Deciphering the spatio-temporal complexity of climate change of the last deglaciation: a model analysis

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Abstract. Understanding the sequence of events occurring during the last major glacial to interglacial transition (21 ka BP to 9 ka BP) is a challenging task that has the potential to unveil the mechanisms behind large scale climate changes. Though many studies have focused on the understanding of the complex sequence of rapid climatic change that accompanied or interrupted the deglaciation, few have analysed it in a more theoretical framework with simple forcings. In the following, we address when and where the first significant temperature anomalies appeared when using slow varying forcing of the last deglaciation. We used here coupled transient simulations of the last deglaciation, including ocean, atmosphere and vegetation components to analyse the spatial timing of the deglaciation. To keep the analysis in a simple framework, we did not include freshwater forcings that potentially cause rapid climate shifts during that time period. We aimed to disentangle the direct and subsequent response of the climate system to slow forcing and moreover, the location where those changes are more clearly expressed. In a data – modelling comparison perspective, this could help understand the physically plausible phasing between known forcings and recorded climatic changes. Our analysis of climate variability could also help to distinguish deglacial warming signals from internal climate variability. We thus are able to better pinpoint the onset of local deglaciation, as defined by the first significant local warming and further show that there is a large regional variability associated with it, even with the set of slow forcings used here. In our model, the first significant hemispheric warming occurred simultaneously in the North and in the South and is a direct response to the obliquity forcing.

1 Introduction

The last deglaciation – ⃒21 to ⃒9 kyr Before Present (BP) – is Earth’s most recent transition from a glacial-like climate to an interglacial-like climate, a type of transition that occurred repeatedly with a periodicity of ⃒100 kyr over the late Quaternary (Hays et al., 1976; Waelbroeck et al., 2002). Milutin Milankovitch was one of the first to propose that this low-frequency variability of the climate system is linked to the variations of the orbit of the Earth around the Sun, thereby modifying the energy received at the top of the atmosphere in summer. He proposed that summer insolation at high northern latitudes could be considered as the main driver of the ice-age cycles as it constrained the capacity of winter snow to survive the summer and hence contributed to the buildup of glacial ice-sheets. During peak glacial periods like the Last Glacial Maximum (LGM) most of North America was covered with a ⃒4 km thick ice-sheet (Dyke et al., 2002; Peltier, 2004) and a good part of northern Europe and western Siberia (Svendsen et al., 2004) as well. Theorbitally-forced changes in insolation received by the Earth are the only long-term forcing truly external to the Earth’s climatic system, whereas ice-sheet waxing and waning and greenhouse gases that strongly affect the climate over similar time periods are only internal feedbacks to that one forcing. The response of the Earth’s system is non-linear and the exact timing of the deglaciation may also be set by a threshold crossing, as is suggested by several authors (Paillard, 1998; Barker et al., 2009; Lamy et al., 2007; Wolff et al., 2009).

Over the years, compelling evidence of how drastic climate changes have been through the last deglaciation have arisen from proxy data retrieved from geological records throughout the world (MARGO Project Members, 2009; North Greenland Ice Core Project members, 2004; EPICA community members, 2004). Although there is no doubt
that this last transition has affected the Earth as a whole, there is still some debate on how changes relate to each other at different geographical locations on Earth (Stott et al., 2007; Huybers and Denton, 2008; Timmermann et al., 2009). Though such debate could in principle be lifted by absolute dating of proxy records and perfect understanding of what is recorded in those proxies, the current science is not there yet.

We therefore propose to help with understanding the sequence of climatic changes of the last deglaciation by performing and analysing results from a model simulation to assess within the physical processes contained in our climate model, when, why and where the climate started to warm in an experiment forced by low-frequency variability arising from greenhouse gases, orbital and ice-sheet distribution changes. We also define a time-period in years needed to distinguish between a large local climate change (such as deglaciation) and local interannual or centennial variability.

