

1 **Impact of grids and dynamical cores in CESM2.2 on**
2 **the surface mass balance of the Greenland Ice Sheet**

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7 **Key Points:**

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Abstract

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Plain Language Summary

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1 Introduction

General Circulation Models (GCMs) are powerful tools for understanding the meteorology and climate of the high-latitudes, which are among the most sensitive regions on Earth to global and environmental change. Despite their importance, the numerical treatment of polar regions in GCMs is handled in vastly-different ways due to the so-called *pole-problem* (D. Williamson, 2007). The pole problem refers to numerical instability arising from the convergence of meridian lines into polar singularities on latitude-longitude grids (e.g., Figure 1a). Depending on the numerics, methods exist to suppress this instability, and latitude-longitude grids may be advantageous for polar processes as structures can be represented with more degrees of freedom than elsewhere in the computational domain. With the recent trend towards globally uniform unstructured grids, any potential benefits of latitude-longitude grids on polar regions may become a relic of the past. In this study a number of grids and dynamical cores (hereafter referred to as *dyccores*) available in the Community Earth System Model, version 2.2 (CESM; <http://www.cesm.ucar.edu/models/cesm2/>), including brand new variable-resolution grids, are evaluated to understand their impacts on the simulated characteristics of the Arctic, with a special focus on the climate and surface mass balance of the Greenland Ice Sheet.

In the 1970's the pole problem was largely defeated through wide-spread adoption of efficient spectral transform methods in GCMs. These methods transform grid point fields into a global, isotropic representation in wave space, where linear operators (e.g. horizontal derivatives) in the equation set can be solved for exactly. While spectral transform methods are still used in the 21st century, local numerical methods have become desirable for their ability to run efficiently on massively parallel systems. The pole problem has thus re-emerged in contemporary climate models that use latitude-longitude grids, and some combination of reduced grids and polar filters are necessary to ameliorate this instability (Jablonowski & Williamson, 2011). Polar filters are akin to a band-aid; they subdue the growth of unstable modes by applying additional damping to the solution over polar regions. This additional damping reduces the effective resolution in polar regions, and the resolved scales are approximately the same everywhere on the grid.

An alternative approach is to use unstructured grids, which allow for more flexible grid structures that permit quasi-uniform grid spacing globally and eliminates the pole-problem entirely (e.g., Figure 1c). This grid flexibility also permits variable-resolution or regional grid refinement (e.g., Figure 2). Grids can be developed with refinement over polar regions that could in principle make up for any loss in polar resolution in transitioning away from latitude-longitude grids (e.g., Figure 2), although this comes at the cost of a smaller CFL-limiting time-step in the refined region (the CFL-condition — short for Courant–Friedrichs–Lewy condition — is a necessary condition for numerical stability when using discrete data in time and space). Unstructured grids also scale more efficiently on parallel systems than latitude-longitude grids, likely resulting in a greater prevalence of unstructured grids as computing power continued to increase over time.

The meteorology and climate of the Arctic is characterized by a range of processes and scales that are difficult to represent in GCMs (Bromwich et al., 2001; Smirnova & Golubkin, 2017; van Kampenhout et al., 2018). For example, while synoptic scale storms are generally well represented at typical GCM resolutions of 1 to 2 degrees (Jablonowski

& Williamson, 2006; Stocker, 2014), mesoscale Polar Lows are not well resolved at these resolutions. These mesoscale systems are prevalent during the cold season and produce gale-force winds that can induce large heat and moisture fluxes through the underlying sea-ice/ocean interface. The Arctic also contains the Greenland Ice Sheet (hereafter denoted as *GrIS*). While it blankets the largest island in the world (Greenland), many of the processes that control the *GrIS* annual surface mass balance (the integrated sum of precipitation and runoff) are only partially resolved at typical GCM resolutions. For example, *GrIS* precipitation is typically confined to the ice-sheet margins (predominately the southeastern margin) where orographic precipitation is generated by steep topographic slopes. *GrIS* ablation areas (marginal regions where seasonal melting exceeds the annual mass input from precipitation) are typically 10s to 100 km wide and confined to low-level areas or regions with limited precipitation. GCMs struggle to resolve the magnitude and extent of these features (van Kampenhout et al., 2018), which can lead to unrealistic ice sheet growth in models with an interactive ice sheet component (e.g., Lofverstrom et al., 2020).

The goal of this study is to characterize the representation of high-latitude regions using the spectral-element and finite-volume dycores in CESM2.2, as these models treat the high-latitudes, e.g., the pole-problem, in different ways. The manuscript is laid out as follows: Section 2 consists of documentation of the grids, dycores and physical parameterizations used in this study. The Arctic refined grids were developed by the authors, and this section serves as their official documentation in CESM2.2. Section 2 also contains a description of the experiments along with reanalysis datasets and post-processing software used for evaluating the model simulations. Section 3 contains the results of the experiments, followed by Section 4 that provides a general discussion and conclusions.

2 Methods

2.1 Dynamical cores

The atmospheric component of CESM2.2, the Community Atmosphere Model, version 6.3 (CAM; Craig et al., 2021), supports a number of different atmospheric dynamical cores. These include dycores using latitude-longitude grids, such as finite-volume (FV; Lin, 2004) and eulerian spectral transform (EUL; Collins et al., 2006) models, and dycores built on unstructured grids, including spectral-element (SE; Lauritzen et al., 2018) and finite-volume 3 (FV3; Putman & Lin, 2007) models. The EUL dycore is the oldest dycore in CAM, and the least supported of all the dycores. FV3 is the newest dycore in CAM, but it was not fully incorporated at the time this work commenced; both the EUL and FV3 dycores are omitted from this study. As such, the results presented in this study are comparing the performance of the SE and FV dycores.

2.1.1 Finite-volume (FV) dynamical core

The FV dycore is a hydrostatic model that integrates the equations of motion using a finite-volume discretization on a spherical latitude-longitude grid (Lin & Rood, 1997). The 2D dynamics evolve in floating Lagrangian layers that are periodically mapped to Eulerian reference grid in the vertical (Lin, 2004), using a hybrid-pressure vertical coordinate. Hyperviscous damping is applied to the divergent modes while Laplacian damping is applied to momentum in the top few layers, referred to as a *sponge layer* (Lauritzen et al., 2011). A polar filter is used to avoid computational instability due to the convergence of the meridians, allowing for a more practical time-step. It takes the form of a Fourier filter in the zonal direction, with the damping coefficients increasing monotonically in the poleward direction (Suarez & Takacs, 1995).

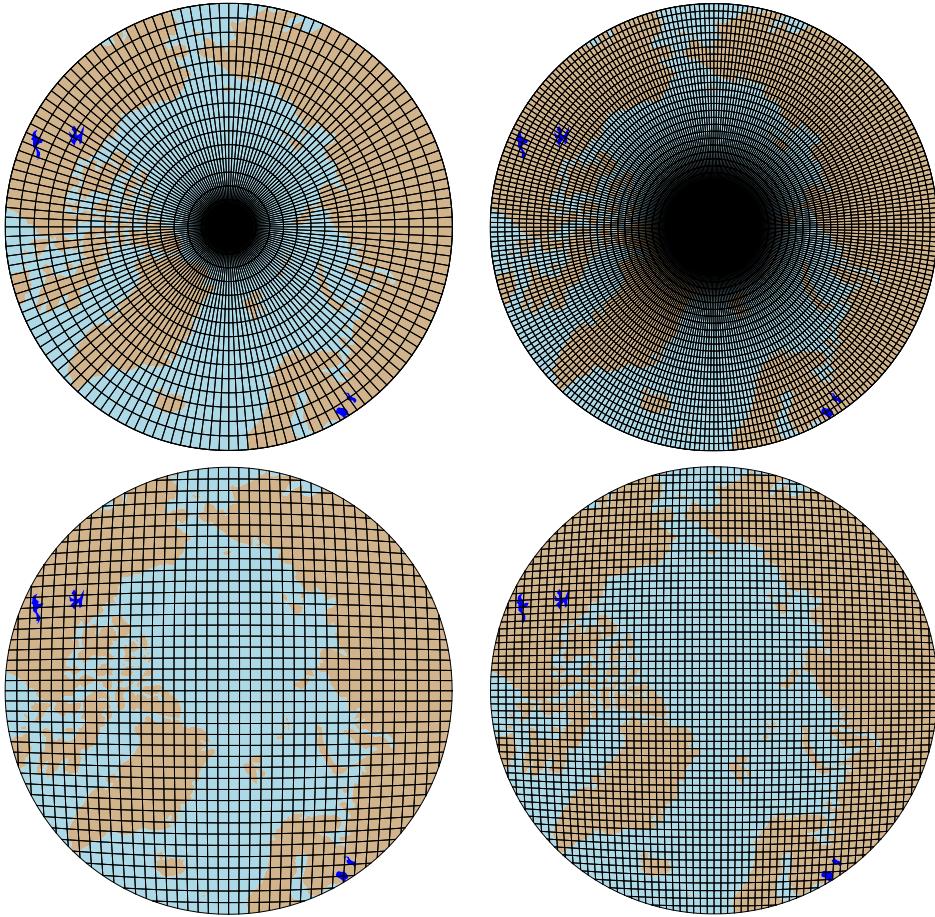


Figure 1. Computational grids for the uniform $1^\circ - 2^\circ$ grids in this study.

