Mechanisms causing reduced Arctic sea ice loss in a coupled climate model

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Abstract

The fully-coupled climate model HadGEM1 produces one of the most accurate simulations of the historical record of Arctic sea ice seen in the IPCC AR4 multi-model ensemble. In this study, we examine projections of sea ice decline out to 2030, produced by two ensembles of HadGEM1 with natural and anthropogenic forcings included. These ensembles project a significant slowing of the rate of ice loss to occur after 2010, with some integrations even simulating a small increase in ice area. We use an energy budget of the Arctic to examine the causes of this slowdown. A negative feedback effect by which rapid reductions in ice thickness north of Greenland reduce ice export is found to play a major role. A slight reduction in ocean-to-ice heat flux in the relevant period, caused by changes in the MOC and subpolar gyre in some integrations, is also found to play a part. Finally, we assess the likelihood of a slowdown occurring in the real world due to these causes.

1 Introduction

The extent of Arctic sea ice, as measured by satellite microwave sensors, has been decreasing in all seasons over the past thirty years. The most severe decline has been observed in the month of September, when the ice shrinks to its annual minimum extent, with a series of record lows being observed in 1995, 2002, 2005 and 2007. The decline has also shown signs of accelerating in recent years, with the past five years having seen the lowest five September mean ice extents on record (Stroeve et al., 2012).

The September minimum ice extent is of particular interest to researchers because of its close relationship to ice volume (which is not currently directly measurable), and because its value defines the difference between a perennial ice cover and a seasonal ice cover. Should the Arctic become nearly, or completely, ice-free in September there would be serious implications for wildlife both on sea and on land, and for native Arctic
peoples. A seasonal ice cover would also open the Arctic to shipping for one or more months of the year, and exacerbate current international tensions over Arctic waters.

The speed of melting of ice during the summer, and hence June and July ice extent, are also closely related to the September minimum extent. Lower ice extent in these months results in more solar energy being absorbed by the mixed ocean layer. This will tend to be released to the atmosphere during autumn and early winter, with consequences for the regional climate. A number of recent studies (Strey et al., 2010; Francis et al., 2009) have suggested that there will also be consequences for weather in mid-latitudes during these seasons.

The projections of mean September sea ice extent submitted to the IPCC AR4 tended to underestimate the rate of decline even before the 2007 record low September extent (Stroeve et al., 2007). However, a small subset of these models, especially those including sophisticated techniques such as a sub-gridscale ice thickness distribution, match the observed decline more closely (Wang and Overland, 2009). The fully-coupled climate model HadGEM1 (Johns et al., 2006) was one of the six models named in the Wang and Overland study. In this paper, we examine the future projections of Arctic sea ice in HadGEM1, and compare its projections in the satellite era to the observational data, especially in the five years that have elapsed since the model data was first available. In particular, we find that a slowing, even stopping, of Arctic sea ice loss is projected between about 2010–2030. We examine possible reasons for this, and attempt to assess the likelihood of this being observed in the real world.

2 Projections of sea ice in HadGEM1

HadGEM1 (Hadley Centre General Environmental Model 1) is a fully coupled atmosphere-ocean general circulation model. The sea ice component is divided between the atmosphere and the ocean sections, but is mostly located within the ocean. It includes elastic-viscous-plastic dynamics (Hunke and Dukowicz, 2002), zero-layer thermodynamics (Semtner, 1976), an ice ridging scheme (Thorndike et al., 1975;
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In this paper we concentrate on two ensembles of the HadGEM1 model; firstly the “ALL” ensemble, which consists of four experiments with historical anthropogenic, solar and volcanic forcing from 1859–2000, each starting from a different point in a long control run. All were continued to 2010, and the last three continued to 2030, using the SRES A1B scenario, an eleven-year assumed solar cycle, and exponentially decaying volcanic forcing. Secondly, we examine the “ANT” ensemble, which consists of four experiments with historical anthropogenic forcing only from 1859–2000, also starting from four different points in a control run. These experiments were continued to 2030, with one ensemble member continued to 2100, using the SRES A1B scenario.

Figure 1a shows the projected mean September Arctic sea ice extent according to all eight experiments in these ensembles, from 1960–2070. Also plotted is the observed sea ice extent from the HadISST dataset, based on passive microwave satellite observations (Rayner et al., 2003), and a 5-yr running mean of the “uncertainty interval” used by Wang and Overland (2009), the values between 20% above and 20% below the HadISST timeseries. To assess the performance of the ensembles in the period of satellite observations, we compare the mean value, the linear trend, and the detrended standard deviation over the period 1979–2010, shown in Table 1. In mean value and annual linear trend, the ALL ensemble values match observations extremely well, and the ANT ensemble quite well. The detrended standard deviations are also similar, although the model has a slightly lower variability than observations.

If we examine the spatial pattern of ice concentration decline, we see differences between HadGEM1 and observations. In the HadISST observations (Fig. 2a), ice loss
is strongly concentrated in the Pacific sector; in the ALL ensemble (Fig. 2b) ice loss is weakly concentrated in the Atlantic sector, while in the ANT ensemble (Fig. 2c) it is fairly evenly spread around the edge of the Arctic. This discrepancy may be due to the simulation of ice thickness in HadGEM1. Although one of the best spatial patterns simulated by an AR4 model (Kwok, 2011), it produces overly thick ice in the Pacific sector, and overly thin ice in the Atlantic sector. This is illustrated by Fig. 2d, showing a superposition of satellite and submarine observations of ice thickness in the 1990s (described more fully in Appendix A), and Fig. 2e, f, showing the modelled ice thickness pattern for this period the ALL and ANT ensembles respectively. This effect would be likely to decrease concentration in the Atlantic (Pacific) sector sooner (later) than has been observed.

