

Climatology of Arctic temperature inversions in current and future climates

Cairo J. Ruman, Adam H. Monahan, and Laxmi Sushama

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11 ABSTRACT

12 Temperature inversions are a common feature of the Arctic climate, affecting the surface
13 energy budget and planetary boundary layer transports. This study investigates the evolution
14 of large-scale temperature inversions (between 925 hPa and 2m) in the context of a changing
15 climate. To this end, two five-member Regional Climate Model (RCM) ensembles, driven by
16 the Canadian Earth System Model, spanning the 1950–2099 period, corresponding to two
17 greenhouse gas emission scenarios (RCP 4.5 and 8.5), are considered. An ERA-Interim driven
18 simulation for the 1979–2005 period is also considered to assess model performance. A
19 comparison of observed atmospheric soundings with the boundary layer variations in the
20 reanalysis-driven simulation indicates that the model captures the temperature inversion
21 characteristics reasonably well, with some positive biases in the temperature inversion strength
22 and frequency. The transient regional climate change simulations suggest substantial decreases
23 in both temperature inversion strength and frequency in winter in future climate for both
24 emission scenarios. These changes are consistent with the reduced sea ice cover and the
25 associated increase in cloud cover that reduce the surface radiative cooling necessary for the
26 formation of strong temperature inversions. Some increases in the frequency and strength of
27 temperature inversions are projected for summer over the Arctic Ocean, possibly linked with
28 increased poleward moisture transport.

29 **Keywords:** temperature inversion, Arctic climate, climate change, regional climate
30 modeling

31

32 **1. Introduction**

33 Temperature inversions, in which atmospheric temperature increases with height, are a
34 frequent feature of the atmospheric boundary layer in the Arctic. The main temperature
35 inversion formation mechanisms in the Arctic are surface radiative cooling (especially during
36 the polar night), the advection of warm air masses over cooler surfaces, and large-scale
37 subsidence warming in high-pressure situations. Consideration of temperature inversions is
38 important to better understand high-latitude climate change, as they are a crucial feature of the
39 Arctic climate, influencing the surface energy balance and contributing to Arctic amplification
40 of anthropogenic climate change (Serreze and Barry, 2013). Each formation mechanism favors
41 a different type of temperature inversion in the Arctic. Surface radiative cooling can lead to
42 surface-based inversions and are more common in winter months, while large-scale subsidence
43 and advection of warm air tend to form elevated inversions, commonly found in the summer.
44 The temperature inversions can have varied types of structures, from simpler types to multiple
45 inversions layers in the boundary layer, with a mix of surface-based and elevated inversions.

46 Arctic temperature inversions have been analyzed using data derived from satellites and
47 radiosondes (Liu et al., 2006; Devasthale et al., 2010; Zhang et al., 2021), a combination of
48 model simulations and radiosondes (Zhang et al. 2011), and climate models and reanalyses
49 (Medeiros et al., 2011; Wetzal and Brümmer, 2011). Many studies have used the algorithm
50 described by Kahl (1990) to compute the temperature inversion strength and height, and
51 recently Fochesatto (2015) published a methodology for determining multilayered temperature
52 inversions. Both methods require data from multiple layers in the lower levels of the
53 atmosphere. In contrast, other studies have measured the temperature inversion by differencing
54 temperatures between a selected pressure level above the planetary boundary layer (PBL) and
55 either 1000 hPa or the surface temperature (e.g. Bintanja et al. (2011) and Medeiros et al.
56 (2011)).

57 In the present study, we focus on the assessment of projected changes to large-scale Arctic
58 temperature inversion characteristics, motivated by their important role in the Arctic climate.
59 We use regional climate model (RCM) simulations for two future emission scenarios and verify
60 the model results using reanalysis, observations from radiosondes, and satellite data. As we
61 are focusing on large scale inversions in an RCM with relatively coarse vertical resolution,
62 multi-layer techniques such as those of Fochesatto (2015) cannot be used and the detailed
63 structure of the temperature inversion structure is not resolved. As such, the present study
64 defines the temperature inversion strength as the temperature difference between the 925 hPa
65 pressure level and the 2m surface temperature. Estimating the temperature inversion strength
66 by the temperature difference between two levels is a crude estimate of the temperature
67 inversion strength, but reasonable when being used to analyze the large-scale structures and
68 the stability of the lower troposphere in climate models which often have a coarse vertical
69 resolution (Medeiros et al. 2011).

70 Most previous research has focused on winter temperature inversions, as they are more
71 prevalent in that season due to the relatively strong radiative cooling. Summer temperature
72 inversions' frequency and properties are much less dependent on surface net radiation and are
73 largely regulated by the surface melt, poleward intrusion of warm air, and subsidence caused
74 by the presence of polar high (Palo et al. 2017; Zhang et al. 2021). External forcings potentially
75 have a stronger influence on the summer temperature inversion properties than local feedbacks.
76 Most of the Arctic temperature inversion studies in the literature focus on present-day
77 climatologies based on different models and reanalyses (e.g. Zhang & Seidel, 2011; Wang et
78 al, 2020; Chang et al., 2021;). Few studies have investigated future changes in Arctic
79 temperature inversions, with Zhang et al. (2021) focusing on the shallowing of temperature
80 inversion depth and Koenigk et al. (2020) showing that the winter temperature inversion
81 strength is greatly reduced in their work on the Arctic climate change in the 21st century. This

82 study will be one of the first to investigate changes to temperature inversion strength and
83 frequency in future climate projections for the Arctic.

84 The outline of this paper is as follows. The model, data, and methods used in this research
85 are presented in section 2. Model assessment of temperature inversion characteristics and their
86 projected changes are presented in Section 3. Conclusions are summarized in section 4.

87 **2. Model, Data, and Methods**

88 *2.1 Model*

89 The regional climate model used in this study is the Global Environment Multiscale (GEM)
90 model (Côté et al. 1998) in a limited area mode (Teufel and Sushama, 2019). It has a non-
91 hydrostatic dynamical core and uses the Arakawa C grid staggering in the horizontal and
92 vertical coordinate based on hydrostatic pressure (Yeh et al. 2002). The numerical scheme is a
93 two-time-level, semi-Lagrangian, implicit scheme. The GEM physics package includes: deep
94 convection following Kain and Fritsch (1990), shallow convection based on a transient version
95 of the Kuo (1965) scheme (Bélair et al. 2005), large-scale condensation (Sundqvist et al. 1989),
96 correlated K solar and terrestrial radiation (Li and Barker, 2005), subgrid-scale orographic
97 gravity wave drag (McFarlane 1987), low-level orographic blocking (Zadra et al. 2003), and
98 turbulent kinetic energy closure to estimate turbulent fluxes (Benoit et al., 1989; Delage, 1997;
99 Delage and Girard, 1992). Lakes, both resolved and subgrid-scale, are represented by the Flake
100 model (Mironov et al. 2010). The land surface scheme in GEM is the Canadian Land Surface
101 Scheme (CLASS) (Verseghy, 2009), which allows a flexible soil layer configuration. CLASS
102 includes prognostic equations for energy and water conservation for the soil layers and a
103 thermally and hydrologically distinct snowpack where applicable.