106 **2.1.2 Spectral-element (SE) dynamical core**

107 The SE dycore is a hydrostatic model that integrates the equations of motion us-
 108 ing a high-order continuous Galerkin method (Taylor et al., 1997; Taylor & Fournier, 2010).
 109 The computational domain is a cubed-sphere grid tiled with quadrilateral elements (e.g.,
 110 Figure 2). Each element contains a fourth order basis set in each horizontal direction,
 111 with the solution defined at the roots of the basis functions, the Gauss-Lobatto-Legendre
 112 (GLL) quadrature points. This results in 16 GLL nodal points within each element, with
 113 12 of the points lying on the (shared) element boundary. Communication between el-
 114 ements happens via the direct stiffness summation (Canuto et al., 2007), which applies
 115 a numerical flux to the element boundaries that reconciles overlapping nodal values and
 116 produces a continuous global basis set.

117 As with the FV dycore, the dynamics evolve in floating Lagrangian layers that are
 118 subsequently mapped to an Eulerian reference grid. A dry mass vertical coordinate was
 119 more recently implemented for thermodynamic consistency with condensates (Lauritzen
 120 et al., 2018). The 2D dynamics have no implicit dissipation and so hyperviscosity op-
 121 erators are applied to all prognostic variables to remove spurious numerical errors (Dennis
 122 et al., 2012). Laplacian damping is applied in the sponge layer.

123 The SE dycore supports regional grid refinement via its variable-resolution config-
 124 uration, requiring two enhancements over uniform resolution grids. (1) As the numer-



Figure 2. Spectral-element grid for the variable-resolution ARCTIC grid in this study. Note that this is not the computational grid; each element has 3×3 independent grid points.

ical viscosity increases with resolution, explicit hyperviscosity relaxes according to the local element size, reducing in strength by an order of magnitude per halving of the grid spacing. A tensor-hyperviscosity formulation is used (Guba et al., 2014), which adjusts the coefficients in two orthogonal directions to more accurately target highly distorted quadrilateral elements. (2) The topography boundary conditions need to be smoothed in a way that does not excite grid scale modes, and so the NCAR topography software (Lauritzen et al., 2015) has been modified to scale the smoothing radius by the local element size.

For spectral-element grids with quasi-uniform grid spacing, a variant in which tracer advection is computed using the Conservative Semi-Lagrangian Multi-tracer transport scheme (CSLAM) is used instead (Lauritzen et al., 2017). CSLAM has improved tracer property preservation and accelerated multi-tracer transport. It uses a separate grid from the spectral-element dynamics, through dividing each element into 3×3 control volumes with quasi-equal area. The physical parameterizations are computed from the state on the CSLAM grid, which has clear advantages over the default SE dycore in which the physics are evaluated at the GLL nodal points (A. Herrington et al., 2018).

2.2 Grids

Six grid are evaluated in this study (Table 1). The FV dycore is run with 1° and 2° grid spacing, referred to as $f09$ and $f19$, respectively (Figure 1a,b). The 1° equivalent of the CAM-SE-CSLAM grid is also run, referred to as $ne30pg3$ (Figure 1c), where ne refers to a grid with of $ne \times ne$ elements per cubed-sphere face, and pg denotes that there are $pg \times pg$ control volumes per element for computing the physics. An additional 1° CAM-SE-CSLAM grid is run, but with the physical parameterizations computed on a grid that contains 2×2 control volumes per element, $ne30pg2$ (Figure 1d; A. R. Herrington et al., 2019).

Two variable resolution meshes were developed as part of the CESM2.2 release that contains grid refinement over the Arctic (Figure 2). The Arctic meshes were developed using the software package SQuadgen (<https://github.com/ClimateGlobalChange/squadgen>). The *ARCTIC* grid is a 1° grid with $\frac{1}{4}^\circ$ regional refinement over the broader Arctic region. The *ARCTICGRIS* grid is identical to the *ARCTIC* grid, but contains an additional patch covering the big island of Greenland with $\frac{1}{8}^\circ$ resolution.

grid name	dycore	Δx_{eq} (km)	Δx_{refine} (km)	Δt_{phys} (s)
<i>f19</i>	FV	278	-	1800
<i>f09</i>	FV	139	-	1800
<i>ne30pg2</i>	SE-CSLAM	167	-	1800
<i>ne30pg3</i>	SE-CSLAM	111	-	1800
<i>ARCTIC</i>	SE	111	28	450
<i>ARCTCIGRIS</i>	SE	111	14	225

Table 1. Grids and dycores used in this study. Δx_{eq} refers to average equatorial grid spacing, Δx_{refine} refers to grid spacing in the refined region (if applicable) and Δt_{phys} refers to the physics time-step. The dycore abbreviation FV refers to the finite-volume dycore, SE the spectral-element dycore and SE-CSLAM the spectral-element dycore w/ CSLAM tracer advection.

156 2.3 Physical parameterizations

157 The CAM6 physical parameterization package (hereafter referred to as the *physics*;
 158 <https://ncar.github.io/CAM/doc/build/html/index.html>) is used in all simulations
 159 in this study. CAM6 physics is most noteably different from it's predecessors through
 160 the incorporation of high-order turbulence closure, Cloud Layers Unified by Binormals
 161 (CLUBB; Golaz et al., 2002; Bogenschutz et al., 2013), which jointly acts as a PBL, shal-
 162 low convection and cloud macrophysics scheme. CLUBB is coupled with the MG2 mi-
 163 crophysics scheme (Gettelman et al., 2015), with prognostic precipitation and classical
 164 nucleation theory in representing cloud ice for improved cloud-aerosol interactions. Deep
 165 convection is parameterized using a convective quasi-equilibrium mass flux scheme (Zhang
 166 & McFarlane, 1995; Neale et al., 2008) inclunding convective momentum transport (Richter
 167 et al., 2010). PBL form drag is modeled after (Beljaars et al., 2004) and orographic grav-
 168 ity wave drag is represented with an anisotropic method informed by the orientation of
 169 topographic ridges at the sub-grid scale.

170 All grids and dycores in this study use 32 levels in the vertical, with a model top
 171 of about 1 hPa or about 40 km. The physics time-step is dependent on grid resolution.
 172 Increases in horizontal resolution permit faster vertical velocities that reduce character-
 173 istic time-scales, and so the physics time-step is reduced to avoid large time truncation
 174 errors (A. Herrington & Reed, 2018). The *ARCTIC* and *ARCTCIGRIS* grids are there-
 175 fore run with a 4× and 8× reduction in physics time-step relative to the default 1800
 176 s time-step used in coarser, uniform resolution grids (Table 1).

177 Initial simulations with the *ne30pg3* spectral-element grid produced weaker short-
 178 wave cloud forcing relative to the tuned up finite-volume dycore. All runs with the spectral-
 179 element dycore have two CLUBB parameter changes in order to provide a more realis-
 180 tic cloud forcing and top-of-atmosphere radiation balance. These are CLUBB's *gamma*
 181 parameter, reduced from 0.308 to 0.270, and *c14*, reduced from 2.2 to 1.6. Briefly, the
 182 *gamma* parameter scales the width of the sub-grid distribution of vertical velocity, and
 183 *c14* controls the strength of the damping term in the equation for the horizontal com-
 184 ponent of turbulent kinetic energy. For a thorough explanation of how CLUBB param-
 185 eters impact the simulated climate, the reader is referred to (Guo et al., 2015).

186 2.4 Experimental design

187 All grids and dycores are run using an identical transient 1979-1998 AMIP-style
 188 configuration, with prescribed monthly SST/sea-ice after (Hurrell et al., 2008). This con-
 189 figuration refers to the *FHIST compset* and runs out of the box in CESM2.2.

data product	years used in this study	citation
ERA5	1979-1998	(Copernicus, 2019)
CERES-EBAF ED4.1	2003-2020	(Loeb et al., 2018)
CALIPSO-GOCCP	2006-2017	(Chepfer et al., 2010)
RACMO2.3	1979-1998	(Noël et al., 2015)
RACMO2.3p2	1979-1998	(Noël et al., 2019)
LIVVkit,v2	1979-1998	(Evans et al., 2019)

Table 2. Description of observational datasets used in this study.

