoDC
CESM2-WACCM	5	99 - 101 - 102	0 - 0 - 0	NoDC
CESM2	3	108 - 108 - 111	0 - 0 - 0	NoDC
CMCC-ESM2	1	1147	73	MidD
CNRM-CM6-1	6	894 - 946 - 1045	22 - 22 - 33	Late
CNRM-ESM2-1	1	901	21	Late
CanESM5-1	8	1011 - 1193 - 1808	43 - 65 - 69	MidD
CanESM5	6	1262 - 1497 - 1969	49 - 55 - 60	MidD
E3SM-1-0	5	492 - 575 - 698	0 - 1 - 4	MidR
E3SM-1-1-ECA	1	497	0	NoDC
E3SM-1-1	1	701	4	MidR
EC-Earth3-Veg-LR	3	1207 - 1439 - 1450	16 - 16 - 31	Late
EC-Earth3-Veg	1	1322	33	Late
FGOALS-f3-L	1	350	0	NoDC
FGOALS-g3	4	229 - 255 - 258	0 - 0 - 0	NoDC
GFDL-CM4	1	166	0	NoDC
GISS-E2-1-G	8	302 - 625 - 875	0 - 0 - 10	MidR
HadGEM3-GC31-LL	4	1325 - 1418 - 2552	47 - 47 - 72	MidD
MRI-ESM2-0	5	352 - 511 - 554	0 - 0 - 2	MidR
NESM3	2	295 - 437 - 579	0 - 0 - 1	MidR
NorESM2-LM	1	92	0	NoDC
UKESM1-0-LL	3	1097 - 1325 - 1517	47 - 47 - 56	MidD

type of A	rctic deep	convection	behaviour	for	the com	posites)) ^b
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^a The ensemble members are those listed in supp. Tables S1 and S2 without an asterisk, i.e.

that do not have deep convection in the Arctic already over 2015-2023.

^b Types are: "NoDC", no deep convection in the Arctic over 2015 - 2100 in SSP5-8.5; "MidR",

rare Arctic deep convection, mid-run; "MidD", Arctic deep convection peaks mid-run and declines; "Late", Arctic deep convection late in the run.

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Table S4. Correlation coefficient^afor each model between the February ocean surface salinity (S), temperature (T), sea ice concentration (SIC) and surface wind speed (Wind), for the Eurasian basin (EB) and, for the models with Arctic deep convection, where the maximum MLD of Fig.

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Model	Region	S vs T	S vs SIC	S vs Wind	T vs SIC	T vs Wind	SIC vs Wind
BCC-CSM2-MR	EB	-0.84	0.26	0.25	-0.42	-0.23	-0.24
CAMS-CSM1-0	\mathbf{EB}	-0.98	-0.26	0.24	-0.24	0.26	-0.39
CAS-ESM2-0	\mathbf{EB}	-0.76	0.86	-0.34	-0.78	0.29	-0.32
CESM2-FV2	\mathbf{EB}	-0.98	0.55	-0.40	-0.69	0.39	-0.36
CESM2-WACCM	\mathbf{EB}	-0.78	0.60	-0.33	-0.89	0.36	-0.36
CESM2	\mathbf{EB}	-0.77	0.76	-0.45	-0.86	0.37	-0.46
E3SM-1-1-ECA	\mathbf{EB}	0.57	-0.41	0.35	-0.75	0.48	-0.58
FGOALS-f3-L	\mathbf{EB}	0.23	0.25	0.27	-0.66	0.29	-0.29
FGOALS-g3	\mathbf{EB}	-0.27	0.54	0.28	-0.32	-0.26	-0.24
GFDL-CM4	\mathbf{EB}	-0.90	0.46	-0.29	-0.58	0.35	-0.31
NorESM2-LM	\mathbf{EB}	-0.89	0.30	0.27	-0.41	0.27	-0.39
ACCESS-ESM1-5	EB	-	-	-	-	-	-
	DC	-	-	-	-	-	-
E3SM-1-0	\mathbf{EB}	-0.25	0.27	0.25	-0.80	0.43	-0.55
	DC	-	-	-	-	-	-
E3SM-1-1	\mathbf{EB}	-0.48	0.40	-0.22	-0.84	0.44	-0.54
	DC	-	-0.27	0.42	-0.86	0.56	-0.60
GISS-E2-1-G	\mathbf{EB}	-0.99	0.64	0.24	-0.65	0.23	-0.30
	DC	0.72	-0.52	0.49	-0.75	0.48	-0.44
MRI-ESM2-0	\mathbf{EB}	-0.70	0.61	-0.25	-0.90	0.31	-0.32
	DC	-	-	-	-	-	-
NESM3	\mathbf{EB}	0.43	-0.48	0.32	-0.84	0.30	-0.35
	DC	-	-	-	-	-	-
CMCC-ESM2	EB	0.64	-0.62	0.40	-0.80	0.47	-0.58
	DC	0.70	-0.66	0.41	-0.76	0.47	-0.57
CanESM5-1	\mathbf{EB}	0.32	-0.40	0.39	-0.84	0.54	-0.63
	DC	0.34	-0.42	0.39	-0.83	0.57	-0.67
CanESM5	\mathbf{EB}	-	-	-	-	-	-
	DC	-	-	-	-	-	-
HadGEM3-GC31-LL	\mathbf{EB}	-	-	-	-	-	-
	DC	-	-	-	-	-	-
UKESM1-0-LL	\mathbf{EB}	0.70	-0.74	0.48	-0.84	0.52	-0.58
	DC	0.72	-0.75	0.50	-0.84	0.55	-0.61
ACCESS-CM2	EB	-0.31	-0.35	-0.23	-0.83	0.40	-0.44
	DC	0.55	-0.53	0.30	-0.82	0.41	-0.45
CNRM-CM6-1	\mathbf{EB}	0.28	-0.30	0.31	-0.82	0.40	-0.50
	DC	0.39	-0.40	0.33	-0.86	0.50	-0.59
CNRM-ESM2-1	\mathbf{EB}	0.38	-0.39	0.34	-0.79	0.50	-0.54
	DC	0.44	-0.44	0.37	-0.84	0.54	-0.57
EC-Earth3-Veg-LR	\mathbf{EB}	0.41,	otember54	0. 2029.33	:11pm0.77	0.43	-0.45
č	DC	0.49	-0.62	0.41	-0.81	0.44	-0.50
EC-Earth3-Veg	\mathbf{EB}	0.41	-0.44	0.32	-0.76	0.34	-0.37
~	DC	0.56	-0.54	0.34	-0.77	0.40	-0.48

^a Only correlations significant at 95% are shown; for models with more than one ensemble member, median of the correlations of the dominating sign.