104 This study uses two GEM ensembles of 5 members each, which are driven by five different
105 members of a second-generation Canadian Earth System Model (CanESM2) initial condition

106 ensemble, spanning the 1950–2099 period. The model experimental domain is displayed in
107 Fig. 1. The two ensembles correspond to RCP 4.5 and 8.5 scenarios (van Vuuren et al. 2011).
108 The RCP 4.5 has an aggressive but not implausible mitigation scenario (IPCC 2013) in which
109 at the end of the current century the global temperature rises slightly above the +2 °C global
110 mitigation aim. The RCP 8.5 is a high-end scenario, where no mitigation is done due to the
111 absence of policies on climate change, and greenhouse gas emissions continue to grow due to
112 high energy demands (Riahi et al. 2011). CanESM2 (Chylek et al. 2011) consists of the
113 physical coupled atmosphere-ocean model, the Fourth Generation Coupled Global Climate
114 Model (CanCM4), coupled to a terrestrial carbon model (CTEM) and an ocean carbon model
115 (CMOC). The simulations driven by CanESM2 will be referred to as GEM-CAN hereafter. An
116 additional ERA-Interim (Berrisford et al. 2009; Dee et al. 2011) driven simulation, named
117 GEM-ERA, for the 1979–2015 period is used to assess the model's ability to simulate the
118 temperature inversion characteristics based on available observations as discussed below. All
119 model results presented below are ensemble means across the five ensemble members for each
120 RCP.

121 ***2.2 Observation Datasets and Reanalysis Products***

122 The monthly averaged daily temperature dataset of the University of East Anglia Climate
123 Research Unit (CRU, version TS 4.03; Harris et al., 2014) is used to assess model simulated
124 temperatures over land. This dataset is available at 0.5° resolution and covers the global land
125 surface. The Aqua/AIRS L3 Daily Standard Physical Retrieval V7 dataset - AIRX3STD (AIRS
126 project 2019), with 1° horizontal resolution is used to assess the simulated temperature and
127 temperature inversion characteristics over the Arctic Ocean. The AIRS satellite has a sun-
128 synchronous orbit, passing over the same location each day at 1h30 AM/PM local time
129 (respectively the ascending and descending passes). Devasthale et al. (2010) showed that over
130 the Arctic region, the difference in the temperature inversion strength is generally minimal

131 between those two passes, so in this analysis, both passes were combined. The ERA-Interim
132 reanalysis (Berrisford et al. 2009; Dee et al. 2011) is also used for the assessment of the 2m
133 temperature. ERA-Interim is a global atmospheric reanalysis produced by the European Centre
134 for Medium-Range Weather Forecasts (ECMWF) and covers the years from 1979 to the present
135 time.

136 Observational radiosonde data from the Integrated Global Radiosonde Archive (IGRA)
137 dataset (Durre et al. 2006) are used to assess model simulated temperature inversion strength
138 and frequency. The radiosonde data are available twice daily (00UTC and 12 UTC) at more
139 than 1500 locations around the globe, 153 of which are located inside the study domain. The
140 IGRA data pass quality assurance algorithms, checking for format problems, climatological
141 outliers, physically improbable values, and temporal and spatial discrepancies in temperature.
142 Only stations with at least 80% of data during the 2003–2015 period are used for model
143 assessment.

144 ***2.3 Methods***

145 Before studying projected changes to temperature inversion strength and frequency, the
146 ability of the model to simulate these is assessed by comparing GEM-ERA with available
147 observations and reanalyses datasets. Reanalyses and satellite data provide a comprehensive
148 set of data-informed climate variables for such comparisons. Based on the study of three
149 different reanalyses, Lindsay et al. (2014) identify ERA-Interim as the most consistent with
150 independent measurements in the Arctic. ERA5 was not used in the GEM-ERA simulation as
151 it was released after the model simulation was completed. The simulated 2m temperatures are
152 assessed using CRU data, AIRS, and ERA-Interim reanalysis, while the temperature inversion
153 strength and frequency are compared with atmospheric soundings available from the IGRA
154 database and AIRS temperature profiles. The temperature inversion strength is estimated as the

155 temperature difference between the 925 hPa level and the 2m temperature. Temperature
156 inversion frequency is calculated as the ratio of the number of temperature inversion
157 occurrences to the total length of the data record over the period of interest.

158 The AIRS satellite data is composed of satellite swaths, which have spatial and temporal
159 sampling differences that can affect comparisons with climate models (Tian et al. 2013). To
160 lessen this temporal and spatial bias, when assessing model data against the AIRS dataset, the
161 model output was resampled hourly, and data were extracted from the model reproducing the
162 satellite swaths on the model domain so that the spatial and temporal sampling of the model
163 data is similar to that of the AIRS data.

164 Hearty et al. (2014) noted that the AIRS data sometimes report nonphysical temperatures
165 near the surface over land, most notably near steep topography. For this reason, comparisons
166 using the AIRS are only made over the oceans. Chang et al. (2018) found that the temperature
167 inversion strength over the ocean is insensitive to cloud fraction variations, making it a good
168 choice for model assessment. Over continental regions, the CRU dataset and the soundings
169 were used instead. Plots using the AIRS data over the land can be found in the supplementary
170 material (Figure S1).

171 Projected changes to the temperature inversion characteristics and 2m temperature for
172 RCP4.5 and 8.5 scenarios are investigated by considering spatial distributions of 30-year
173 averages of two future periods, mid-century (2040-2069) and late-century (2070-2099), and
174 their differences with the current period of GEM-CAN simulations.

175 The statistical significance of the comparison between the model and observation datasets,
176 and between the historical and transient period is assessed with the two-sample t-test, with the
177 null hypothesis (H_0) given as $\mu_2 = \mu_1$ and the alternative hypothesis (H_1) given as $\mu_2 \neq \mu_1$,

178 where μ_1 and μ_2 are the observations and the model data, or the historical and projected data
179 in the climate projection analysis. We test the null hypothesis at the 95% significance level.

180 **3. Results**

181 This section is divided into two parts: 1) model assessment and climatology of the
182 temperature inversions from model and observations and 2) future projections of temperature
183 inversion characteristics. All results from the GEM-CAN simulations presented in this section
184 are based on averages across the five-member ensemble.

185 ***3.1 Climatology of the Temperature Inversions: Model and Observations***

186 Comparisons of model-simulated 2m temperatures with CRU (land), AIRS (ocean), and
187 ERA-Interim for winter (DJF) and summer (JJA) are shown in Fig. 2. Comparison with CRU
188 data indicates cold biases between -1 and -6 °C over Europe and West Asia for DJF, while
189 warm biases are noted over the northeastern parts of Eurasia. A cold bias is also present over
190 Northeastern North America (NA) and a warm bias over the Western and Northwestern NA.
191 The pattern and values of the bias over North America are similar to those found by Šeparović
192 et al. (2013) when analyzing a GEM simulation over the NA domain. The strong positive bias
193 relative to the CRU temperatures over Northeast Asia and the negative bias over Greenland
194 should be interpreted with caution due to the sparse network of observations over the Arctic
195 region; the bias is not as strong when comparing the model to the ERA-Interim reanalysis, and
196 it is present when comparing the ERA-Interim with the CRU dataset. We can see that
197 differences between the observational datasets can be as large as or larger than those between
198 the model simulation and observations. Over the ocean, GEM-ERA has a similar spatial bias
199 for both AIRS and ERA-Interim, with a negative bias over the central Arctic of -1 to -2 °C
200 (ERA-Interim) and of -2 to -3 °C (AIRS) and positive bias of 1 to 3 °C over the coastal region
201 of Siberia and Alaska. For winter, the cold biases over the Rockies and Greenland are presumed

202 to be related to the representation of topography in the model. Batrak and Müller (2019)
203 suggests that the warm biases in reanalysis and models are due the missing representation of
204 snow layer on top of the sea-ice. For the summer months, the model shows a persistent cold
205 bias of -1 to -6 °C compared to both CRU and ERA-Interim. This cold bias is most prominent
206 in East-Central Asia and the eastern Canadian Arctic. The cold bias in summer over Arctic
207 Canada and mountain regions is presumed to be related to snow on the ground. Over the Arctic
208 Ocean, GEM-ERA shows a negative bias of -1 to -2 °C when compared with AIRS and ERA-
209 Interim. From the differences of the datasets in figure 2, we found that they are significant for
210 most of the domain. These biases notwithstanding, visual inspection demonstrates that the
211 means from each dataset are reasonably similar for most of the domain, and the model shows
212 good capacity in simulating the temperature patterns for each season, with spatial variability
213 similar to the biases. As the temperature inversion is strongly coupled with trends in surface
214 temperature (Zhang et al. 2011), knowing the surface temperature bias can lead to a better
215 understanding of the model simulation of the temperature inversions.