190 The surface mass balance (SMB) of the Greenland Ice Sheet (GrIS) is simulated
 191 in all grids and dycores in this study. The SMB is the sum of the mass source term, ac-
 192 cumulation (i.e., precipitation), and the mass sink term, ablation. Ablation can be ex-
 193 pressed as evaporation/sublimation plus total runoff, with runoff being a combination
 194 of liquid precipitation and snow and ice melt. Not all liquid precipitation becomes runs
 195 off the ice sheet; rain may penetrate pore spaces in the firn layer and freeze, forming ice
 196 lenses in the subsurface. These processes are represented by different components in CESM,
 197 but it is the Community Land Model, version 5 (CLM; Lawrence et al., 2019), that ag-
 198gregates these processes and computes the SMB.

199 CLM runs on the same grid as the atmosphere, but also uses a downscaling tech-
 200 nique to account for sub-grid variability in SMB. In short, the ice sheet patch in a CLM
 201 grid cell is subdivided into 10 elevation classes (EC), weighted by their respective area
 202 fractions at each EC, which is derived from a high resolution GrIS elevation dataset. The
 203 near surface air temperature, humidity and air density are calculated at each EC using
 204 an assumed lapse rate and the elevation difference from the grid mean, and the precip-
 205 itation rates from CAM are repartitioned into solid or liquid based on the temperature
 206 of the EC. Ice accumulation is modeled as a capping flux, or snow in excess of a 10 m
 207 snow cap, and refreezing of liquid within the snowpack additionally acts as a source of
 208 ice. A unique surface energy balance and SMB is computed for each EC. Integrating over
 209 all ECs using the area weights provides a more accurate SMB. For a more detailed de-
 210 scription of how the SMB is computed in CESM, the reader is referred to (Lipscomb et
 211 al., 2013; Sellevold et al., 2019; van Kampenhout et al., 2020).

212 Since the 10 m snowcap needs to be reached in the accumulation zone to simulate
 213 the SMB, the snow depths in the variable-resolution grids were spun-up by forcing CLM
 214 in standalone mode, cycling over a 20 year *ARCTIC FHIST* run for about 500 years.
 215 The uniform resolution grids are all initialized with an SMB from an existing *f09* spun-
 216 up initial condition.

217 2.5 Observational Datasets

218 Several observational datasets are used in this study to understand the performance
 219 of the simulations. A list of the datasets used in this study are shown in Table 2. Sev-
 220 eral of these products (ERA5, CALIPSO and CERES) are near-global gridded datasets
 221 commonly used to evaluate GCMs. Surface mass balance datasets are gathered from sev-
 222 eral sources. RACMO2.3 11km and RACMO2.3p2 5.5km are regional model simulations
 223 targeting Greenland, forced by ERA interim and ERA5 reanalysis at its domain bound-
 224 aries. The RACMO simulations have been shown to performs very skillfully against ob-
 225 servations and is therefore considered an ideal modeling target (Noël et al., 2015, 2019).
 226 The Land Ice Verification and Validation toolkit (LIVVkit), version 2.1 (Evans et al.,
 227 2019) maintains a repository of snow pit and and ice core SMB measurements, as well
 228 as the IceBridge radar accumulation dataset. The LIVVkit dataset is compared against
 229 model simulations by finding the nearest grid cell center to the location of each obser-
 230 vation.

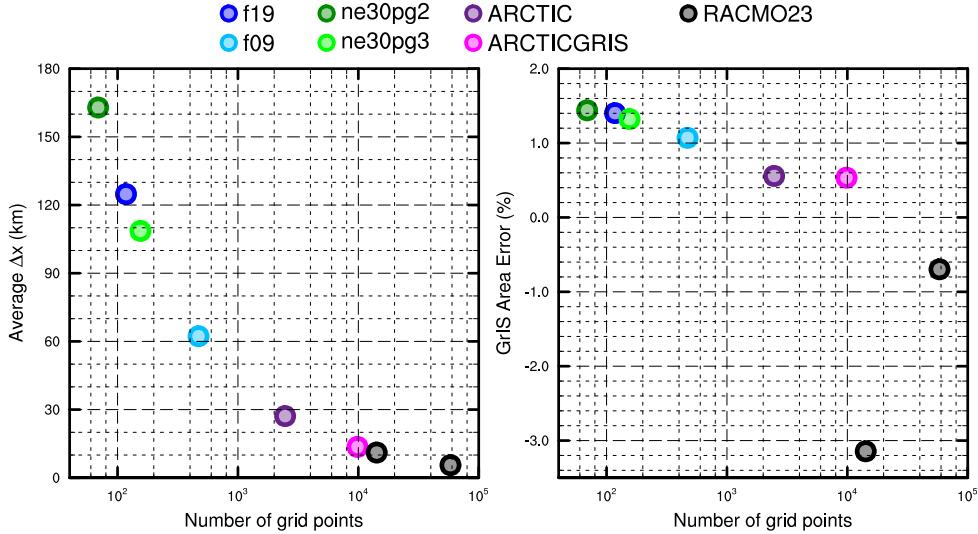


Figure 3. The spatial properties of the GrIS as represented by different grids in this study. (Left) approximate average grid spacing over GrIS, (right) GrIS area error, computed as the relative differences from a 4km dataset used to create the CESM ice masks.

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2.6 SMB Analysis

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A common high resolution dataset is used to generate the GrIS boundary conditions in all grids, using the CLM dataset creation tools. Since we are interested in the total ice sheet SMB, we seek to integrate various components of the SMB over a common ice mask to get the total mass change of the GrIS. Figure 3 shows the GrIS ice mask area across the different grids, as a function of the number of grid points. Due to the use conservative regridding in the CLM tools, the interpolation errors are small and ice mask areas have less than 1.5% errors relative to the raw ice mask dataset. RACMO2.3, however, uses a smaller ice mask, about 3% smaller than the raw ice mask dataset.

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Figure 3 suggests integrating quantities over the native ice mask of the six grids would probably not suffer from large errors due to differing ice masks, but we seek to compare these integrated quantities to RACMO2.3. Therefore, we have taken the approach of mapping all model fields to the lowest resolution grids and integrating over the respective low resolution ice masks. Due to the sensitivity of mapping errors to grid coordinates (i.e., unstructured or structured), all quantities are evaluated on both the *f19* and *ne30pg2*, the lowest resolution grid for both dycores considered in this study. In addition, two remapping algorithms are used; ESMF conservative and TempestRemap high-order, monotone algorithm. In all, each integrated quantity is evaluated (at most) four times to provide an estimate of uncertainty due to differences in grid coordinates and remapping algorithm.

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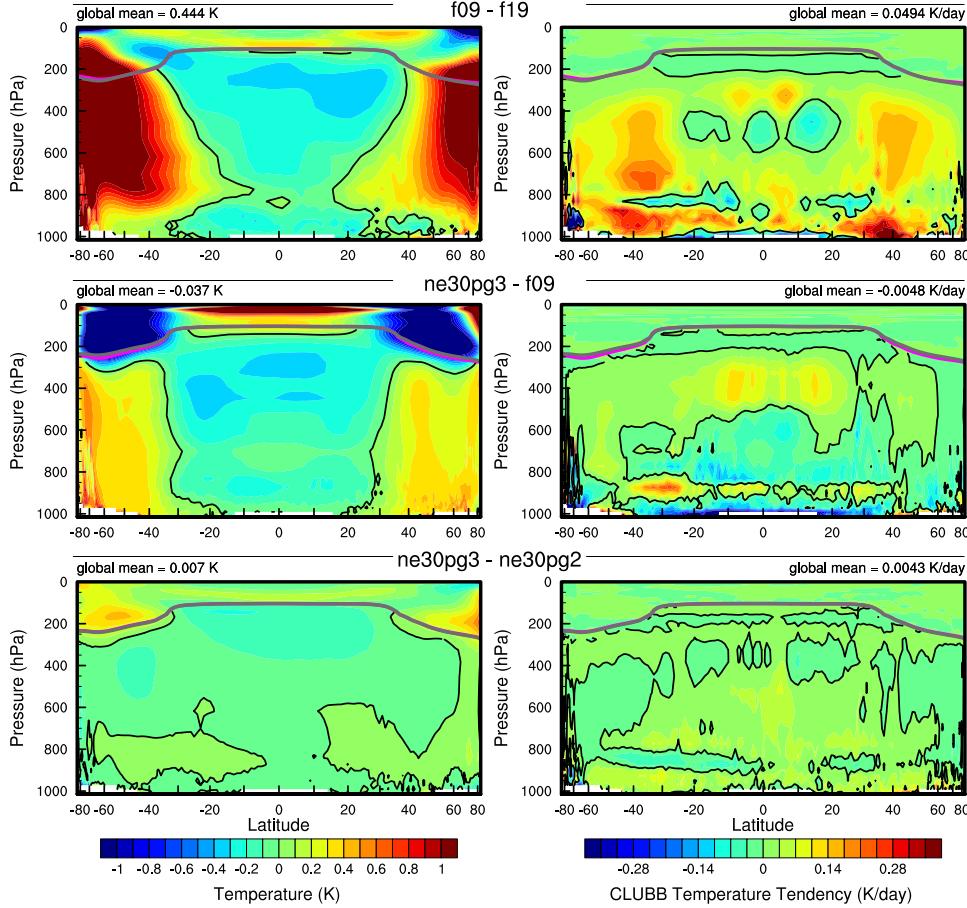
3 Results