Having briefly evaluated the performance of the HadGEM1 ensembles up to the present, we now examine its future projections. In Fig. 1a, the most surprising aspect of the timeseries post-2010 is a slowdown, even a temporary cessation, of sea ice loss projected in both ensembles. After 2030 ice loss appears to resume in the single continuing ensemble member. In order to examine the rates of decline more closely, a 15-yr running gradient was plotted for each experiment, using the gradient of the least-squares linear fit to each 15-yr interval (Fig. 1b). The “flattening” is more clearly visible in the ALL ensemble than the ANT; there is a clear minimum in the gradients in the late 1990s, followed by an increase to near, or above, zero at around 2010.

Formally, a shallowing of the decline was declared to be significant for any individual experiment if there existed two successive 15-yr periods for which the gradient intervals did not overlap. The gradient interval for any 15-yr period was defined as $[\text{gradient} - 2 \times (\text{stderr}) ; \text{gradient} + 2 \times (\text{stderr})]$, where stderr is the standard error in the linear fit to the 15-yr timeseries. Using this method, we find that the shallowing was significant in two out of the three ALL members continued to 2030 (ALL 3 and 4), and two out of the four ANT members (ANT 2 and 3). The same analysis was then carried out on the September ice volume, a variable with less interannual variability. The shallowing in this variable was significant for all of the above experiments (and in addition
for ANT 1), demonstrating that there is a clear slowing of ice loss in many of the model runs, at similar times.

The timeseries of mean global temperature, and mean temperature north of 70°N (“Arctic temperature”), were examined, from 1980–2030, for the ALL and ANT experiments (not shown). Sea ice extent was also plotted against these two indices (Fig. 3). Arctic temperature varies roughly linearly with sea ice extent in all experiments; for example, in the experiment ALL 4 the timeseries rises to a maximum of −12.8°C in the late 2000s, then decreases to −14.8°C in the mid-2010s as ice extent increases, before slowly rising again. The behaviour of global temperature is more complicated, as evidenced by the greater scatter seen in Fig. 3a than in Fig. 3b. Only one of the ANT experiments, ANT 3, displays a slowing of global temperature rise; by contrast, all three of the continuing ALL experiments show this. In ALL 2 and 3 the gradient minimum occurs in 2005 and 2011 respectively, but ALL 4 does not show a defined minimum, the temperature remaining roughly constant from 1999–2020. The magnitude of the slowing in global temperature rise in these experiments is such that the Arctic temperature changes alone are not sufficient to explain them, which suggests the existence of other factors, external to the Arctic, which are helping to cause both.

Prior to 2010, the experiments have projected sea ice loss with reasonable accuracy, so here we analyse why the slowdown occurs, to identify the model processes responsible, and to assess the likelihood of the slowdown being observed in the real world.

3 Arctic heat budget

3.1 Methods and error evaluation

To examine the Arctic energy budget, we divide the region into the three components of atmosphere, ice and ocean. Energy fluxes passing between these components, and across the Arctic boundary, were calculated (Fig. 4). The Arctic was defined for this
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3.2 Results

Decadal means of calculated fluxes for 1960–2030 are plotted in Fig. 5, showing the ensemble mean for the ALL and ANT experiments respectively. For each ensemble, the time mean control value of each flux from 1960–2030 has been subtracted, to enable the decade-to-decade changes to be assessed. In the discussion paragraphs below, the label “1980s” refers to decadal means from 1980–1989, and similarly for all other decades.

For the ALL ice heat budget, we see a decrease in the ice heat uptake (IHU) of 2.30 TW from the 2000s to the 2010s, to near-zero levels, corresponding to the purpose as the Arctic Ocean, shown in Fig. 4. Vertical fluxes were calculated directly from flux diagnostics. Ice heat uptake and transport were calculated using rate-of-change diagnostics, and ocean heat uptake calculated as the rate of change of ocean heat content. Ocean heat content and ocean heat transport were calculated as integrals, and atmospheric heat transport as a residual. In addition, the “residual” ocean error term was calculated (the ice residual was three orders of magnitude lower than the other terms, and the atmosphere residual zero by design). It was assumed that atmospheric heat uptake was of negligible size. The methods used are described in more detail in Appendix B.

The heat budget allows easy comparison between the ice volume changes and the energy balance in the ice because the ice component of the HadGEM1 model is “zero-layer”, effectively assuming the ice to have no heat capacity (although a very thin “skin layer” at the top is allowed to vary in temperature to better simulate the diurnal cycle). Thus any excess heat flux into the ice will simply melt the ice, while an energy deficit will cause a proportional amount of ice to be created. Ice volume and ice heat energy become effectively the same quantity, related by the constant $-\rho_{\text{ice}} q_{\text{ice}}$, $\rho_{\text{ice}}$ being ice density and $q_{\text{ice}}$ being specific latent heat of melting. Ice heat uptake is equivalent to ice volume loss, and “ice heat transport” into the Arctic is equivalent to advection of ice out of the Arctic Ocean region.
slowdown in ice loss. The ocean-to-ice heat flux (OI) contributes by some margin the largest amount to this change, decreasing by 1.85 TW. There is a corresponding increase in the ice heat uptake, of 2.67 TW, from the 1980s to the 1990s, as the ice melt speeds up. This appears to be overwhelmingly driven by an increase in OI of 3.33 TW, opposed by a smaller decrease (1.09 TW) in atmosphere-to-ice heat flux (AI). From the 1990s to the 2000s there is little change in IHU, as increases in OI (2.08 TW) and AI (3.29 TW) are more than balanced by a very large decrease in ice heat transport (IHT) of 6.91 TW. This may be partly due to the ongoing ice loss itself, as the ice flowing out of the Arctic Ocean will be expected to be thinner. It appears likely from this discussion that the OI is the predominant driver, in the ALL runs, of the different rates of melting between 1980–2030, although the large drop in IHT after the 1990s also has some effect. We therefore next examine the ocean heat budget (Fig. 5c) to try and determine the causes of the changes in OI.