216 The next step in the assessment compares the surface temperature inversion strength and
217 frequency in GEM-ERA with those of the sounding data over land and with those of the AIRS
218 dataset over the ocean, again for both winter and summer. The spatial plots in Fig. 3 show the
219 simulated temperature inversion strength and its frequency and the difference between model
220 values and soundings and AIRS for the Pan-Arctic domain for the 2003–2015 period, with the
221 radiosonde data represented by the filled circles.

222 Radiative cooling of the surface is one of the key processes responsible for surface-based
223 temperature inversion formation, and therefore higher-latitude locations, particularly in
224 Greenland, Alaska, Canada, and Eastern Siberian Russia, experience more frequent and more
225 intense temperature inversions than lower latitude stations in winter, with temperature
226 inversion frequency between 80-100% and temperature inversion strength between 9 and 12°C.

227 Across the radiosonde data, the highest temperature inversion frequency occurs in the Arctic
228 Canada region and Siberia, with values ranging across the same range of values as the GEM-
229 ERA model, between 80-100%, and the highest temperature inversion strength occurs in the
230 interior of Alaska and Siberia, with values up to 9°C in Alaska and up to 12°C in Siberia. The
231 temperature inversions are stronger and more frequent in winter due to the polar night and the
232 weakening of air-surface thermal coupling by snow cover (e.g. Van de Wiel et al., 2017).

233 The patterns for the range of the temperature inversion strength and frequency values of
234 North America (NA) are well represented in the simulation, with the largest winter inversion
235 strength biases of 4°C occurring at a few coastal locations (three on the Arctic Ocean and one
236 on west Greenland). Summer temperature inversion strength bias has a pattern of negative
237 values ranging from -0.5 °C to -2 °C on the west NA and positive bias from 0.5°C to 2°C
238 Nunavut, Quebec, and Greenland regions. Temperature inversion frequency has small
239 differences, up to ±20% in both seasons, with a few outliers on coastal Greenland and the west
240 NA Rockies region. The continental Eurasian patterns also show good agreement for both
241 temperature inversion characteristics. The winter temperature inversion strength shows a
242 positive bias of up to 3 °C in the European region and negative bias up to -4 °C in Siberia.
243 Eurasian summer values do not show a defined pattern, with values ranging from -1.5 °C to 2
244 °C. Over the Arctic Ocean in winter, the GEM-ERA simulation has stronger and more frequent
245 temperature inversions, limited to the region north of NA and the Baffin Bay. Summer
246 produces a small positive bias of up to 1 °C for the temperature inversion strength and up to
247 20% for the temperature inversion frequency over the Arctic Ocean. The model overrepresents
248 the winter and summer frequency of the temperature inversion in the center of the Arctic Ocean,
249 showing up to 50% more inversions in winter and 30% more inversions in the summer.

250 Table 1 shows the radiosonde values and standard deviation with corresponding values
251 from the GEM-ERA simulation for summer and winter at eight locations above latitude 60°

252 selected as the two best and worst results for each positive and negative winter bias of the
253 temperature inversion strength. Results show good agreement between radiosondes and model
254 values, as all model values are within the soundings' one standard deviation range. Overall, the
255 spatial patterns of the temperature inversion strength and frequency from the GEM model are
256 broadly in agreement with the results obtained from the radiosondes and AIRS for both seasons
257 considered, with the largest differences occurring at a few stations in mountainous regions or
258 the Greenland region. The results from the model display patterns and magnitudes that are
259 broadly in accordance with the soundings for the Arctic region, for both seasons.

260 When comparing the winter temperature inversion strength and frequency in GEM-ERA
261 with those in GEM-CAN (Fig. 4), the main differences are over regions where the sea-ice
262 concentration differs in the two simulations (not shown). This difference between the GEM-
263 ERA and GEM-CAN simulations illustrates the strong relationship between sea ice and
264 temperature inversion in higher latitudes. Pavelsky et al. (2011) showed that sea ice is the
265 principal driver of spatial variability in inversion strength in the Arctic Ocean. Besides the
266 difference in regions where the sea-ice differs between simulations, the similarity of the two
267 simulations demonstrates that driving by the GEM model does not increase the biases of
268 simulated temperature inversions.

269 ***3.2 Climate Projections***

270 Fig. 5 and 6 show the winter season current climate (period 1976-2005) and projected
271 changes of temperature inversion strength and frequency in the GEM-CAN simulation for the
272 2040–2069 and 2070–2099 future periods with respect to the 1976–2005 current reference
273 period. Projections for RCP 8.5 indicate that winter season inversions will become weaker and
274 less frequent, with the greatest change over the Arctic Ocean, where the temperature inversion
275 strength is reduced by up to 5°C and 7 °C by the mid and late century, respectively. Over much

276 of the Arctic Ocean, the temperature inversion frequency ranges from 60% to 100% lower for
277 the 2070–2099 period than at present, with reductions of up to 100% in the ice-free regions
278 (Figure S2, supplementary material), meaning that regions without sea-ice loses the main
279 contributor for temperature inversion formation in winter. Mid-century projected changes are
280 milder, with most of the Arctic Ocean experiencing between 40 and 60% fewer temperature
281 inversions, up to a 100% reduction in the Barents Sea.

282 The changes in temperature inversion strength and frequency are consistent with the sea-
283 ice loss and increased cloud cover in winter, with the Arctic Ocean becoming increasingly ice-
284 free and cloudy as warming progresses. Wintertime overall cloud cover increases in all regions
285 north of the 70° latitude, for both RCPs (Figure S3, supplementary material). The increased
286 cloud cover is consistent with the enhanced evaporation rates resulting from reduced sea ice
287 cover (Boisvert and Stroeve, 2015). The reduction of sea-ice also increases the surface sensible
288 heat flux, which increases the air temperature and specific humidity in the lower troposphere,
289 leading to an increase of cloud-cover and downward longwave radiation (Kim et al. 2019),
290 creating a vertical feedback mechanism, reducing the sea-ice even further. In the leads and
291 polynyas (Gultepe et al. 2003), and often over open sea areas of the Arctic, the sensible heat
292 flux is often larger than the latent heat flux. Although the GEM model uses a subgrid-scale tiles
293 approach to calculate the heat fluxes, the SST and sea ice fraction are prescribed from the
294 coarse resolution CanESM2, meaning that fine-scale processes associated with these fields are
295 not represented in the model. Increased ice-free areas in the summertime Arctic Ocean absorb
296 more shortwave energy, resulting in increased latent heat flux later in the year (Morrison et al.,
297 2019). The sensible heat flux also increases, which can lead to reduced cloud formation by
298 heating the air and increasing the saturation specific humidity. In the present GEM simulation
299 results, latent heat fluxes dominate these competing processes, resulting in an increase in cloud
300 cover. As the main factor of strong surface-based temperature inversion formation over the

301 Arctic is radiative cooling, the progressively ice-free ocean, the increase of cloud cover and
302 surface temperature in winter favors a radiative-turbulent near-equilibrium state of the Arctic
303 atmosphere characterized by a cloudy-sky state (e.g. Abraham and Monahan, 2019). The
304 formation of temperature inversions is inhibited, and when temperature inversions occur, they
305 are weaker.

306 The largest changes to temperature inversion properties are limited to the Arctic Ocean and
307 adjacent water-covered regions. Only the simulations using RCP 8.5 show large spatial changes
308 over continental areas, with less frequent and weaker inversions. The largest changes of
309 temperature inversion strength over land are seen in the Canadian Arctic Archipelago and
310 Northern Siberia for both future periods considered. The changes of temperature inversion
311 frequency over the continents follow broadly the same pattern as that of temperature inversion
312 strength, with changes in the inversion strength decreasing with increasing distance from the
313 ocean. Lawrence et al. (2008) showed that the warming signal due to rapid sea-ice loss can
314 penetrate inland, leading to terrestrial snow cover reduction changing the surface fluxes and
315 planetary boundary layer characteristics (Alexander et al. 2010; Deser et al. 2010) that affect
316 temperature inversion formation.