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3.1 Tropospheric temperatures

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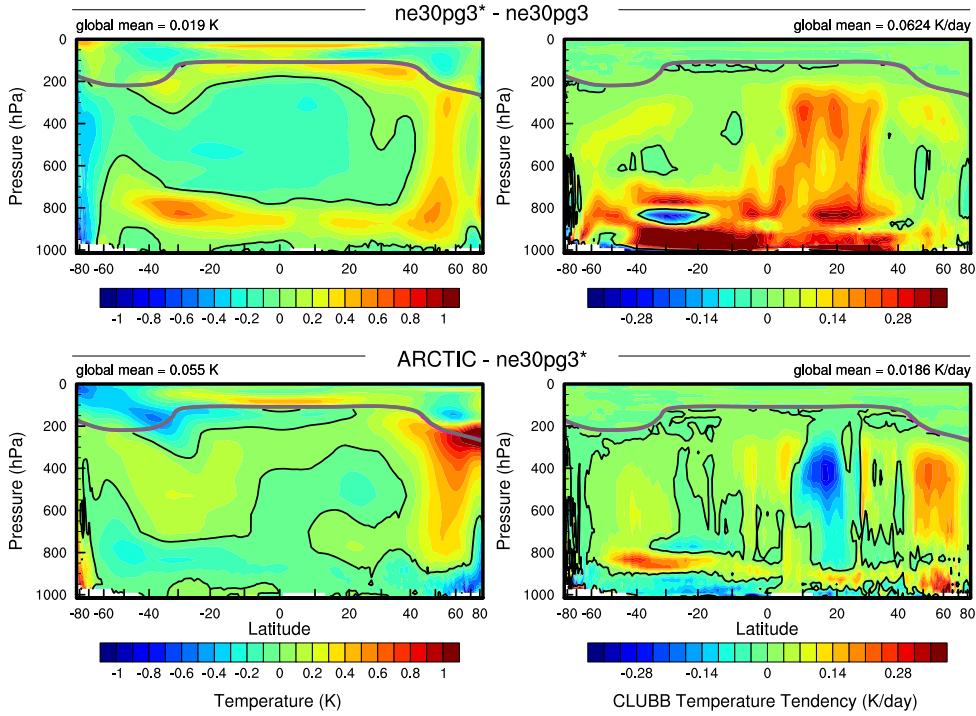
Before delving into the simulated characteristics of the Arctic, the global mean differences between the various grids and dycores are assessed. Figure 4 shows 1979-1998 annual mean, zonal mean height plots expressed as differences between the uniform resolution grids and dycores. The *f09* grid is warmer than the *f19* grid, primarily in the

**Figure 4.**

mid-to-high latitudes and throughout the depth of the troposphere. This is a common response to increasing horizontal resolution in GCMs (Pope & Stratton, 2002; Roeckner et al., 2006), and (A. R. Herrington & Reed, 2020) has shown that this occurs in CAM due to greater resolved vertical velocities that in turn, facilitate greater condensational heating in the macrophyics routine in CLUBB. The right columns in Figure 4 supports this interpretation, which shows an increase in the climatological CLUBB heating in the low and mid-latitudes in the *f09* grid.

As the SE dycore is less diffusive than the FV dycore, the resolved vertical velocities are larger in the SE dycore, and so a modest, resolution-like sensitivity occurs in which *ne30pg3* is warmer than *f09* (Figure 4). The stratosphere has a different response, in which *ne30pg3* is much cooler than *f09* in the mid-to-high latitudes. Figure 4 also shows differences in temperature between *ne30pg3* and *ne30pg2*, which are small, although there is a slight warming near the tropopause at high latitudes. This is consistent with the similar climates found between these grids in (A. R. Herrington et al., 2019).

Comparing the variable-resolution grids to the uniform resolution grids is complicated because we simultaneously increase the grid resolution and reduce the physics time-step, both of which noticeably impact the solution (D. L. Williamson, 2008). An additional *ne30pg3* simulation is run with the physics time-step used in the *ARCTIC* grid, referred to as *ne30pg3**. Figure 5 shows the change in the climatological summer temperatures

**Figure 5.**

in zonal-mean height space between *ne30pg3** and *ne30pg3*. A similar warming response to increasing resolution occurs when the time-step is reduced, and the mechanism is similar in that the shorter time-step facilitates greater condensational heating by CLUBB. Figure 5 shows the difference in climatological summer temperature between the *ARCTIC* grid and the *ne30pg3** grid. The greater condensational heating and warmer temperatures are confined to the regionally refined region when the impact of physics time-steps is removed from the analysis.

It's useful to understand summer temperature biases, instead of annual means, due to its control on ice/snow melt (Ohmura, 2001; Huybers & Tziperman, 2008). Figure 6 shows the 1979-1998 lower troposphere summer temperature bias relative to ERA5. It is computed from the 500 hPa-1000 hPa geopotential thickness, solving for the layer mean virtual temperature using the hypsometric equation. The results generally track with the analysis of the zonal mean height plots; increasing resolution from *f19* to *f09* leads to a warmer climate, and the 1° spectral-elements grids are warmer than the finite-volume grids. The summer temperatures in the finite-volume grids are persistently colder than ERA5 at high latitudes, whereas the 1° spectral-element grids are warmer than ERA5 at only very high-latitudes, north of 80° . All grid illustrate a north-south gradient in bias over Greenland, in which the summer temperature bias becomes more positive in the northward direction. This pattern is also evident in the 2m summer temperature bias over Greenland (not shown).

The *ARCTIC* grid has similar summer temperatures to the 1° spectral-element grids, but it is a bit warmer over northern Eurasia and the North Pole. An anomalous cooling patch forms to the west of Greenland, centered over Baffin Island. The *ARCTICGRIS* grid is warmer than the *ARCTIC* grid over most of the Arctic region, but approximately maintains the same pattern of summer temperature bias as in the *ARCTIC* grid.

Some of these temperature anomalies may be related to summer shortwave cloud forcing differences across the different grids and dycores. Figure 7 shows the summer shortwave cloud forcing bias in the runs, using the CERES-EBAF product. All the uniform 1° – 2° grids have similar biases, with the clouds reflecting 20-40 W/m 2 too much shortwave radiation over a wide swath of the Arctic, primarily over the land masses. There's also a halo of low cloud forcing bias around the oceanic perimeter of Greenland. The *ARCTIC* grid has much smaller cloud forcing biases over the Arctic land masses, although still too reflective, whereas the *ARCTCGRIS* grid vastly improves the cloud forcing bias over Eurasia, and improves the bias over N.America compared to the *ARCTIC* grid. In both variable-resolution grids, the halo of too weak cloud forcing bias around the perimeter of Greenland is absent.

While the summer cloud forcing biases are consistent with the summer temperature biases in Figure 6 –regions where clouds are too reflective coincide with regions that are too cold– it is not clear whether the cold biases are caused by the cloud biases, or whether the cold biases amplify the cloud forcing bias.

3.2 Shortwave radiation over Greenland

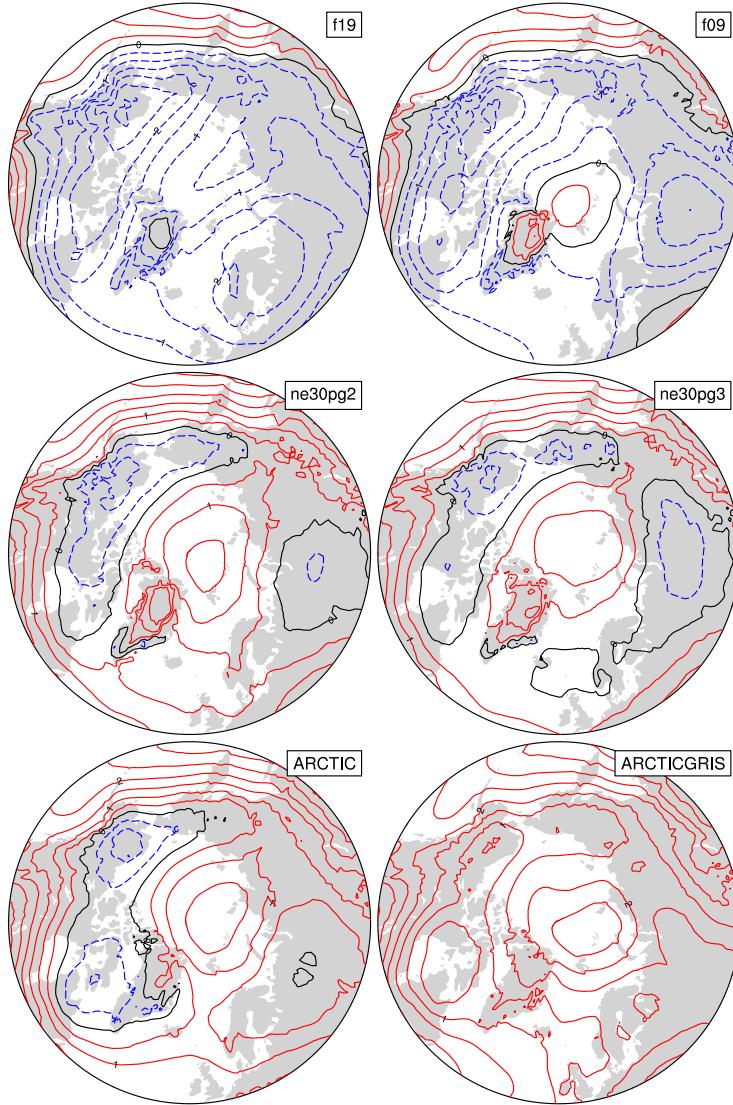
In addition to summer temperatures, shortwave radiation is also an important determinant of snow/ice melt. Figure 8 shows the summer incident shortwave radiation bias at the surface, zoomed in over Greenland. The top panel computes the bias using the CERES-EBAF dataset, and the bottom panel using RACMO2.3p2 dataset. This halo of excessive incident shortwave radiation around the coasts of Greenland is apparent in both datasets, consistent with the shortwave cloud forcing biases in Figure 7.