Ocean heat transport (OHT) rises steeply from the 1980s to the 2000s (increasing by 11.87 TW and then 10.61 TW from decade to decade), then decreases slightly by 1.93 TW into the 2010s. From the 1980s–1990s the large increase is mainly balanced by oceanic heat uptake (OHU), which rises by 6.58 TW; from the 1990s–2000s it is balanced by atmosphere-to-ocean flux (AO), which falls by 9.73 TW. The AO and ice-to-ocean (IO, equal and opposite to OI) fluxes are in fact consistently decreasing between the 1980s and 2000s. We conclude that the changes in OHT are entirely driving the changes in OHU, and are opposed by the changes in AO and IO; increased heat transport into the Arctic is resulting in ocean warming and increased heat flow to the atmosphere and ice. From the 1980s to the 1990s the increase principally causes ocean warming; from the 1990s to the 2000s there is no increase in ocean warming, but instead greatly increased heat flow to the atmosphere, while ocean-to-ice flux increases throughout. From the 2000s to the 2010s, by contrast, the OHT does not increase significantly, even reducing slightly, and hence the ocean-to-ice heat flux decreases as well.
In summary, the slowdown in ice melt in the ALL runs appears to be principally attributable to the slowdown in the increase of oceanic heat transport into the Arctic from the 2000s to the 2010s. However, the sharp decrease in ice transport from the 1990s to the 2000s may also have had some effect, and would bear further investigation.

The ice heat uptake in the ANT experiments displays slightly different behaviour to that of the ALL experiments (Fig. 5e). Ice heat uptake begins from a fairly high value in the 1980s (1.18 TW). It displays a small decrease of 1.30 TW into the 1990s, a larger increase of 1.82 TW into the 2000s, and finally a large decrease of 3.10 TW into the 2010s corresponding to the slowing of ice loss. This last decrease appears to be entirely caused by a decrease in IHT (5.37 TW), with the flux terms from the atmosphere and the ocean actually opposing the change, decreasing by 1.47 and 0.80 TW respectively. Advective effects are acting to slow ice loss, while thermodynamic effects are acting to increase it; in this short time period the advective effects are “winning”. The decrease from the 1980s to the 1990s appears to be caused by another decrease in IHT (2.52 TW), opposed by a rise in OI (1.91 TW); the increase from the 1990s to the 2000s caused by a large increase in Al (2.60 TW), opposed by another small decrease in IHT (0.71 TW). In this experiment, the OI does not appear to be nearly as important as the IHT.

To summarise, the slowdown in ice loss shows clearly in the heat budget of both ensembles as a drop in IHU from the 2000s–2010s. It is accompanied by steep drops in IHT in both ensembles, although the ALL drop occurs a decade earlier than that of IHU. There is also a decrease in OI in the ALL ensemble, probably caused by the slowdown in the rise of OHT. Together, the heat budgets suggest the following questions: What causes the drop in ocean heat transport in the ALL ensemble, and why is a similar drop not observed in the ANT ensemble? Why do the steep drops in ice heat transport occur, and why do they occur when they do?
3.3 Changes in ice export

A decreasing trend in ice export from the Arctic is a feature of all of the simulations. The ice transport in the ALL runs shows a short but fairly continuous period of decline (Fig. 6a), from about 30 TW in 1990 to about 23 TW in 2010. In the ANT runs, by contrast, there is a sharper decrease around 2010, from 27 TW to 20 TW. This decline is almost certainly playing a part in the slowing of ice loss. It could itself be caused by a decrease in the thickness of ice leaving the Arctic, and/or a decrease in the speed of outflow.

Ice export can also be calculated as \( \int h_{\text{ice}} v_{\text{ice}} \, ds \), where \( h_{\text{ice}} \) and \( v_{\text{ice}} \) denote ice thickness and velocity respectively, and \( S \) is the Arctic Ocean boundary. To examine the relative effects of thickness and velocity on the transport timeseries, the ice heat transport across the Fram Strait was calculated in this way, holding the ice thickness (velocity) fields constant, using the time mean field, in order that only the ice velocity (thickness) was varying in time.

There was no qualitative change to the patterns observed when ice export was calculated as \( \int h_{\text{ice}} v_{\text{ice}} \, ds \); the same patterns of decline for both ensembles were observed (Fig. 6b). The same was true when the velocity fields were replaced by the time mean (Fig. 6c). However, when the thickness fields were replaced by the time mean the patterns changed completely, and ice transport actually increased in every run over the period 1960–2030 (not shown). From this it can be concluded that the changes in ice transport are entirely due to decreases in the thickness of the ice arriving at the boundary, and increased ice velocity is acting to oppose this slightly. This does not immediately rule out a change in atmospheric circulation as a cause; more ice could for example be being advected from the eastern Arctic after 2010, a region of generally divergent, and therefore thinner, ice.

Two sub-regions of the Arctic adjacent to the Fram Strait were identified; the region between 120°W–0°W, north of 79°N, known hereafter as the “Western Arctic” (WA), and the region between 0°E–120°E, north of 79°N, known hereafter as the “Eastern
Arctic” (EA). In order to attribute the decline in Fram Strait ice transport to (a) changes in ice thickness in the WA; (b) changes in ice thickness in the EA; (c) changes in the proportion of ice coming from each region, mean ice thickness in the whole Arctic Ocean (AO), WA and EA was plotted for each ensemble, and for each ensemble mean, from 1960–2030, and 15 yr gradients were calculated.