317 The RCP 4.5 simulations show similar patterns as RCP 8.5, but with less pronounced
318 changes. For both the temperature inversion strength and frequency, the late 21st century for
319 RCP 4.5 has similar spatial patterns and magnitudes of projected changes as the RCP 8.5
320 simulation in mid-century. The similarity is consistent with the corresponding RCPs warming
321 signals due to increased CO₂ concentrations. For RCP 4.5, CO₂ stabilizes at ~500 ppm in the
322 late 21st century. This value is reached by RCP 8.5 around mid-century. In the RCP 4.5
323 simulation, the late 21st century shows a 50% decrease of inversion frequency and inversions
324 up to 5 °C weaker. Mid-century RCP 4.5 has temperature inversions up to 4°C weaker with a
325 30% reduction in temperature inversion frequency. The regions with complete loss of

326 wintertime sea-ice in the Arctic Ocean region show a 100% reduction of the inversion
327 frequency in both periods.

328 Summer season values for the current climate and projected changes for the temperature
329 inversion strength and frequency are shown in Fig. 7 and 8, respectively. For both RCPs,
330 projected changes in the summer season show increased temperature inversion strength and
331 frequency over the Arctic Ocean, except for the ocean regions of the Barents Sea and the
332 Norwegian Sea. These regions show temperature inversions up to 1.5 °C weaker, and up to
333 40% reduction in temperature inversion frequency, in both RCPs. The late 21st century in RCP
334 8.5 has up to 50% increased inversion frequency over the Arctic Ocean with inversion strength
335 1.5°C stronger. Except for the Ellesmere Island in the Canadian Arctic Archipelago, Arctic
336 coastal regions show up to 40% fewer inversions, up to 2 °C weaker, with the southern region
337 of the Canadian Arctic Archipelago having the most pronounced changes. The mid-21st century
338 shows similar patterns as the late century, with weaker changes. These relatively large changes
339 in temperature inversion frequency occur in regions in which summer temperature inversions
340 are relatively common (> 40%) in the present climate. The work of Wang et al. (2022) relates
341 positive anomalies of the summer temperature inversions in the Arctic to the sea-ice minimum
342 extremes in September. With the increased frequency of such temperature inversions in the
343 transient projections, the effect on the sea-ice minima is presumed to increase. The increase in
344 frequency and strength of summer temperature inversions can also make Arctic fog events
345 more common, as inversions factor in fog formation (Gilson et al. 2018).

346 Similar to the winter results, RCP 4.5 has the same pattern of changes as the RCP 8.5 for
347 the inversion strength and frequency, but with smaller changes of both properties. The changes
348 for the mid-century and late 21st century is similar, with the late century showing only slightly
349 higher values. This can be explained by the stabilization of the climate in RCP 4.5 late 21st
350 century.

351 The increase in strength and frequency of temperature inversions in summer is likely related
352 to the increase in moist static energy transport in the Arctic, increasing the frequency of
353 temperature inversion formation by advection of warm air masses over a cooler surface. Hwang
354 et al. (2011) show an increase in poleward moisture transport in climate projections driven by
355 the increased moisture as the climate warms. Tjernström et al. (2019) discuss the increased
356 atmospheric water content in the Arctic due to the increased intrusion of warmer air masses,
357 also causing strong surface temperature inversions due to subsidence in the summer months.
358 They hypothesize that as the warm air mass is advected over the melting sea-ice, it triggers the
359 formation of a sharp surface temperature inversion and fog. As the surface is cold, even low
360 humidity in the advected air results in supersaturation. As mechanical surface mixing and
361 cloud-top buoyant mixing act on this system, the Arctic boundary layer gradually becomes
362 more well mixed. The fog and the well-mixed layer deepen, the cloud lifts from the surface
363 and the Arctic boundary layer transforms into a persistent well-mixed cloud-capped boundary
364 layer.

365 The results discussed above show the changes in temperature inversion characteristics for
366 summer and winter. In winter, the main mechanism of formation is surface long-wave radiation
367 cooling. Reduction in sea ice cover not only results in the loss of a cold surface but allows for
368 increased moisture flux (Serreze et al., 2009; Screen and Simmonds, 2010a; Screen and
369 Simmonds, 2010b;). The increased wintertime cloud cover reduces the surface longwave
370 cooling rate, greatly diminishing the temperature inversion formation by radiative cooling. For
371 summer, subsidence and the transport of southern air masses likely drive the formation of
372 temperature inversions.

373 **4. Summary and Conclusions**

374 In this study, climate projections of large-scale Arctic temperature inversion characteristics
375 were investigated using an ensemble of the GEM climate model simulations for RCPs 4.5 and
376 8.5. This study includes the assessment of the model performance and characterization of the
377 spatial distribution of the temperature inversion strength and frequency in summer and winter
378 focusing on the large-scale structure of the temperature inversions, as the model cannot capture
379 the fine-scale characteristics of the lower atmosphere.

380 The model slightly overestimates the temperature inversion strength and frequency in the
381 current climate, with some coastal regions reaching 4°C difference in the temperature inversion
382 strength, and up to 20% difference in the temperature frequency, consistent with the fact that
383 numerical models have difficulty in representing the stable boundary layer, particularly in the
384 Arctic (Fernando and Weil, 2010; Holtslag et al., 2013). For both summer and winter, the
385 results from the model show patterns and magnitudes that are in overall accordance with
386 soundings and the AIRS satellite data in the Arctic region. The spatial variability of the Arctic
387 temperature inversions is strongly influenced by the surface type. In winter, ice-free zones in
388 the North Atlantic have a low frequency of temperature inversions, in contrast with the sea ice
389 covered ocean and snow-covered regions where occurrence frequency exceeds 80%. Summer
390 values of the temperature inversion frequency are higher for near-coastal regions (up to 70%
391 frequency), in contrast to the central Arctic (40% frequency). Analyses of the transient climate
392 change simulations indicate that the loss of sea ice and related increase in cloud cover in winter
393 leads to weaker and less frequent temperature inversions. This reduction in the occurrence of
394 very strongly stratified conditions is consistent with the reduction in surface radiative cooling
395 resulting from increased cloudiness. The increase of cloud cover is associated with an increased
396 surface moisture flux during the winter seasons after the mostly ice-free sea absorbed energy
397 in the summer. It is important to note that changes in cloud cover are model-dependent. Some
398 climate models predict increased autumn cloud cover in the central Arctic (Liu et al. 2012;

399 Vavrus et al. 2012), while other models predict a decrease (Koenigk et al. 2013), which can
400 change the outcome of temperature inversions in a future climate. Therefore, additional studies
401 should be made with different RCMs and driving GCMs on the area, due to variations in spatial
402 patterns from different driving data and climate scenarios. For ocean regions in the summer,
403 there is an increase in temperature inversion frequency and strength at the end of the 21st
404 century, consistent with an increase in poleward moist energy transport.

405 With the changes in the Arctic atmosphere seen in the future climate, the current
406 predominant clear cold atmospheric state becomes less frequent. Cloudy and warm conditions
407 that inhibit the formation of temperature inversions become more common. For the winter
408 season, the reduction of sea ice and the increased cloud cover prevents the radiative cooling
409 necessary to the formation of strong temperature inversions, and for the summer, the increased
410 poleward moisture transport likely due to increased synoptic-scale warm air advection over the
411 cooler ice-free ocean causes a higher incidence of temperature inversions.