2 Experimental setup

2.1 Model description

In the present study, we use the LOVECLIM earth system model of intermediate complexity in its version 1.0 (Driesschaert et al., 2007). In the version applied here, components for atmosphere (ECBilt), an ocean (CLIO) and vegetation (VECODE) are activated. It is a follow-up of the ECBilt-CLIO-VECODE coupled model that has been successful in simulating a wide range of different climates from the Last Glacial Maximum (Roche et al., 2007) to the future (Driesschaert et al., 2007) through the Holocene (Renssen et al., 2005, 2009) and the last millennium (Goosse et al., 2005). The atmospheric component (ECBilt) is a quasi-geostrophic model at T21 spectral resolution (≈5.6° in latitude/longitude) with additional parametrizations for the non-geostrophic terms (Opsteegh et al., 1998). ECBilt has three vertical layers in which only the first contains humidity as a prognostic variable. Precipitation is computed from the precipitable water of the first layer and falls in the form of snow if the temperature falls below 0°C. The time step of integration of ECBilt is 4h. The oceanic component (CLIO) is a 3-D Oceanic General Circulation Model (Goosse and Fichefet, 1999) run on a rotated B-grid at approximately 3° × 3° (lat-lon) resolution. It has a free surface that allows the use of real freshwater fluxes, a parametrization of downsloping currents and a realistic bathymetry. CLIO also includes a dynamical-thermodynamical sea-ice component (Fichefet and Morales Maqueda, 1997, 1999) on the same grid. The interactive vegetation component used is VECODE (Brovkin et al., 1997), a simple dynamical model that computes two Plant Functional Types (PFT: trees and grass) and a dummy type (bare soil). The vegetation model is resolved on the atmospheric grid (hence at T21 resolution) and allows fractional allocation of PFTs in the same grid cell to account for the small scale needed by vegetation. The different modules exchange heat, stress and water. It should be noted that there is a precipitation correction needed to avoid the large overestimation of precipitation over the Arctic and the North Atlantic that is present in ECBilt. This surplus of fresh water is removed from the latter regions and is added homogeneously to the North Pacific surface (cf. Goosse et al. (2010) on this aspect). The advantages of the LOVECLIM model when compared to other EMICs are the 3-D oceanic general circulation model and the dynamical atmosphere with actual moisture transport; other models are often energy-moisture balance models.

2.2 Deglacial forcings

Our goal is to perform a transient simulation of the last deglaciation, from the Last Glacial Maximum (LGM, around 21 kyr BP) to the early phase of the Holocene period (around 9 kyr BP). It shall be noted that there is still some ice present in North America over the Quebec region at this last date; the Northern Hemisphere ice-sheets reaching a near present-day extent around 7 kyr BP (cf. Renssen et al., 2009, 2010 for an analysis of the impact of the remnants of the Laurentide ice-sheet on the climate evolution of the Holocene). We start our integration at the LGM from the climatic state described in Roche et al. (2007). From 21 kyr BP onwards, we force the model with insolation changes arising from the long-term changes in orbital parameters (the so-called “Milankovitch forcing”), greenhouse gases changes and ice-sheet distribution, since our model version does not include an interactive ice-sheet component. The orbital parameters are taken from Berger (1978). For greenhouse gases, we prescribe changes in carbon dioxide, methane and nitrous oxide as recorded in air bubbles from ice cores (cf. Fig. 1). Ice-sheet evolution is taken from the ICE-5gV1.2 reconstruction (Peltier, 2004) for both northern and Southern Hemisphere ice-sheets, and interpolated on the T21 grid of the atmospheric component of our coupled climate model. Between the given ICE-5gV1.2 time slices reconstructions, we linearly interpolate in time with a time step of 50 yr. We both prescribe the orography and ice-mask so as to ensure their joint evolution during the deglaciation run, whereas the land-sea mask is kept fixed at LGM. Indeed, it is not obvious how changes in the land-sea mask should be taken into account from the oceanic perspective in order to properly conserve mass, momentum and salinity. Using this approach means that the Barents and Kara seas but also the Hudson bay remain as land throughout and that the Bering strait is kept closed at all times. This means that climatic anomalies in regions like the Bering Strait, the Barents Sea or the Argentinian shelf should be regarded with caution. Similarly, the bathymetry of the ocean was kept to LGM conditions, that is we reduced its depth by 120 m, with a cut-off to the closest vertical model level. This is known to have important implications for the sensitivity of the oceanic
As we focus here on the long-term changes, we tested the first significant warming occurring after the LGM at every location. The first significant warming is, a priori, an asynchronous event in each grid cell of the model, though some regional patterns are expected to emerge. In reality, this first significant warming would be the first detectable warming in the temperature recorded by any method. In the following, we will define the first significant warming using a statistical test. It requires the knowledge of the “internal” (modeled) variance of the LGM climate, computed here from the last 500 yr of an equilibrium run under constant LGM boundary conditions. Our 12 000 yr deglaciation run is first divided in 120 samples of 100 yr that we tested independently with respect to the control LGM climate. We also performed the analysis with samples of 25 and 200 yr to assess the robustness of the method. In the following, we first performed a standard Fischer test on the variances to assess whether they differed or not. When sample variances were equal, we tested against the specified period. From a statistical point of view, the reference LGM period is thus the null hypothesis compared to the deglacial sample considered. In the following, we consider anomalies that are significant at a 5 % level, that is when $t_{\text{value}} > 1.962$ for a sample of 100-year (two-sided t-test as we test directly if temperatures are above or below the mean, with reference to both the observed and the modelled temperature at the NorthGRIP until around 16 ka BP, when a sudden cooling in Greenland interrupted the warming trend. This cooling has been associated with the North Atlantic Heinrich Event 1 (cf. Hemming (2004) for a review) that modified the sea surface conditions by the addition of excess freshwater to that area. The subsequent sequence of events was likely responding or forced in the same manner. As we do not include the appropriate forcing for such events, we will focus in the following on the long-term trend in climate in a more abstract framework. A detailed data–model comparison will be the focus of further studies.