The gradients of ice thickness change in the WA show pronounced minima in the late 1990s and late 2000s for the ALL and ANT ensembles respectively, indicating step decreases in WA ice thickness at these times (Fig. 6e). The gradients of change of AO and EA ice thickness (not shown) also show minima close to these times, but these are higher and closer in value to the rest of the gradient timeseries. The step decreases in WA ice thickness occur, in both ensembles, immediately before the steepest decrease in ice export; it is therefore likely that this decrease is the most important cause of the ice export decrease.

As a second check, the ice export was calculated again as \( \int h_{\text{ice}} v_{\text{ice}} \, ds \); the velocity at the boundary was again replaced by the time mean field, but the thickness at the boundary was also replaced by the time mean field plus the timeseries of anomalies in WA ice thickness. Very similar patterns in ice export were seen (Fig. 6d), but exaggerated and shifted earlier in time; a step change from \( \sim 23 \) to \( \sim 15 \) TW is seen in the late 2000s in the ANT runs, with a slightly steadier decrease being observed in the ALL runs during the 1990s, from \( \sim 28 \) to \( \sim 15 \) TW. This suggests that the changes in WA ice thickness are accounting for the changes in ice export, but that there is a time delay and that the effects are ameliorated somewhat by ice transport from other areas of the Arctic.

To identify reasons for the sudden decreases in WA ice thickness, the heat budget analysis was performed again for the WA region. The step decreases were visible in most integrations as increased ice heat uptake. There is large variation between the runs in the timing of changes in the other ice terms, but there are a number of common themes, including “flushing” events (years with exceptionally high ice export) shortly before the decreases, and periods of high ocean-to-ice heat flux. In three of the runs
(ALL 3, ANT 1 and ANT 2) these periods coincide with periods of particularly high ocean heat convergence in the WA region, while in ALL 2 high OI coincides with a maximum in the mixed layer depth. In ALL 4 and ANT 3, the cause of the high OI is difficult to determine.

3.4 Changes in ocean heat transport

In HadGEM1, the principal source of heating of the Arctic Ocean by water transport is from the Atlantic Ocean, through Fram Strait and the Barents Sea. The rate of heating from the Atlantic is controlled by the temperature of the water at inflow, and the speed of inflow. To examine the relative effects of these terms, a similar analysis to that in section 3.3 was performed: the oceanic heat transport $OHT = \int T_{ocn} u_{ocn} \cdot dS$ was recalculated twice, $T_{ocn}$ and $u_{ocn}$ being held constant in turn by replacing each term by its respective time mean field. $T_{ocn}$ and $u_{ocn}$ are ocean temperature and current velocity respectively, and $S$ is the Arctic Ocean boundary.

Figure 7 shows OHT as calculated in each of these three ways. Figure 7a shows OHT as in the heat budget, with time-varying temperature and velocity; in the ALL runs we see a steep rise in OHT from the late 1980s, checked and briefly reversed in the mid-2000s, and resuming from about 2020. There is variation within the ensemble; OHT is reversed most strongly in ALL 4, but checked hardly at all in ALL 3. The ANT ensemble displays a very similar pattern, shifted 7–8 yr into the future. These patterns are to a large degree replicated when the velocity field is replaced by the time mean (Fig. 7b); however, when the temperature field is replaced by the time mean (Fig. 7c), the timeseries becomes virtually flat, although there are still weak falls in OHT around the time of the reversals. This suggests that the water flowing into the Arctic is warming, then cooling, then warming again, and that around the time of the reversals there is also a slight reduction in the into-Arctic current velocities.

To seek an explanation for these effects in turn, we examine two important aspects of the North Atlantic circulation; the Meridional Overturning Circulation (MOC) and the Subpolar Gyre (SPG). We define the strength of the MOC in the usual way, as the
maximum overturning streamfunction in the North Atlantic. The SPG is more complex to define as there are two separate cyclonic gyres in the North Atlantic, in the Labrador Sea and in the Nordic seas. We concentrate on the Nordic gyre as it influences conditions at the Arctic-Atlantic boundary more directly, and define its strength to be the maximum gyre streamfunction between 66–80° N.

The two indices are plotted, for all integrations, in Fig. 8. The MOC decreases in strength in all integrations; the ALL ensemble exhibits a step decrease at around 2008, but in the ANT ensemble the decrease is more gradual. The distribution of the ALL MOC indices in the period (1993–2007) was compared to that in the period (2008–2023). The initial period has mean index 20.1 and spread 0.8; the later period mean 18.5 and spread 1.2. The distributions are significantly different at the 5% confidence level, i.e. there is a significant weakening of the MOC in 2008 of around 1.6 Sv. A decrease in the MOC index would be expected to have a significant negative effect on the ocean temperature at the Arctic-Atlantic boundary; therefore it is likely that this step decrease is a major cause of the changes.

In the SPG index there is no systematic decrease. However, in ALL 4 a short-term weakening is visible, with a minimum in 2013 2–3 Sv below the long-term mean. Interestingly, there is a similar short-term weakening in ALL 2 in the late 1980s which also coincides with a period of relatively high ice extent (Fig. 1) and low OHT (Fig. 7). There is in general no correlation between SPG index and September ice extent, but in the two cases mentioned above low SPG index coincides with low temperatures in the top 70 m of the Nordic Seas, low OHT, low Arctic OI heat flux and low September ice volume as well. It is therefore reasonable to assume that while the SPG is not in general a major factor influencing sea ice, in these cases long periods of an extremely low index are causing greater sea ice cover than would otherwise be the case. While changes in the SPG index cannot therefore explain the slowing of ice loss across the ensemble, it may explain why the slowing is most pronounced in ALL 4.