412 These results provide information about how large-scale temperature inversions are
413 represented in climate models and about the effects of a warming climate in the Arctic planetary
414 boundary layer, as changes in temperature inversion characteristics have important
415 implications for the Arctic climate system. For example, temperature inversions mediate the
416 surface energy balance (Bradley et al. 1992; Medeiros et al. 2011), restrict the vertical exchange
417 of momentum, heat, moisture, and pollutants between PBL and the free atmosphere (Wetzel
418 and Brümmer, 2011) and are an important contributor to Arctic Amplification (Bintanja et al.
419 2011). In this study, we analyzed Arctic temperature inversions looking at relatively coarse
420 structures of the planetary boundary layer. Future work benefiting higher vertical resolution
421 and better representation of near-surface processes could focus on simulating the changes of
422 the more detailed vertical structure of the atmosphere.

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438 **Availability of data and material:** The data presented in this study are available on
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440 **Code availability:** The code used to generate the plots and to analyze the data can be
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442 **Authors' contributions:** C. J. Ruman analyzed the data and wrote the manuscript with
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TABLES

689 **Table 1: Subset of radiosonde values plus standard deviation for the temperature inversion strength (ΔT) and the**
 690 **temperature inversion frequency (FREQ), with corresponding values from the GEM-ERA simulation for summer and**
 691 **winter.**

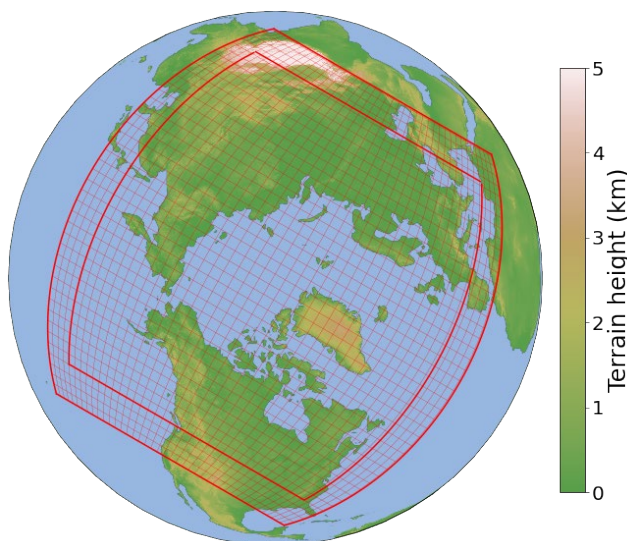
Location (lat, lon)	DJF				JJA			
	Soundings		GEM-ERA		Soundings		GEM-ERA	
	ΔT (°C)	FREQ (%)	ΔT (°C)	FREQ (%)	ΔT (°C)	FREQ (%)	ΔT (°C)	FREQ (%)
ZYRYANKA (65.7, 150.9)	11.8±6.6	87	7.5±5.2	60	1.8±1.5	13	2.6±2.0	19
SEJMCHAN (62.9, 152.4)	9.4±5.1	79	5.2±4.3	51	3.2±2.2	25	3.8±3.1	29
SODANKYLA (67.4, 26.6)	7.9±6.1	53	7.5±5.2	66	2.7±1.9	21	1.8±1.5	8
OSTROV DIKSON (73.5, -80.4)	6.3±3.7	78	5.9±3.9	59	3.1±2.4	32	3.2±2.2	41
ARHANGEL'SK (64.6, 40.5)	5.1±3.8	44	5.2±3.8	42	2.6±1.8	23	1.4±1.2	8
BAKER LAKE (64.3, -96.1)	7.8±4.2	90	8.1±5.5	75	2.8±2.2	20	2.1±1.8	20
INUVIK (68.3, -133.5)	7.4±4.9	80	10.7±7.6	74	2.3±2	25	2.3±1.8	22
ALERT (82.5, -62.3)	5.8±4.0	82	9.5±5.9	84	3.0±2.1	43	4.5±3.1	52

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FIGURES

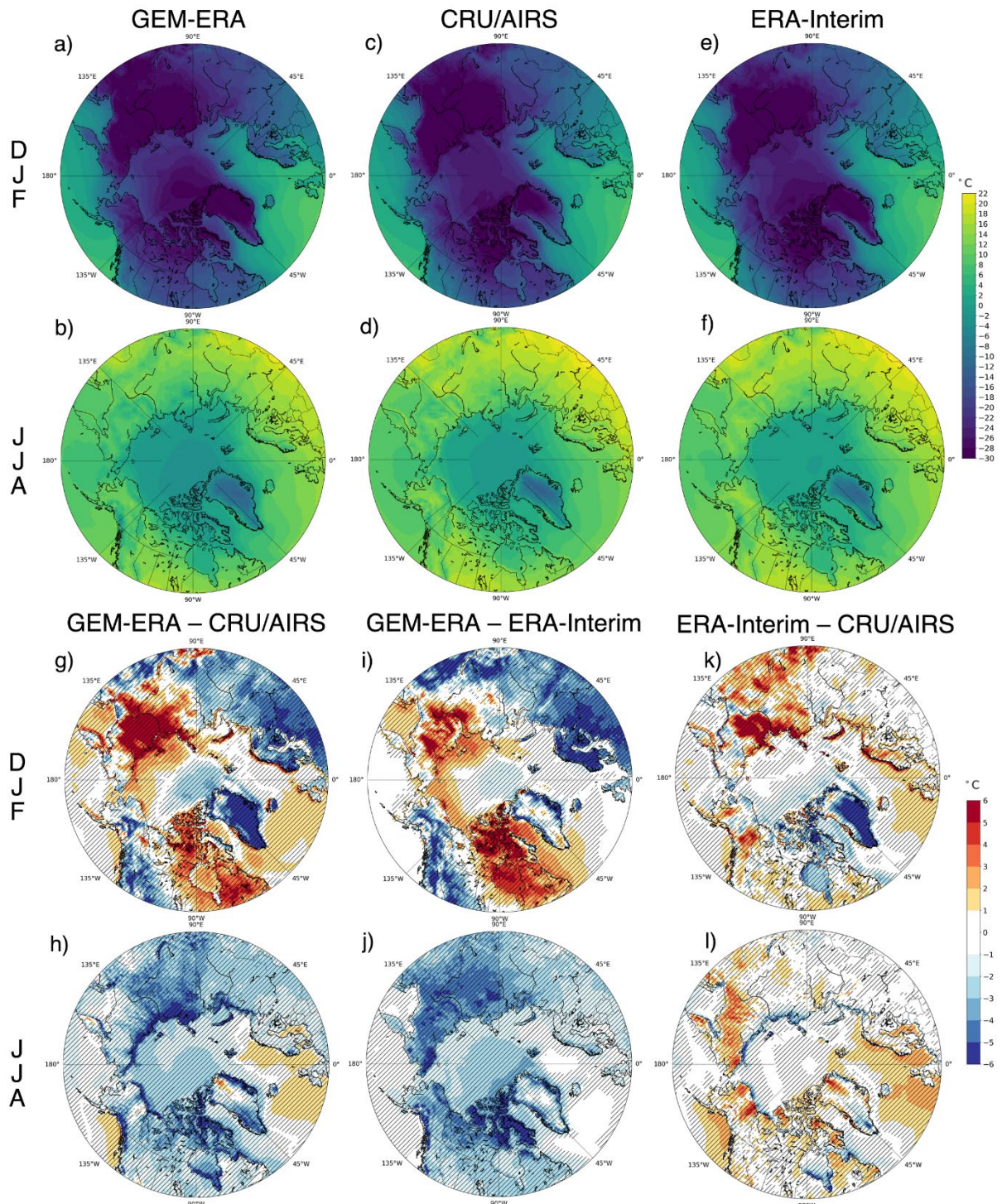
694 **Figures should include full captions. Appendix figures should be included here following**
 695 **any main manuscript figures (if any).**



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Fig. 1 The Pan-Arctic experimental domain, with every fifth grid point shown. The inner bold line separates the blending and free zones. The topography is shown in color.