The interior of the ice sheet receives too little shortwave radiation in the coarser grids. In the variable-resolution grids, both the interior deficit in shortwave and the excessive shortwave around the oceanic perimeter of Greenland are improved. This suggests that the coarse grids clouds are too thick in the interior of Greenland, and too thin around the perimeter of Greenland, and that increasing horizontal resolution balances out these biases. This is consistent with total summer cloud fraction bias, computed from the CALIPSO cloud dataset (Figure 9). Note that total cloud fraction characterizes the cloud field at all vertical levels, but attenuates any changes arising from any single layer due to the maximum overlap assumption used to compute this quantity. Despite the attenuated signal, the total cloud fraction does indicate a reduction in cloud coverage in the interior, and an increase in cloudiness about the oceanic perimeter, in the variable-resolution grids.

The agreement of the cloud biases over Greenland in multiple independent datasets indicates this is a robust feature of the coarser grids. The reduction of these cloud biases in the variable-resolution grids suggests they are a result of insufficient horizontal resolution in coarse grids.

3.3 Greenland surface mass balance

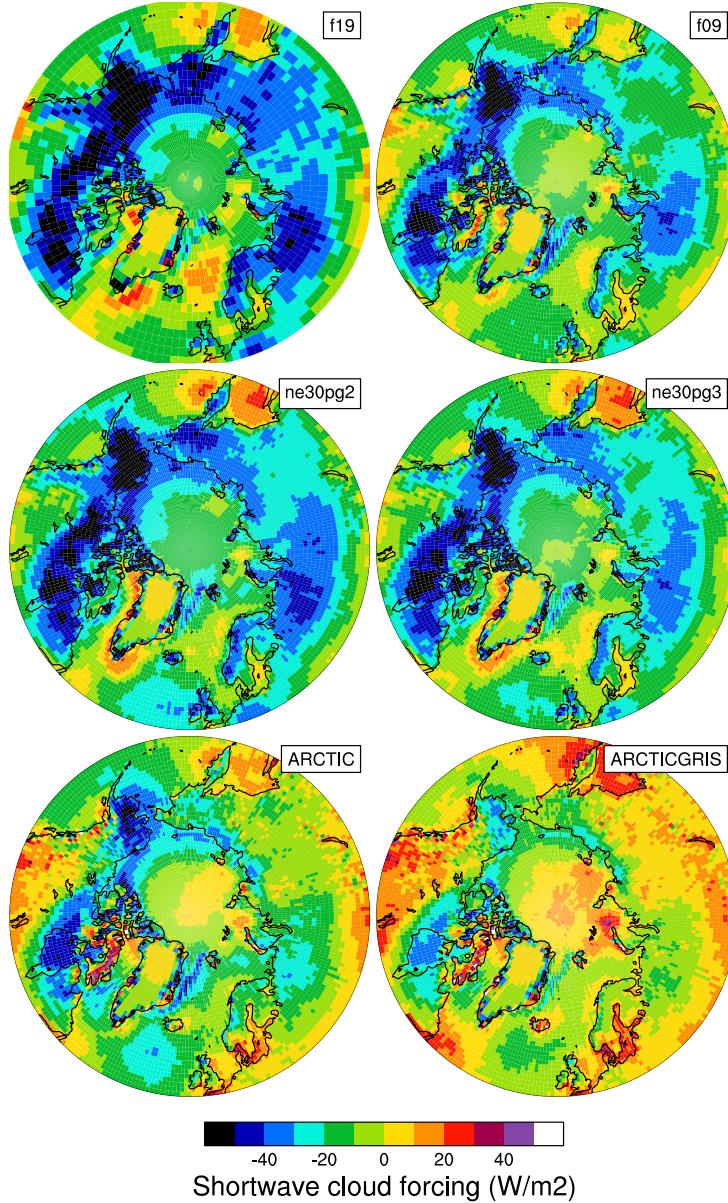
The accuracy of the simulated SMB is expected to be sensitive to grid resolution. Figure 9 shows the average grid spacing over the Greenland Ice Sheet (GrIS) in all six grids in this study. The *ne30pg2* grid has the coarsest representation with an average $\Delta x = 160\ km$, and the *ARCTCGRIS* grid has the highest resolution with an average $\Delta x = 14.6\ km$, similar to the grid spacing of the $11\ km$ RACMO2.3 grid. The *ne30pg3* grid has an average $\Delta x = 111.2\ km$, which is substantially more coarse than the *f09* grid, with an average $\Delta x = 60\ km$. This is interesting because *ne30pg3* and *f09* have similar average grid spacing over the entire globe, and comparable computational costs, but due to the convergence of meridians the finite-volume model has enhanced resolution over GrIS. The *ARCTIC* grid has an average grid spacing of $\Delta x = 27.8\ km$, and

**Figure 6.** .

is about 10 times more expensive than the 1° models (whereas the *ARCTICGRIS* grid is about twice as expensive as the *ARCTIC* grid).

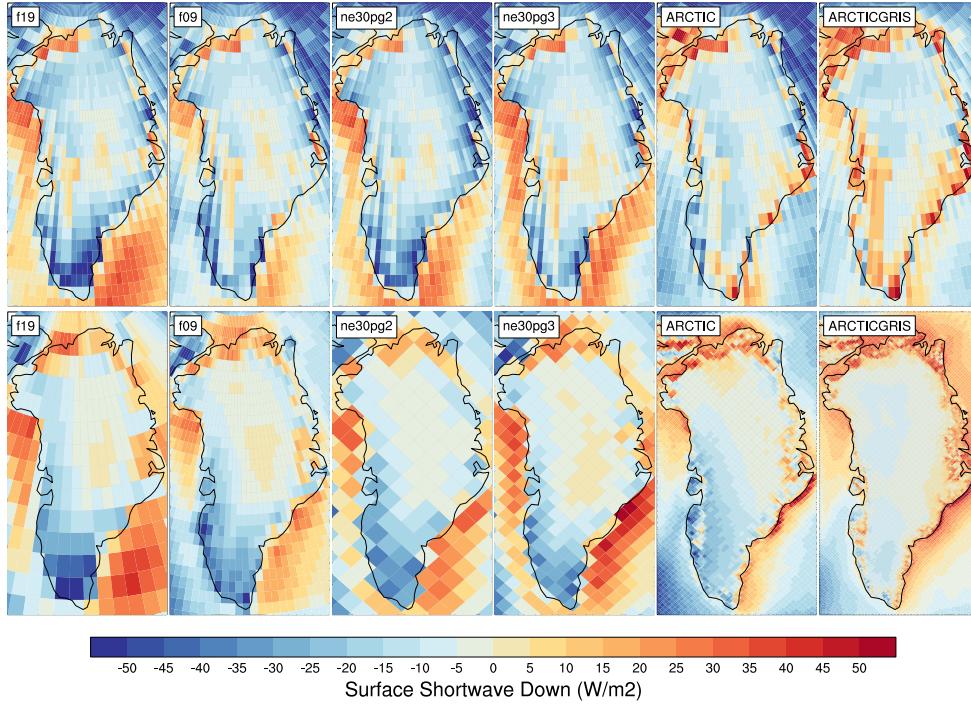
The summer climatological mean precipitation bias over GrIS is shown in Figure 3, expressed as the fractional difference from the RACMO2.3p2 solution. What sticks out is that the coarse $1^\circ - 2^\circ$ grids have large, positive biases centered over the southern dome. The *ARCTIC* run improves this bias substantially, and the *ARCTICGRIS* run improves this bias further. This suggests the southern dome bias is due to inadequate horizontal resolution, which is consistent with the original GrIS variable-resolution experiments in (van Kampenhout et al., 2018).