The MOC changes in particular may help to partly explain the behaviour in global temperature timeseries seen in Sect. 2. The sharp change in the MOC in the ALL
runs in late 2008, with the concurrent stable Arctic sea ice and Arctic temperatures, may have helped to keep global temperatures at a roughly constant level for several years (although as for ALL 3 and 4 the “flattening off” of global temperatures occurred around 2000, this was probably not the only factor). The more gradual decline in the MOC in the ANT runs is consistent with the more constant rate of temperature rise in this ensemble.

4 Discussion

HadGEM1 has projected ice loss with reasonable accuracy up to the present day, but projects a slowdown in ice loss to occur in the near future (from 2010–2030), in the ALL ensemble a little earlier than in the ANT ensemble. The timing and severity of the slowdown varies between individual experiments. A brief halt in the rise of global average temperature is simulated at a similar time to the slowing of ice loss.

In this paper, three principal mechanisms have been identified as being responsible for this slowdown, two closely related. Firstly, there is the weakening of the MOC, which appears to reduce (or slow the rise of) the heat flowing into the Arctic Ocean, and hence reduce the ocean to ice heat flux. The weakening is sudden in the ALL ensemble, occurring at around 2008; it is slower, and about a decade later, in the ANT ensemble, and thus goes some way towards explaining the difference in timing of the slowdown between the two ensembles. It may also be partially responsible for the halted rise in global temperatures.

Secondly, there is the negative ice export feedback identified, whereby a thinning ice cover will reduce the rate at which ice is exported from the Arctic. Ice loss in the ALL experiments reaches its maximum rate in the late 1990s (Fig. 1b). Most of the ANT experiments do not display clear minimum gradients of ice loss, but ANT 3 displays a similar gradient minimum in the early 2000s. All of these experiments then display similar rates of rise in gradient as ice loss slows. The negative ice export feedback would be expected to begin to take effect 3–7 yr after the gradient minima, based on
the average residence times for ice in the Arctic. Therefore, the timing of fastest loss may also provide an explanation as to the differing times of the slowdown. Whether the value of the steepest rate of change affects the strength of this feedback is not clear.

Lastly, there are the sudden decreases in ice thickness in the region immediately north of Greenland, in the late 1990s and late 2000s in the ALL and ANT ensembles respectively, which together with the effect above appear to precipitate the sudden declines in ice export from the Arctic in around 2000 and 2010 respectively. Clearly these are related to some extent to the time of fastest total ice loss, but other than this their cause is difficult to determine. In some experiments “flushing” events appear to play a role. In others, temporary periods of increased oceanic heat convergence to this region appear to be responsible.

Do we expect any of these effects, and a consequent slowing of ice loss, to be observed in the near future? A slowing of the MOC during the 21st century is widely projected to occur by climate models, as a result of warming and freshening of the North Atlantic. However, this is likely to lead to a smaller slowing of ice loss than that projected in HadGEM1, for the reasons detailed in Sect. 2: while in HadGEM1 regions of present-day ice loss are spread fairly evenly around the edge of the Arctic, in observations the region of ice loss is strongly concentrated in the Pacific sector. We would therefore expect ice loss to be less strongly affected by changes in the Atlantic. It has also been found that a decreasing MOC will not necessarily decrease OHT into the Arctic Ocean (Bitz et al., 2006) as in HadGEM1 the centres of overturning tend to move north in the long term, closer to the Arctic-Atlantic boundary.

The negative feedback of ice export is almost certainly a real effect which has been already acting to retard ice loss slightly (relative to the loss that would have been observed with a constant rate of ice export). Nevertheless, it is unlikely that the feedback could on its own slow ice loss as time progresses. This can be shown using a simple box model of the Arctic, evolving ice volume and ice export rate in time, described in Appendix C. Ice loss could probably only be briefly slowed by short periods of fast ice loss in certain parts of the Arctic, as described above.
The causes of the sudden ice losses north of Greenland are too varied and complex for the likelihood of their occurring in the real world to be easily assessed, although increased ocean heat convergence has been found to be important in causing Arctic-wide rapid ice loss events in other models beside HadGEM1 (Holland and Bitz, 2006). The “flushing” events seen in some simulations suggest that, following these, total ice loss may for some time continue at a slower rate than that which was occurring before the event. A close precedent for this may have actually occurred in the early 1990s. It is generally believed that a sudden increase in the Arctic Oscillation index around 1989–1990 resulted in a large amount of thick multiyear ice being expelled from the Arctic (Rigor and Wallace, 2004; Rothrock et al., 2003), and about this time there appears a local minimum in ice extent in all months. However, in the following years ice extent increased again, recovering to 1980s levels by 1995, whereupon another sharp decrease occurred (Fig. 9). Reduced ice export from the Western Arctic region may have helped in this temporary recovery. It would be theoretically possible for this mechanism to temporarily slow ice loss at any point in the future.

Given the causes discovered, we cannot conclude, from the HadGEM1 projections, that a slowdown in ice loss is to be expected soon, particularly as the slowdown has already started in the present year of the experiments. However, the model has provided clues as to what mechanisms might be causing a slowdown, were one to be observed at any point in the future.

Appendix A

Combining satellite and submarine observations of ice thickness in one figure

A1 Description of datasets

The spatial pattern of mean ice thickness from 1994–2000 shown in Fig. 2d is derived from two datasets: (i) a multiple regression of moored upward-looking sonar submarine

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measurements of ice draft (Rothrock et al., 2008); (ii) gridded measurements of ice thickness from satellite radar altimetry (Laxon et al., 2003). The submarine observations cover a large area of the central Arctic known as the SCICEX box, including the North Pole; the satellite observations cover the area south of 81.5° N. Together they cover much of the Arctic Ocean. The figure was produced for illustrative purposes only and is not itself a new dataset; due to the very different nature of the two sources, it should be used with great caution.