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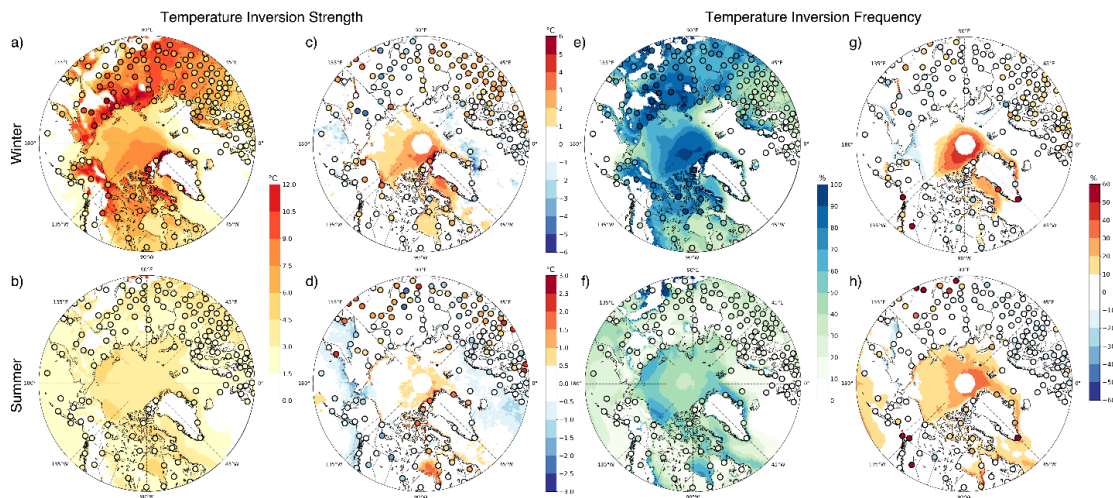
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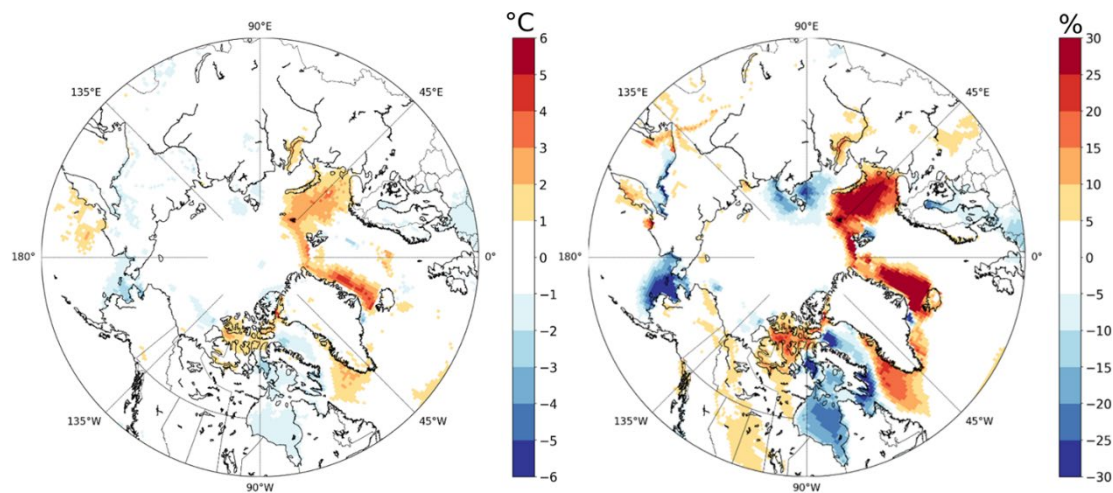
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Fig. 2 Spatial plots showing 2m temperatures for the 2003-2015 period for: GEM-ERA simulated climatological winter (a) and summer (b); CRU (land) and AIRS (ocean) winter (c) and summer (d); ERA-Interim winter (e) and summer (f). Differences between GEM-ERA and CRU/AIRS 2m temperatures for winter (g) and summer (h), and

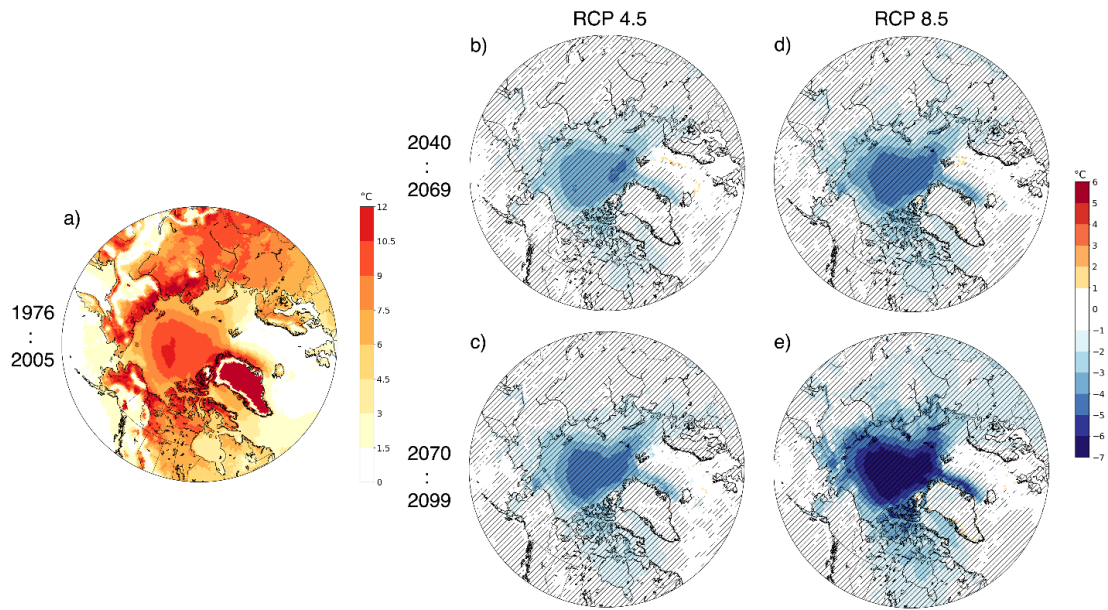
703 differences between GEM-ERA and ERA-Interim for winter (i) and summer (j). Hatched regions show where
 704 projected changes are statistically significant with the two-sample t-test at the 5% significance level.



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 706 Fig. 3 GEM-ERA simulated winter (a) and summer (b) temperature inversion strength. Winter (c) and summer
 707 (d) differences between GEM-ERA and ARIS (ocean) and atmospheric soundings (circles over land) for the
 708 temperature inversion strength. e)-(h): same as (a)-(d) for the temperature inversion frequency.