### 3 Analysis method

Analysing climate change throughout the last deglaciation is complex and could be based on different variables (temperature, precipitation, etc.). The most obvious change that comes to mind when thinking of deglaciation is warming. We thus chose to concentrate on the phasing of climate evolution throughout the last deglaciation, with a focus on the first significant warming occurring after the LGM at every location. The first significant warming is, a priori, an asynchronous event in each grid cell of the model, though some regional patterns are expected to emerge. In reality, this first significant warming would be the first detectable warming in the temperature recorded by any method. In the following, we will define the first significant warming using a statistical test. It requires the knowledge of the “internal” (modeled) variance of the LGM climate, computed here from the last 500 yr of an equilibrium run under constant LGM boundary conditions. Our 12 000 yr deglaciation run is first divided in 120 samples of 100 yr that we tested independently with respect to the control LGM climate. We also performed the analysis with samples of 25 and 200 yr to assess the robustness of the method. In the following, we first performed a standard Fischer test on the variances to assess whether they differed or not. When sample variances were equal, we tested the means with a standard Student t-test. When not, we made use of a t-test with two unequal variances defined as (Welsh’s test):

$$t_{\text{value}} = \frac{\chi_{\text{ref}} - \chi_{\text{sample}}}{\sqrt{\frac{\sigma^2_{\text{ref}}}{N_{\text{ref}}} + \frac{\sigma^2_{\text{sample}}}{N_{\text{sample}}}}}$$

where $\chi$ denotes the mean of the climatic variable over the considered period, $N$ the size (in timesteps) of the period and $\sigma^2$ the variance of the climatic variable. “ref” denotes the reference period (LGM) while “sample” denotes the sample tested against the specified period. From a statistical point of view, the reference LGM period is thus the null hypothesis compared to the deglacial sample considered. In the following, we consider anomalies that are significant at a 5 % level, that is when $t_{\text{value}} > 1.962$ for a sample of 100-year (two-sided t-test as we test directly if temperatures are above or below the mean, with reference to both the observed and the modelled temperature at the NorthGRIP until around 16 ka BP, when a sudden cooling in Greenland interrupted the warming trend. This cooling has been associated with the North Atlantic Heinrich Event 1 (cf. Hemming (2004) for a review) that modified the sea surface conditions by the addition of excess freshwater to that area. The subsequent sequence of events was likely responding or forced in the same manner. As we do not include the appropriate forcing for such events, we will focus in the following on the long-term trend in climate in a more abstract framework. A detailed data–model comparison will be the focus of further studies.
Fig. 2. Insolation anomaly to the 0–30 kyrs BP mean for the last deglaciation computed from Berger (1978) for (A) the annual mean, (B) the northern summer, (C) the northern winter. For (B) and (C) the summer (winter) is computed as the second (fourth) quarter of year after the spring equinox.

less than the mean). We will consider significant temperature anomalies at a given time or the timing of the first significant anomaly as a marker of the local start of the deglaciation period as modeled with the imposed slow forcings.

4 Results for Surface Air Temperature (SAT) evolution

4.1 Annual mean

In the following we concentrate on a 100-year sample for discussion. Figure 4 introduces the spatial distribution of the timing of first significant warming from 21 ka BP onwards.