The two datasets, in their raw form, represent different quantities. The submarine data, obtained via the formula given in Rothrock et al. (2008), is effectively gridbox mean ice draft (including open water areas). To convert to gridbox mean ice thickness the field was multiplied by \( \frac{\rho_{\text{water}}}{\rho_{\text{ice}}} \), the ratio of water and ice densities. The satellite data, however, is in the form of mean ice thickness not including thin ice and open water. “Thin ice” is defined by Laxon et al. (2003) as ice thinner than 0.5 to 1 m. It was necessary to carry out a transformation on this field, based on the HadISST ice concentration data and described below, to convert it to approximate gridbox mean ice thickness.

### A2 The “thick ice” to “gridbox mean ice” transformation

Let \( H_1 \) denote mean thickness of “thick ice”, as in the raw dataset, \( \gamma \) ice concentration, \( H_{\text{MOI}} \) mean ice thickness over ice (i.e. not including open water), and \( H_{\text{GBM}} \) gridbox mean ice thickness. We say that \( H_1 \) is mean thickness of ice greater than \( \eta \) metres thick, knowing that \( 0.5 \leq \eta \leq 1 \). Let \( g(h) \) be the ice thickness distribution function, and assume that for \( 0 < h \leq \eta \), ice growth rate is inversely proportional to ice thickness \( \left( \frac{dh}{dt} = \frac{C}{h} \right) \), and that \( g(h) \propto \frac{1}{dh/dt} \) (i.e. distribution density at \( h \) is inversely proportional to rate of growth at \( h \)). Then \( g(h) = g_1 h \), some constant \( g_1 \), for \( 0 < h \leq \eta \).

\[ (A1) \]
Assume that for high concentration $\gamma$, $g_1$ is inversely proportion to $H_{MOI}$, i.e.:

$$H_{MOI} = \frac{1}{\lambda g_1} \tag{A2}$$

Then

$$H_{GBM} = \int_0^\infty hg(h).dh = \int_0^n g_1 h^2 .dh + \int_n^\infty hg(h).dh = \left[ \frac{1}{3} g_1 h^3 \right]_0^n + H_1 \int_n^\infty g(h).h \tag{A3}$$

$$\Rightarrow \gamma H_{MOI} = \frac{1}{3} g_1 \eta^3 + H_1 \left( \gamma - \int_0^n g_1 h .dh \right)$$

$$\Rightarrow \frac{\gamma}{\lambda g_1} = \frac{1}{3} g_1 \eta^3 + H_1 \left( \gamma - \left[ \frac{1}{2} g_1 h^2 \right]_0^n \right) = g_1 \eta^2 \left( \frac{\eta}{3} - \frac{H_1}{2} \right) + H_1 \gamma$$

$$\Rightarrow g_1^2 \eta^2 \lambda \left( \frac{\eta}{3} - \frac{H_1}{2} \right) + H_1 \gamma \lambda g_1 - \gamma = 0 \tag{A4}$$

This quadratic in $g_1$ solves to give

$$g_1 = \frac{-\lambda H_1 \gamma \pm \sqrt{(\lambda H_1 \gamma)^2 + 4 \gamma \lambda \eta^2 \left( \frac{\eta}{3} - \frac{H_1}{2} \right)}}{2 \lambda \eta^2 \left( \frac{\eta}{3} - \frac{H_1}{2} \right)} \tag{A5}$$

We take the “+” solution as it allows $g_1 \to 0$ as $H_1 \to \infty$; intuitively we would expect the proportion of thin ice to be small for large $H_1$. For the quantity inside the square root to be positive, $\lambda \geq \frac{4 \eta^2}{\gamma} \left( \frac{1}{2 H_1} - \frac{\eta}{3 H_1^2} \right)$ is sufficient; and for $H_1 \geq 1$, $\eta \in [0.5, 1]$, $\lambda \geq 3/4 \gamma$ is sufficient. If we restrict $\gamma \geq 1/2$ then we can set $\lambda = 2$ for simplicity.
So for $0.5 \leq \gamma \leq 1$, we say
\[
g_1 = \frac{-2H_1\gamma + \sqrt{4(H_1\gamma)^2 + 8\gamma \eta^2 \left(\frac{n}{3} - \frac{H_1}{2}\right)}}{4\eta^2 \left(\frac{n}{3} - \frac{H_1}{2}\right)}
\]  
(A6)

For $0 \leq \gamma \leq 0.5$, we try
\[
g_1 = \gamma - \alpha(H_1)\gamma^2,
\]  
(A7)

so that for small $\gamma$ almost all ice is thinner than $\eta$. The functions must match up at $\gamma = 0.5$; therefore
\[
\alpha(H_1) = 2 + \frac{H_1 - \sqrt{H_1^2 + 4\eta^2 \left(\frac{n}{3} - \frac{H_1}{2}\right)}}{\eta^2 \left(\frac{n}{3} - \frac{H_1}{2}\right)}
\]  
(A8)

so $g_1(H_1, \gamma, \eta)$ can be calculated for all values of $\gamma$ and $H_1$, obtained from the Laxon et al. dataset and the HadISST dataset respectively, using Eqs. (A6) and (A7). Then
\[
H_{GBM} = g_1\eta^2 \left(\frac{n}{3} - \frac{H_1}{2}\right) + H_1\gamma,
\]

as in Eqs. (A3), (A4).

For four sample months, $H_{GBM}$ was calculated for $\eta = 0.5, 0.6, 0.7, 0.8, 0.9, 1.0$, and the resulting fields compared. The largest difference in $H_{GBM}$ between these different options was 28.2 cm. For the final calculation, $\eta = 0.75$ was used.