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 710 Fig. 4 Difference between GEM-ERA and GEM-CAN simulations in DJF for temperature inversion strength
 711 (left column) and inversion frequency (right column for the 1980-2005 period)



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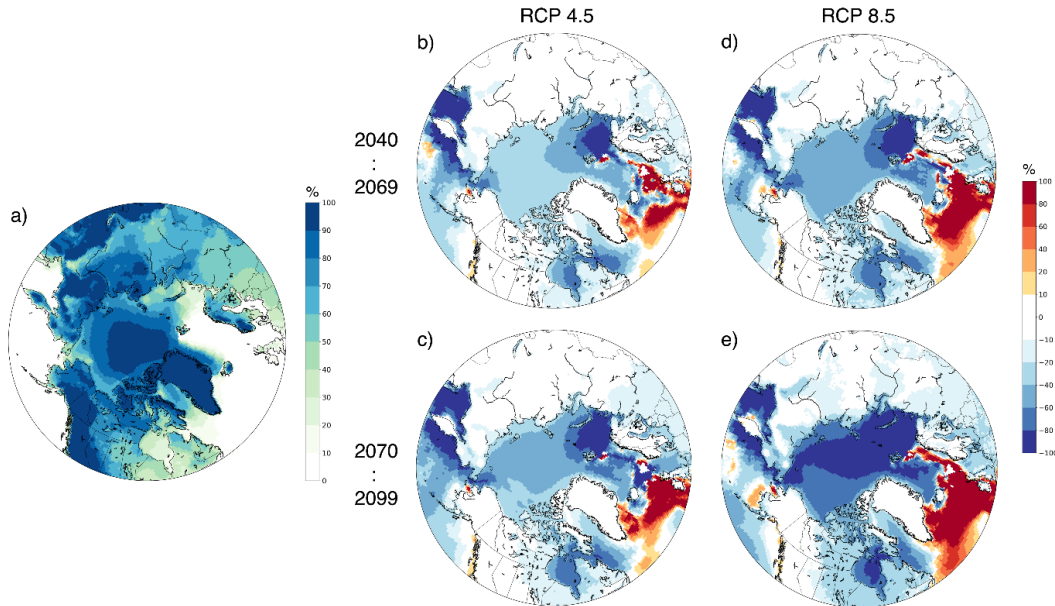
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Fig. 5 GEM-CAN simulated temperature inversion strength for DJF (a) and RCP 4.5 projected changes to DJF temperature inversion strength for the 2040–2069 (b) and 2070–2099 (c) future periods with respect to the current 1976–2005 period. (d)-(e): same as (b) and (c) but for RCP 8.5. Hatched regions show where projected changes are statistically significant with the two-sample t-test at the 5% significance level.



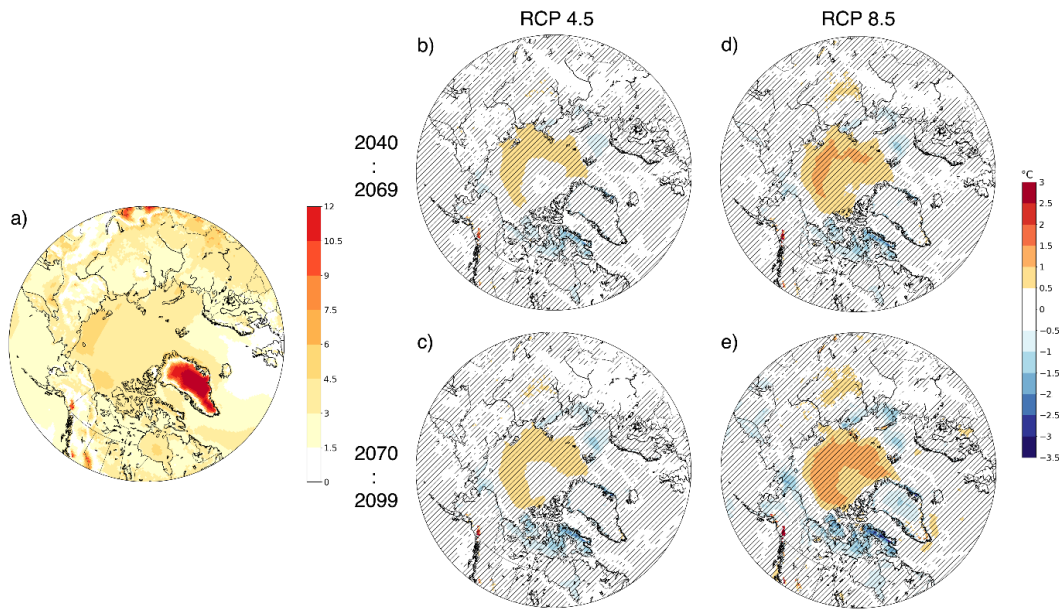
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Fig. 6 GEM-CAN simulated temperature inversion frequency for DJF (a) and RCP 4.5 projected changes to DJF temperature inversion strength for the 2040–2069 (b) and 2070–2099 (c) future periods with respect to the current 1976–2005 period. (d)-(e): same as (b) and (c) but for RCP 8.5.



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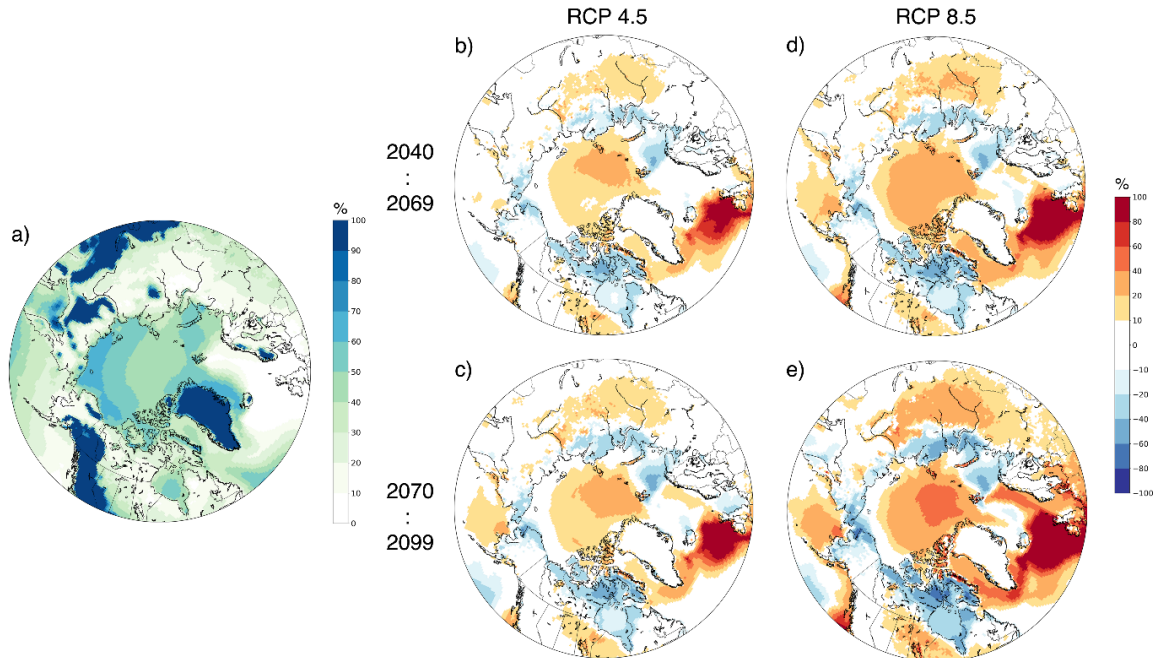
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Fig. 7 GEM-CAN simulated temperature inversion strength for JJA (a) and RCP 4.5 projected changes to JJA temperature inversion strength for the 2040–2069 (b) and 2070–2099 (c) future periods with respect to the current 1976–2005 period. (d)-(e): same as (b) and (c) but for RCP 8.5. Hatched regions show where projected changes are statistically significant with the two-sample t-test at the 5% significance level.