The first regions to respond (between 21 and 20 ka BP) are the Arctic Ocean, the northernmost part of Siberia and patches in the Southern Ocean. These are regions affected by the presence of sea-ice, or neighbouring continental regions. The first response is immediately followed (20 to 19 ka BP) by a significant response of all latitudes poleward of 35° north and south. During that given period of time, the only forcing is the orbital forcing, greenhouse gases and ice-sheet forcing being quasi constant (CO₂ concentration changes are ≃5 ppm). Sea-ice being sensitive to the total amount of energy received throughout the year, increasing the energy received in any season will limit (or even reduce) the sea-ice extent and the buffering effect of the underlying ocean will extend (in time) the anomaly to a year-round effect. The early response seen in the northern North Atlantic and adjacent regions is therefore an effect of the obliquity increase during the early part of the deglaciation that increases the total amount of energy received by the Earth at high latitudes, as depicted in Fig. 2. The fact that the early significant warmings observed in the model are obliquity-driven is further reinforced by the in-phase changes of the Northern
and Southern Hemispheres. This warming is accompanied by a reduction in the sea-ice extent around Antarctica, a result found in other simulations of the last deglaciation (Timmermann et al., 2009). We do not find any delay between the Northern and Southern Hemisphere responses in our model. The sea-ice change in the South therefore responds primarily to the local orbital (obliquity) forcing and not to a delayed response to the North Atlantic warming through upwelled waters as found for other climatic periods (Duplessy et al., 2007; Renssen et al., 2010). By increasing the total energy received from the sun at high latitudes, the obliquity signal forces an in-phase response of both hemispheres at high latitudes regions (cf. Fig. 2).

A later response (17 to 15 ka BP) is then observed in most oceanic tropical regions. Given the simplified representation of the physical equations for motion in the atmospheric part of our model, caution is needed in interpreting this pattern. We observe some changes in the precipitation pattern at the same time (cf. Sect. 4.3) that may be linked to ITCZ changes in response to the changing Equator-to-pole gradient as well as change in ice-sheet topography. However, a precise assessment of what is occurring in the tropics would require a model with more complex atmospheric physics (Khodri et al., 2009). The time period around 16 ka BP is also a period when the global greenhouse gas forcing starts to become significant (CO₂ at around 220 ppm, cf. Fig. 1) enough to counterbalance the obliquity-induced cooling of the tropics (cf. Fig. 2). The later response of the tropical regions are also due to the rather small absolute temperature changes from 21 to 9 ka BP (cf. Fig. 5). It is therefore difficult to discriminate between a small temperature change and year-to-year variability within the model in such areas. Two different types of regions are lagging the response of the rest of the planet.

The first type are areas where the year-to-year variability – as characterized by the sample variance – is higher during the deglaciation and the local temperature change over the deglaciation is not that large. Thus, the deglacial warming becomes significant only late in the deglaciation. Characteristic examples are the continental regions of India and China and the tropical Pacific ocean, becoming significant only after 15 ka BP (with the notable exception of the Himalayas). Figure 6 shows the temperature simulated for one location in the Pacific Ocean (15° N, 120° W) for the LGM and 9 ka BP. The two density distributions are relatively well separated, but not to a 5% significance level. The deglacial sample has a larger variance (variability) than the glacial one, as characterized by the width of the density peak. Within the sample, some years cannot be statistically distinguished from one another as for example a series of 20 yr between years 20 and 40. Thus, one can argue that the climate depicted by those two samples is not very different at a 5% confidence level, i.e. a relatively high confidence level.

The second type are regions where the deglacial warming from 21 ka to 9 ka is never significant (shown by black shading on Fig. 4). They are located in equatorial regions in Africa and south America, offshore California in the northern tropical Pacific Ocean and in the sea of Okhotsk. These regions are simulated to be colder on an annual mean at 9 ka BP than during the LGM (cf. Fig. 5) and thus never encounter lasting, significant deglacial warming.

4.2 Seasonal means

To pursue a more in-depth analysis on the complex seasonal timing of deglacial warming, Fig. 7 presents results for December-January-February (DJF, northern winter) and June-July-August (JJA, northern summer).