A3 Producing the figure; discussion of error

For Fig. 2d, the transformed fields were plotted together; where a gridbox contained data from both fields, a simple mean was used. The field is a time mean of all months from January 1994 to December 2000; in order to reduce the bias of missing summer
months in the satellite field, gridpoints were filled in using the mean March ice thickness value with the seasonal cycle in ice thickness found by Rothrock et al. added (but cut off at 0 m). There are obviously large errors associated with both fields; for the satellite field these stem from: (i) uncertainty in the value of \( \eta \); (ii) the assumptions leading to the equation \( g(h) = g_1h \) for thin ice; (iii) incomplete coverage of certain areas for some months of the year. For the submarine field the errors principally stem from the multiple regression itself; they are discussed in Rothrock et al. (2008), who estimate the standard deviation of variation not explained by their model to be 25 cm. For the (transformed) satellite field the errors will be larger still, hence the need for the figure to be used for illustrative purposes only.

Appendix B

How the Arctic heat budget was calculated

B1 Vertical fluxes

The vertical fluxes were calculated directly from diagnostics; all are positive downwards. Top-of-atmosphere flux

\[
\text{TOA} = \text{SW}_{\text{down}} - \text{SW}_{\text{up}} - \text{LW}_{\text{up}},
\]  

(B1)

where \( \text{SW}_{\text{down}} \) denotes downward shortwave radiation at the top of the atmosphere, and similarly for \( \text{SW}_{\text{up}} \) and \( \text{LW}_{\text{up}} \); atmosphere-to-ocean heat flux

\[
\text{AO} = (\text{penetrating solar flux}) + (htn),
\]  

(B2)

where \( htn \) is a diagnostic representing all non-solar fluxes from atmosphere to ocean. Atmosphere-to-ice heat flux

\[
\text{AI} = \text{topmelt} + \text{botmelt} + q_{\text{ice}}(\text{sub–snow}),
\]  

(B3)
where $q_{\text{ice}}$ is the specific latent heat of melting, topmelt is the top melting heat flux, botmelt is the diffusive heat flux through the ice, and sub, snow are heat fluxes due to sublimation from and snowfall on to the ice respectively. Ocean-to-ice heat flux

\[ OI = \text{ocnmelt} + \text{frazil+misc}, \]  

(B4)

where (ocnmelt) denotes diffusive ocean-to-ice heat flux, (frazil) heat flux due to frazil ice formation, and (misc) small “correction” terms that pass heat between the ocean and ice when ice disappears, and when snow falls into the ocean during ice ridging.

**B2 Horizontal fluxes and heat uptake**

The horizontal transports into the Arctic are more difficult to compute. The atmospheric heat transport (AHT) into the Arctic was computed as a residual \( (AHT = AO + AI - TOA) \), assuming that the atmospheric heat uptake was negligible.

The ice heat transport and uptake (IHT, IHU), were calculated using “rate of change” diagnostics:

\[ \text{IHT} = q_{\text{ice}} \left( \rho_{\text{ice}} \int \left[ \frac{dh_{\text{ice}}}{dt} \right]_{\text{dyn}} .dA + \rho_{\text{snow}} \int \left[ \frac{dh_{\text{snow}}}{dt} \right]_{\text{dyn}} .dA \right) \]  

(B5)

\[ \text{IHU} = q_{\text{ice}} \rho_{\text{ice}} \int \left( \left[ \frac{dh_{\text{ice}}}{dt} \right]_{\text{dyn}} + \left[ \frac{dh_{\text{ice}}}{dt} \right]_{\text{therm}} + \left[ \frac{dh_{\text{ice}}}{dt} \right]_{\text{ridge}} \right).dA \]

\[ + q_{\text{ice}} \rho_{\text{snow}} \int \left( \left[ \frac{dh_{\text{snow}}}{dt} \right]_{\text{dyn}} + \left[ \frac{dh_{\text{snow}}}{dt} \right]_{\text{therm}} + \left[ \frac{dh_{\text{snow}}}{dt} \right]_{\text{ridge}} \right).dA \]  

(B6)

where $\rho_{\text{ice}}$ is ice density, $h_{\text{ice}}$ gridbox mean ice thickness; $\rho_{\text{snow}}$ and $h_{\text{snow}}$ are the analogous quantities for snow $t$ denotes time. The subscripts dyn, therm and ridge indicate rates of change of ice and snow thickness due to advection of ice, thermodynamic processes, and ridging processes respectively.
The oceanic heat transport (OHT) was calculated as

\[ \text{OHT} = \int T_{\text{ocn}} u_{\text{ocn}} \cdot dS \]  

(B7)

where \( T_{\text{ocn}} \) and \( u_{\text{ocn}} \) are ocean temperature and current velocity respectively, and \( S \) is the boundary surface of the Arctic Ocean as defined in Fig. 4.

In order to compute the heat uptake of the ocean (OHU), the annual mean heat content was computed first:

\[ \text{OHC} = \rho_{\text{water}} c_{\text{water}} \int T_{\text{ocn}} \cdot dV \]  

(B8)

where \( \rho_{\text{water}} \) and \( c_{\text{water}} \) are sea water density and specific heat capacity at constant pressure respectively.

The heat uptake for any given year was then calculated to be

\[ \text{OHU}(t) = \frac{\text{OHU}(t + 1) - \text{OHU}(t - 1)}{2} \]  

(B9)

While this is obviously a very inaccurate method of calculating year-on-year heat uptake, the long term trends in heat uptake will be captured. It is very difficult to compute the exact heat uptake in the ocean for a given year using annual mean diagnostics. While monthly mean ocean temperature diagnostics were available, it was chosen not to use these for reasons of data storage and computational time, and because annual mean data was considered sufficient to compute decadal means.