Southeast Greenland has the largest accumulation rates in GrIS due to synoptic systems moving in from the southeast. These systems are orographically lifted by the steep southeast ice sheet margin, dumping large amounts of precipitation along the southeast coast. At lower horizontal resolutions, the topography is too smooth and large amounts

**Figure 7.**

of moisture penetrates further inland, incorrectly dumping precipitation onto the interior of the ice sheet. A similar bias occurs in northwest Greenland, in particular during the summer, when it's common for synoptic systems to arrive from the southwest. The ability of the variable-resolution grids to more accurately simulate the orographic precipitation process in Greenland is consistent with all the cloud results up to this point. Since the precipitation centers move from the interior towards the coasts, and even out over the ocean with increasing resolution, the cloud decks should, and are, moved accordingly, eliminating this halo of low cloud bias around the oceanic perimeter of Greenland.

Figure 10 shows time-series of annually integrated precipitation and snow/ice melt over the GrIS in the various different grids and dycores, with both versions of RACMO shown in black. The 1979-1998 climatological mean values are provided as circles on the

**Figure 8.** .

right side of the panels. The uniform $1^\circ - 2^\circ$ grids all show a distinctive positive bias in precipitation due to this over-prediction of interior precipitation rates. The variable-resolution grids have the smallest precipitation biases, providing a comparable solution to RACMO. The *f19* and *f09* perform similarly, with +110 Gt/yr bias, whereas *ne30pg3* is biased by about +165 Gt/yr and *ne30pg2*, +230 Gt/yr. The results suggest that uniform resolution spectral-element grids have larger biases than the finite-volume grids, consistent with spectral-element grids having a coarser representation of GrIS (Figure 3).

The combined annual snow/ice melt integrated over the GrIS is given by the bottom panel of Figure 10. The *ARCTIC* grid simulates the most realistic melt rates, with all other grids tending to have larger melt rates than RACMO. The *ARCTICGRIS* grid over predicts melting by about 125 Gt/yr. This is likely due to an anomalously warm lower troposphere during the summer, relative to the *ARCTIC* run (Figure 6). The *f19* and *f09* melting rates are improved over *ARCTICGRIS*, simulating too much melt by about 70-90 Gt/yr. The spectral-element grids have the largest positive melt bias, between 200-220 Gt/yr. It is more difficult to attribute these differences to resolution alone, since the finite-volumes grids have colder summer temperatures than the uniform resolution spectral-element grids. However, that the *ARCTCIGRIS* grid has the warmest summer temperatures, yet has a lower melting bias than the uniform spectral-element grids, suggests that increasing resolution improves the ablation process.

Figure 11 shows the distribution of point-wise differences from LIVVkit observational database as violin plots. The IceBridge dataset is exclusively from the interior of the ice sheet and represents accumulation rates. The uniform $1^\circ - 2^\circ$ grids have similar median errors of about +35-50 mm.w.e, while the variable-resolution errors are noticeably smaller. The in-situ observations in the accumulation zone, shown in the middle plot looks very similar to the IceBridge errors, providing confidence that the variable-resolution grids are outperforming the uniform grids in the interior accumulation zone.

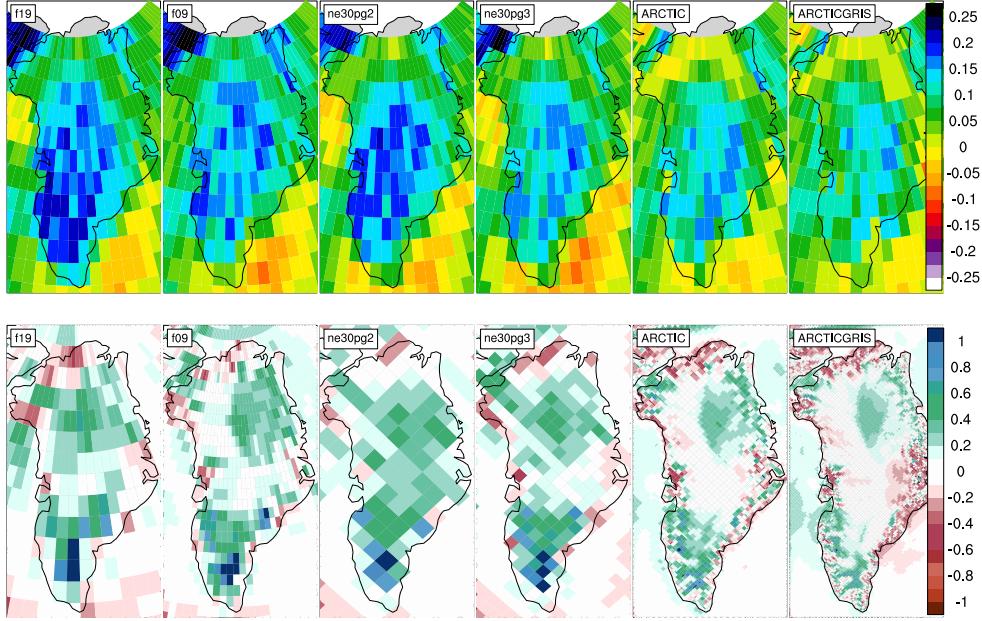


Figure 9. Climatological (1979-1998) annual precipitation rate bias (in mm/yr) relative to the RACMO2.3p2 5.5km resolution data product (Noël et al., 2019).

grid name	accumulation (Gt/yr)	ice/snow melt (Gt/yr)
<i>RACMO</i>	768.5	-347.2
<i>f19</i>	882.5	-440.3
<i>f09</i>	874.8	-418.4
<i>ne30pg2</i>	1000.	-549.4
<i>ne30pg3</i>	934.9	-568.8
<i>ARCTIC</i>	795.9	-367.3
<i>ARCTICGRIS</i>	708.7	-471.6

Table 3. 1979-1998 Surface Mass Balance of the Greenland Ice Sheet.

The errors evaluated at in-situ ablation zone measurements are in tension with the RACMO results in Figure 10. They indicate that the uniform 1° – 2° grids perform similarly, and that the *ARCTICGRIS* performs best, and an improvement over the *ARCTIC* grid. However, the in-situ ablation measurements are sparse in time and space, and so there is a large amount of uncertainty in extending these results to the entirety of the ablation zone.

The “k-transect” is perhaps the most well studied transect in Greenland, as it has been regularly monitored since the 1950s. The LIVVkit has compiled the k-transect observations, shown in Figure 12 as elevation vs. SMB along the transect, with the model’s simulated transects shown as well. The in-situ observation points are nicely replicated by the *ARCTICGRIS* run, whereas the *ARCTIC* grid is biased positive in the higher-elevations of the ablation zone. The *f09* grid is surprisingly competitive with the variable-resolution grids, capturing a realistic slope of the elevation-SMB curve, although it exhibits a similar positive bias as the *ARCTIC* grid. The elevation-SMB slopes of the uniform spectral-element and *f19* grid are too shallow, in particular at the higher elevation regions of the transect.