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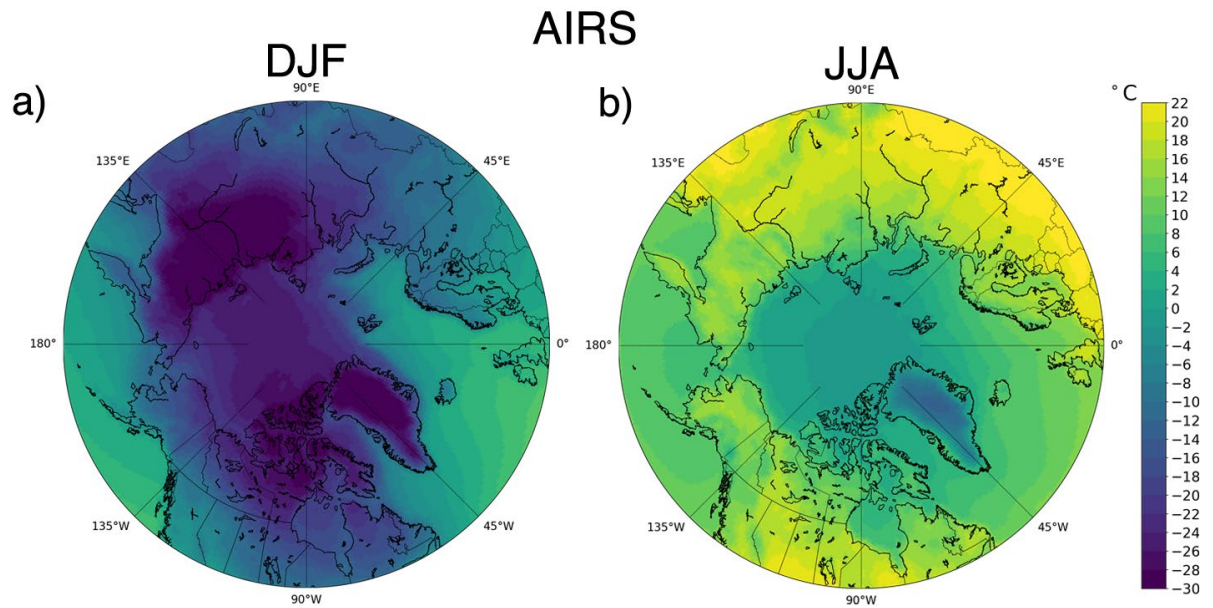
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Fig. 8 GEM-CAN simulated temperature inversion frequency for JJA (a) and RCP 4.5 projected changes to JJA temperature inversion frequency for the 2040–2069 (b) and 2070–2099 (c) future periods with respect to the current 1976–2005 period. (d)-(e): same as (b) and (c) but for RCP 8.5.

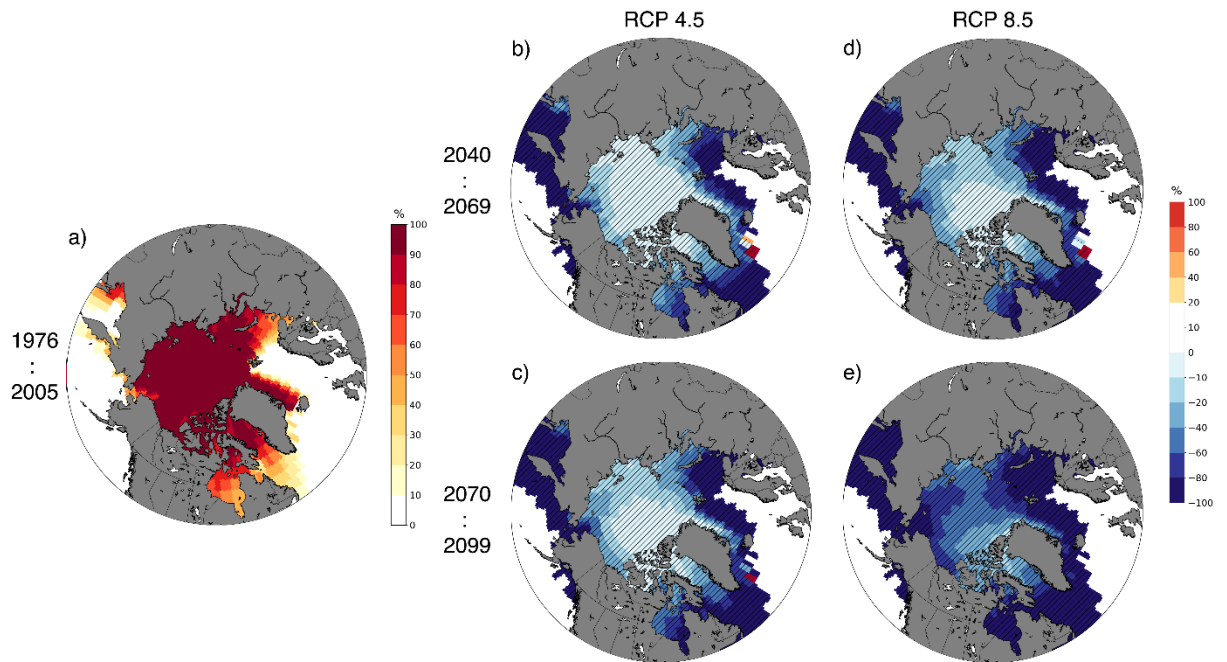
Supplementary Figures



12

13 Fig. S1: Spatial plots showing 2m temperatures for the 2003-2015 period for the AIRS dataset for the winter (a)

14 and summer (b) months.

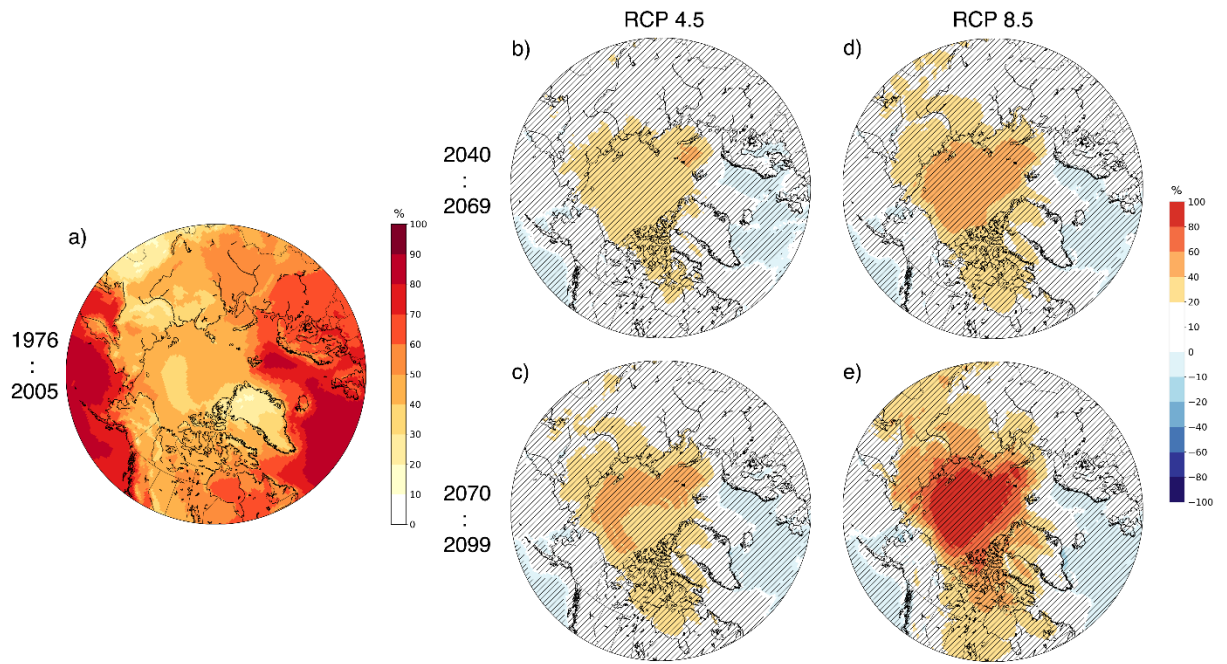


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16 Fig. S2: GEM-CAN simulated sea-ice for DJF (a) and RCP 4.5 projected changes to DJF sea-ice for the 2040-

17 2069 (b) and 2070-2099 (c) future periods with respect to the current 1976-2005 period. (d)-(e): same as (b) and (c)

18 but for RCP 8.5.



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Fig. S3: GEM-CAN simulated total cloud cover for DJF (a) and RCP 4.5 projected changes to DJF total cloud for the 2040–2069 (b) and 2070–2099 (c) future periods with respect to the current 1976–2005 period. (d)–(e): same as (b) and (c) but for RCP 8.5.