**B3 Residuals**

To evaluate errors in the calculation, a “residual” term can be calculated for each of the atmosphere, ice, and ocean – the energy entering the system that is left over, and does not appear in the heat uptake. Because the atmospheric heat transport into the Arctic
is itself defined as the residual of the other terms, this was zero in the atmosphere component. It was extremely small in the ice component ($\sim O(1GW)$) due to the accuracy of the methods used for calculating the into-ice fluxes. The methods used for calculating OHU and OHT are less accurate; the residual was therefore of significant size, and is plotted on Figs. 5c and 4f with the other ocean fluxes.

Appendix C

A model of the ice export negative feedback effect

We say the Arctic contains $V(t)$ cubic metres of ice at any time $t$, and that volume changes according to advective and thermodynamic effects, specifically:

$$\frac{dV}{dt} = -\mu V + \lambda (V_0 - V)$$

(C1)

where $-\mu V$ and $\lambda (V_0 - V)$ represent ice volume change due to advective and thermodynamic effects respectively, and $\lambda$ and $\mu$ are positive constants. Thus ice export is directly proportional to ice volume, and thermodynamic effects act to “push” ice volume towards a value $V_0$ in which the ice is in thermodynamic equilibrium. (1) solves easily to give an exponential solution in which $V$ approaches $V_0 \times \frac{\lambda}{\lambda + \mu}$, the “steady state” volume.

We now simulate a constant rate of ice melt by replacing $\lambda (V_0 - V)$ with $\lambda (V_0 (1 - t/T) - V)$ in (1), such that the “thermodynamic equilibrium volume” decreases linearly from $V_0$ to 0 in time $T$, and require that $V(0) = V_0 \times \frac{\lambda}{\lambda + \mu}$, the steady state volume at time $t = 0$. This also solves exactly to give

$$V(t) = V_0 \times \frac{\lambda}{\lambda + \mu} \times \left[ 1 - \frac{t}{T} + \frac{1}{T(\lambda + \mu)} \times \left( 1 - e^{-(\lambda + \mu)t} \right) \right]$$

(C2)

Thus $\frac{d^2V}{dt^2} = -\frac{\lambda V_0}{T} \times e^{-(\lambda+\mu)t}$, which is always negative, and the rate of ice loss never slows.

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References


Table 1. Statistics of September ice extent timeseries in the ALL and ANT ensembles and in HadISST observations, from 1979–2010. All values are in millions of square kilometres.

<table>
<thead>
<tr>
<th></th>
<th>Mean value</th>
<th>Annual linear trend</th>
<th>Detrended stddev</th>
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</thead>
<tbody>
<tr>
<td>ALL</td>
<td>6.50</td>
<td>−0.078</td>
<td>0.46</td>
</tr>
<tr>
<td>ANT</td>
<td>6.74</td>
<td>−0.059</td>
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<tr>
<td>HadISST</td>
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<td>−0.077</td>
<td>0.53</td>
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</table>
Fig. 1. (a) Arctic September mean sea ice extent in the HadGEM1 historical forcing runs; (b) 15-yr running gradient of extent.
Fig. 2. (a), (b), (c) show linear trend in September ice concentration (yr$^{-1}$) from 1979–2011 in the HadISST observations, and in the HadGEM1 ALL and ANT ensembles respectively. (d), (e) and (f) show mean ice thickness from 1994–2000 in a superposition of satellite and submarine observations from that period (described more fully in Appendix A), and in the ALL and ANT ensembles respectively.
Fig. 3. (a) shows annual mean global temperature (°C) plotted against September mean ice extent (x10^6 km^2); (b) shows annual mean Arctic temperature (mean temperature north of 70° N) plotted against September mean ice extent, for the ALL and ANT ensembles.
Fig. 4. Schematic to show how the Arctic heat budget was calculated. The bottom panel shows the Arctic Ocean as defined in Sect. 3. The Western Arctic (WA) and Eastern Arctic (EA) regions used in Sect. 3.3 are also shown.
Fig. 5. Decadal means of terms in the Arctic heat budget for the ALL ensemble mean (a), (b) and (c)) and the ANT ensemble mean ((d), (e) and (f)). For each term, the time mean of the equivalent periods in the control runs from 1980–2030 has been subtracted.
Fig. 6. Ice export from the Arctic Ocean in the HadGEM1 runs. Showing (a) exact ice export as calculated in the heat budget; (b) ice export through the Fram Strait, calculated as $\int h_{\text{ice}} v_{\text{ice}} \, dS$; (c) ice export calculated as in (b), but with the time-varying $v_{\text{ice}}$ fields replaced by their time mean; (d) ice export calculated as in (c), but with the $h_{\text{ice}}$ fields replaced by their time mean, plus the timeseries of anomalies of ice thickness in the Western Arctic region as defined in Sect. 4. (e) Shows mean ice thickness in the Western Arctic region, with 15-yr running gradients.
Fig. 7. Ocean heat transport into the Arctic Ocean in the HadGEM1 runs. Showing (a) OHT=$\int T_{\text{ocn}} \ u_{\text{ocn}} \ dS$, as calculated in the heat budget; (b) as in (a), with velocity fields replaced with the time mean; (c) as in (a), with temperature fields replaced by the time mean. Each figure shows ocean heat transport (top sets of lines, left scales) and 15-yr running gradients (bottom sets of lines, right scales).
Fig. 8. Variation of the MOC (dark colours, top set of lines) and the Sub-Polar Gyre (light colours, bottom set of lines) in the HadGEM1 experiments. The SPG index of ALL 4 has been highlighted from 2006–2021.
Fig. 9. Timeseries of monthly mean Arctic sea ice extent anomalies from the HadISST data set, 1981–2000. Anomalies are taken relative to the 1981–2000 mean extent for each respective month.