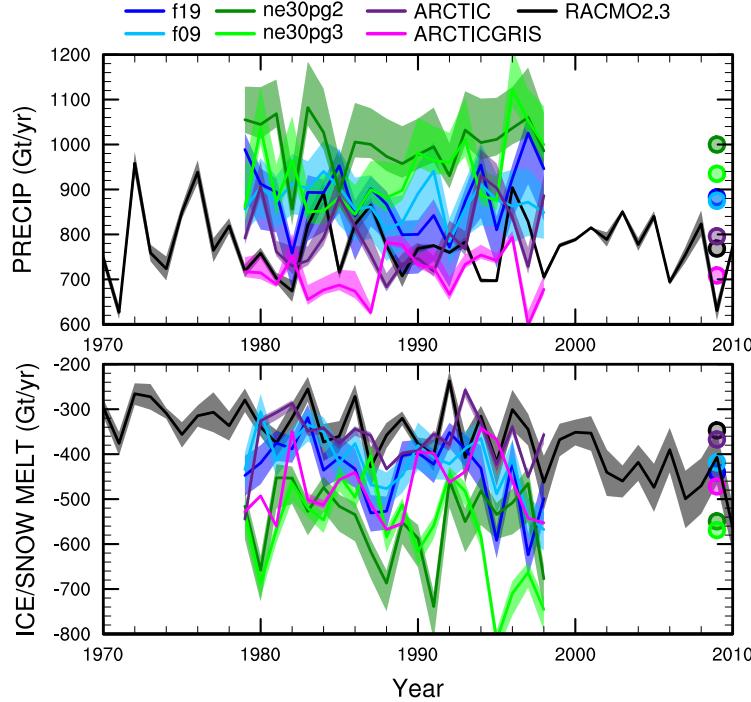
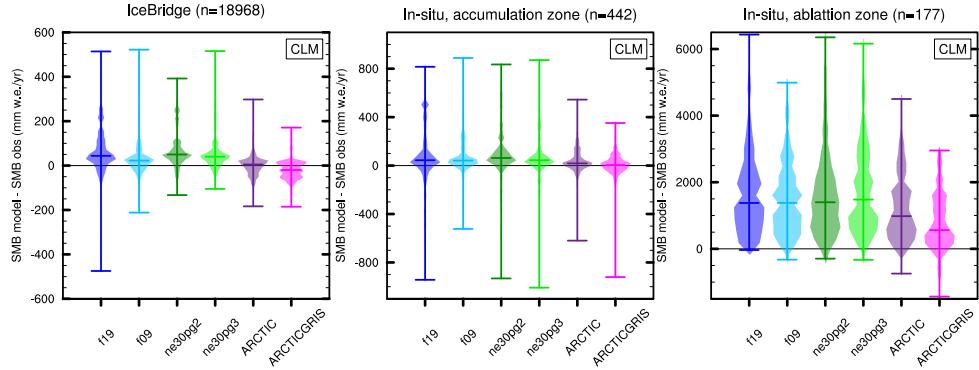
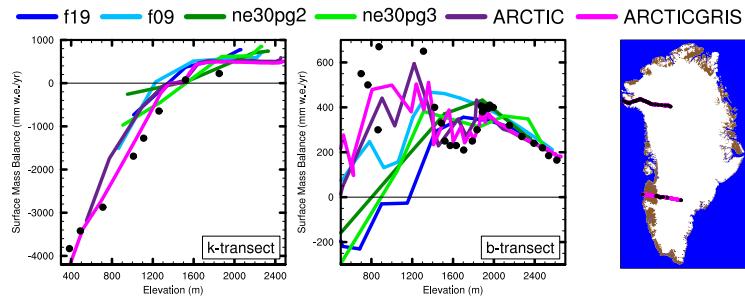


Figure 10. Time-series of annual (solid+liquid) precipitation (top) and annual runoff (bottom) integrated over the Greenland Ice Sheet for all six simulations and compared to RACMO3.2. The raw fields are mapped to two target low resolution grids, f19 & ne30pg2, and using two different remapping methods, conservative ESMF and high order TempestRemap. The remapped values are then integrated over the ice mask of the target grid. This gives four time-series for each simulation (three for f19 & ne30pg2), with the mean value given by the solid line and shading spanning the extent of the remapped solutions.

Figure 13 shows the representation of the surface of the ice sheet along these transects in the different grids, compared to the high resolution dataset used to generate CAM topography boundary conditions. The 1° – 2° grids are noticeably coarse, with only a handful of grid cells populating the transect. The f09 grid is a bit of an exception –since the grid cells become very narrow in the meridional direction at high latitudes, a larger number of grid cells can populate the east-west transect, consistent with its skillful representation of the ablation zone.

What the authors refer to as the “b-transect” in northwest Greenland (Figure 12) is characterized by orographic precipitation, resulting in the accumulation zone extending down to the ice margin. The variable resolution grids perform relatively well, with larger SMBs at lower elevations where the precipitation rates are highest, whereas the 1° – 2° grids underestimate the SMB in these lower elevation regions. Only the variable resolution grids can capture the local reduction in SMB in the 1500-2000 m region. The skill of the variable-resolution grids is clearly related to the more accurate representation of the steepness of the transect, while also capturing the protrusion around 1500-2000 m that coincides with the local minimum in SMB (Figure 13).

To give an idea of what the ablation zones look like in the highest resolution *ARCTCIGRIS* grid, Figure 14 is a still from a visualization produced by the authors. It shows the daily mean surface mass balance during the height of the melt season in year 1981 of the sim-

**Figure 11.** .**Figure 12.** .

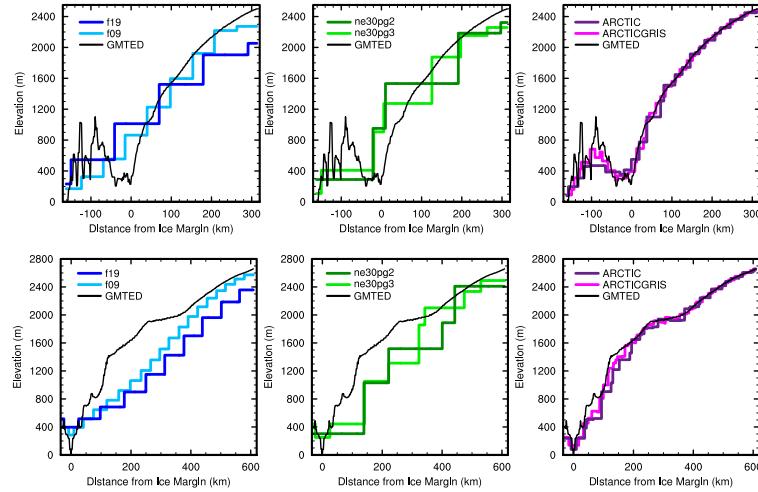
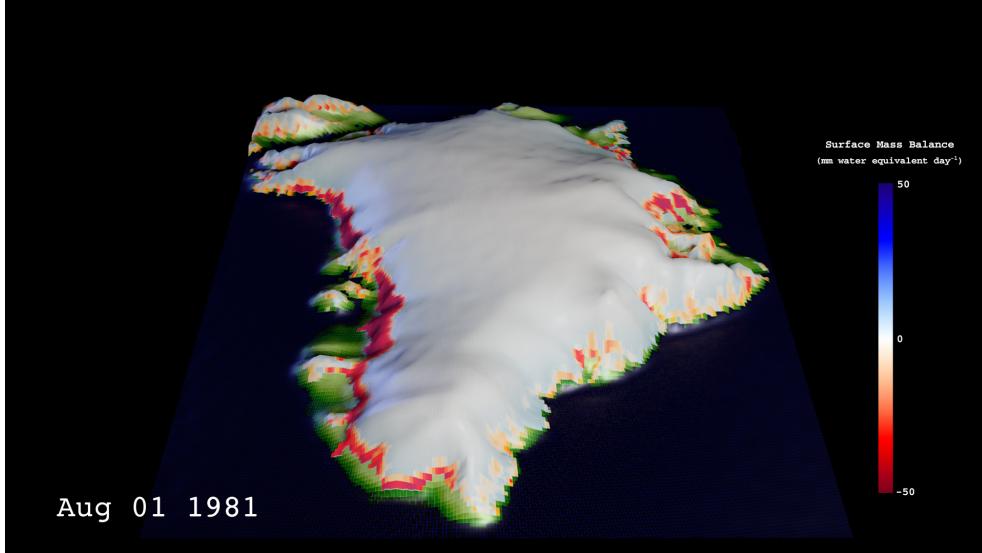
435 ulation. The ablation zone appears to be well resolved in the *ARCTICGRIS* grid. The
 436 full visualization is publicly available on youtube.com¹.

437 3.4 Impact of horizontal resolution on extremes

438 Synoptic storms are tracked and analyzed using TempestExtremes (Ullrich et al.,
 439 2021). As the *ARCTIC* grid contains $\frac{1}{4}^{\circ}$ refinement north of about 45° latitude, the storm
 440 tracker is applied to this region for the *ARCTIC* and *ne30pg3* run to identify differences
 441 in storm characteristics due to horizontal resolution. Figure 15 shows the mean precip-
 442 itation rates averaged over all January storms identified by TempestExtremes. The iconic
 443 comma structure of the synoptic cyclones is simulated in *ne30pg3* and *ARCTIC* grids,
 444 with the magnitudes about the same in these two grids, with perhaps a marginal increase
 445 in precipitation rates in the storm center of the *ARCTIC* grid. For good measure, the
 446 *ne30pg3** run is also plotted, and looks more-or-less identical to the *ne30pg3* run.

447 As has been previously reported, horizontal resolution can have large impacts on
 448 extreme precipitation events. Figure 16 is a PDF of the precipitation rates associated
 449 with synoptic storms, by month. The PDFs are constructed by sampling all the precip-
 450 itation rates within 30° of the storm center, for each point on the storm track and for
 451 all storms. The PDFs are evaluated on an identical composite grid for all runs, and so
 452 storm statistics are not impacted by differences in output resolution. The *ARCTIC* run
 453 has larger extreme precipitation rates compared to *ne30pg3* in every month, but the in-

¹ https://www.youtube.com/watch?v=YwHgqDu75s8&t=4s&ab_channel=NCARVisLab

**Figure 13.****Figure 14.**

crease is greatest in the summer months, which coincides with the most extreme events of the year. This is primarily due to an increase in resolution, and not the reduced physics times-step; the *ne30pg3** only marginally increases the precipitation rates compared with *ne30pg3*.

The extreme precipitation rates in the *ARCTIC* run are closer to the ERA5 reanalysis than in *ne30pg3* (Figure 16). It's difficult to know the extent that the extreme precipitation rates in ERA5 are constrained by data assimilation, or whether these precipitation rates are due to using a similar $\frac{1}{4}^\circ$ model as the *ARCTIC* grid. However, it is well documented that $\frac{1}{4}^\circ$ models are more skillful at simulating extreme events (Bacmeister et al., 2013; O'Brien et al., 2016), and so this is an additional benefit of the variable-resolution grids.

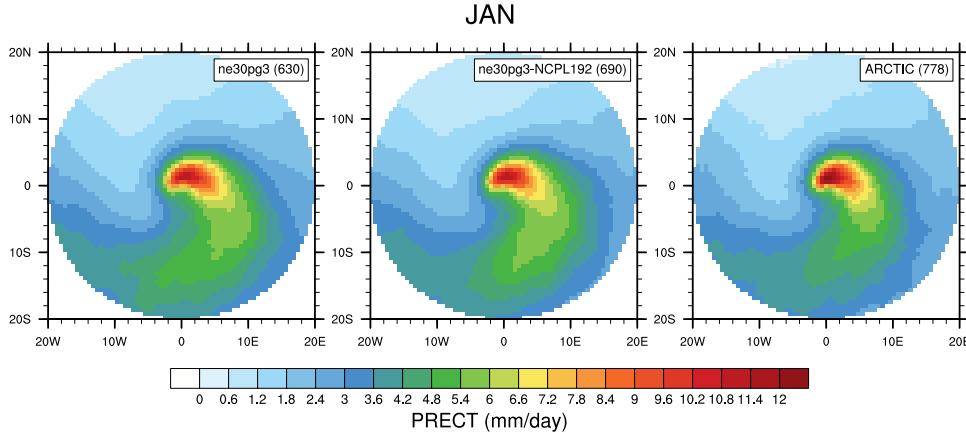


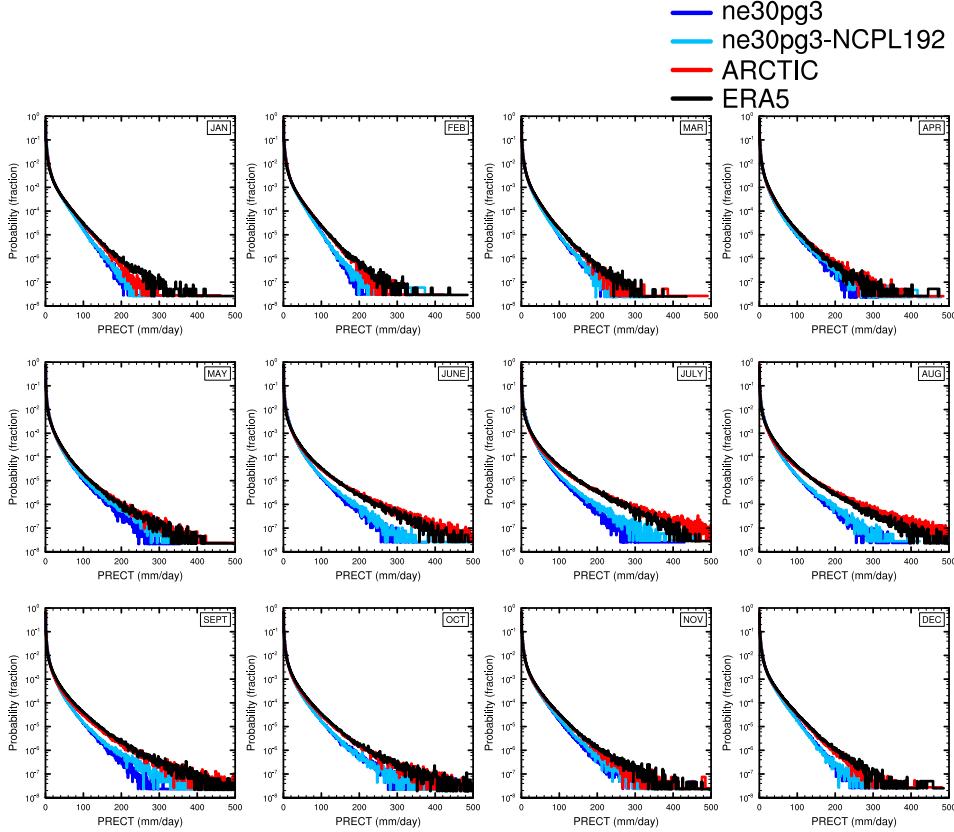
Figure 15. .

4 Conclusions

Six grids from different dynamical cores in CESM2.2 are evaluated in an AMIP-style configuration for their performance over the Arctic and in simulating the Greenland Ice Sheet (GrIS) surface mass balance (SMB). The $1-2^\circ$ finite-volume grids have enhanced resolution over Polar regions due to the convergence of meridian lines, although a polar filter is employed to prevent features from forming at this higher resolution. Spectral-element grids comparable to the resolution of the finite-volume grids have an isotropic grid structure, meaning the grid resolution is similar over the entire domain. Two variable-resolution grids were developed and introduced into CESM2.2 as part of this work. They use the spectral-element dycore; the *ARCTIC* grid has $\frac{1}{4}^\circ$ refinement over the broader Arctic, whereas the *ARCTICGRIS* grid is identical except it has an $\frac{1}{8}^\circ$ patch of refinement over Greenland.

In general, the finite-volume grids have colder summer temperatures over the Arctic, and the spectral-element grids (incl. the variable-resolution grids) are warmer. The cloud biases in all the uniform resolution grids, whether finite-volume or spectral-elements, are similar, in general being too cloudy over Arctic land masses. The variable-resolution grids largely improve the cloud biases. It should be emphasized that our analysis is specific to the Arctic summer due to its relevance to melt rates over GrIS; improved clouds in the Arctic should not be taken to mean that lower latitude regions have an improved cloud field as well.

At the regional level, a halo of negative cloud bias is found around the oceanic perimeter of Greenland in all $1-2^\circ$ grids, and is absent in the variable-resolution grids. This halo bias coincides with a positive cloud bias over the interior of the ice sheet. This has been traced back to the inadequacy of the lower resolution grids to resolve the orographic precipitation process in Greenland. Synoptic systems moving into Greenland are not sufficiently lifted when encountering the steep ice margins due to overly smooth topography in the $1 - 2^\circ$ grids. As a result, moisture penetrates across the ice margin and dumps excess precipitation into the interior of the GrIS, instead of being concentrated closer to the coastal margins as indicated by observations. This results in a positive precipitation and cloud bias in the ice sheet interior, and a halo of low cloud bias about the perimeter of Greenland. The agreement of different observational data products on this bias lends confidence in attributing the cause of the precipitation and cloud biases. The variable-resolution grids compare more favorably to the observations and indicate the orographic precipitation process in Greenland is largely resolved.

**Figure 16.** .

499 The primary source and sink terms of the SMB equation are integrated over the
500 GrIS, and evaluated in the six grids. The uniform $1 - 2^\circ$ grids all have large positive
501 accumulation biases owing to their inability to resolve the orographic precipitation pro-
502 cess. The uniform spectral-element grids have larger accumulation biases, suggesting that
503 the finite-volume grids are more skillful at resolving the precipitation processes due to
504 their finer grid spacing over Greenland, despite the polar filter. The variable resolution
505 grids have the most accurate accumulation rates of all the grids.

506 The primary mass sink term of the GrIS, ice/snow melt, expresses similar biases
507 as the accumulation rates. The uniform resolution spectral-element grids melt too much,
508 while the finite-volume grids have reduced biases. It's more difficult to attribute these
509 biases to grid resolution alone; the finite-volume grids have colder summers that is con-
510 sistent with their lower melt bias. However, the *ARCTICGRIS* grid has the warmest
511 summer temperatures of all grids, yet it has reduced melting compared to the uniform
512 resolution spectral-element grids. This suggests that grid resolution is responsible for a
513 large fraction of the melt biases. That the melt process would be sensitive to grid res-
514 olution is intuitive, since local elevation impacts on temperature and melting is largely
515 handled by the land model, which does not have a pole problem because there is no hor-
516 izontal dynamics. The larger number of grid cells in the ablation zone in the finite-volume
517 grids seems to more accurately resolve the melt process relative to the uniform resolu-
518 tion spectral-element grids, in which the ablation zone is made up of substantially fewer
519 grid cells.

520 **Acknowledgments**

521 This material is based upon work supported by the National Center for Atmospheric Re-
 522 search (NCAR), which is a major facility sponsored by the NSF under Cooperative Agree-
 523 ment 1852977. Computing and data storage resources, including the Cheyenne super-
 524 computer (doi:10.5065/D6RX99HX), were provided by the Computational and Infor-
 525 mation Systems Laboratory (CISL) at NCAR.

526 The data presented in this manuscript is available at <https://github.com/adamrher/2020-arcticgrids>.

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