DJF shows the largest areas of non-significant warming over the deglaciation. The reason for this is similar to the late warming previously described, i.e. the local variability is too
Fig. 6. Comparison of two temperature samples from the deglaciation in the Pacific Ocean, offshore Mexico (15° N, 120° W). The left panel shows the annual mean temperature evolution over a 100-year sample taken from the reference LGM run (in red) and from the deglaciation (at 12 ka BP, in blue). The right panel shows the density function associated to that sample on the same temperature axis. The dashed vertical lines in the right panel show the means of the series, the filled areas are the respective intervals corresponding to the standard deviations of each series (±σ).

high for the local warming to become statistically significant. This may be interpreted as regions where a 100-year mean is more representative of interannual to decadal variability than of climate in the sense of a 30-year mean. The time-length of the sample needed for the warming to become significant during the transition is discussed in Sect. 4.4. For a relatively large area centered on the Bering Strait as well as for the Gulf of Mexico, there is no significant warming in DJF at 9 ka BP relative to the LGM as the two regions are significantly cooling in our model. Other large areas of non-significant warming (part of the eastern Pacific, continental tropical regions and eastern Eurasia) are characterized by small temperature anomalies as a whole (below 1 °C in DJF), a change that is hardly significant with respect to the model interannual variability in the same regions. We nonetheless note the early response of sea-ice regions, first in the northern North Atlantic (19 ka BP) and of the Southern Ocean sea-ice north of 60° S, better marked than in the yearly mean. Conversely, JJA shows the smallest non-significant warming from 21 k to 9 k. A striking feature is that most regions have a significant warming early in the deglaciation mostly before 18 ka BP. Three large areas are standing out as being earlier than that: the northern North Atlantic, the Southern Ocean around 60° S and the northern Equatorial regions. The first is due to sea-ice changes and circulation changes as was noted before, followed by the neighbouring Arctic. Accordingly, the Southern Ocean region is linked to shrinking sea-ice winter extent.

Fig. 7. Timing of first significant warming during the deglaciation from a 100 yr sample at 5% significance. DJF (top) and JJA (bottom). Color scale is the date in kyr BP Black denotes regions without significant warming over the deglaciation.

4.3 Precipitation evolution

Not all proxies for climate change are primarily sensitive to temperature changes during the last deglaciation. The same is true for specific regions, for instance the intertropical regions, where the main response to the last deglaciation is likely to be a change in annual precipitation, not annual warming (Roche et al., 2007). Performing the same kind of analysis, we obtain a geographical distribution of the significant increase (decrease) in precipitation. It should be noted that the precipitation distribution is not normally distributed, but that the logarithm of precipitation is (Vrac et al., 2007). The analysis performed in this section is thus comparable to the previous ones using the logarithm of precipitation as a variable. To simplify the interpretation, we also mask the regions where precipitation changes are significant for both an increase and a decrease. That is, we retain areas where there is only an increase (or decrease) in precipitation throughout the deglaciation. The complete sequence of events from the precipitation point of view would require a dedicated study involving an analysis of the Intertropical Convergence Zone (ITCZ) movements through time.
The annual mean precipitation (Fig. 8) shows a significant decrease in a zonal belt in the southern equatorial regions during 20 and 16 kyrs BP. This pattern is due to the northward shift of the ITCZ in response to the warmer climate conditions. Indeed, under LGM climate conditions, most models simulate a southward shift of the ITCZ in response to the imposed boundary conditions (Braconnot et al., 2007), a shift consistent with data evidence (Leduc et al., 2007). This shift was shown to respond primarily to global temperature changes (Khodri et al., 2009). As the beginning of the deglaciation (21–16 kyrs BP) is marked by relatively low greenhouse gases changes, we observe an ITCZ shift mainly in response to the change in insolation forcing.

The annual mean figure for a significant increase in precipitation shows a more complex pattern. We can note that there are many more areas with increased precipitation than with decreased precipitation. Indeed, as the atmosphere warms during deglaciation, it can hold more moisture. The global LGM to early Holocene change in moisture content is thus towards an increase, thus leading to an increase in precipitation. We can distinguish four different areas. First is the very early change over the Arctic regions; during the LGM these regions are very dry. Thus, a small increase in precipitation at the start of deglaciation is immediately significant. Second, the northern equatorial regions show a relatively zonal pattern; the latter is a counterpart of the decrease seen in the southern equatorial regions and is likely responding to the same ITCZ shift. The timing of this increase is coherent with the decrease seen previously. Third, the southern mid-latitudes consistently show an increase in precipitation between 20 and 16 kyrs BP. This is coherent with the regions previously identified in the temperature fields as early warming (around 60° S) and bears a link to seasonal changes in the sea-ice field. Fourth, the desertic and semi-desertic regions (Sahara, Arabia, Pakistan etc.) display an increase in precipitation starting 17–16 kyrs BP. This pattern is generally consistent with an observed transition to more humid conditions in northern Africa (Gasse, 2000; deMenocal et al., 2000; Tjallingii et al., 2008; Timm et al., 2010) in response to warmer conditions. However, the role of abrupt events of the last deglaciation (that we do not take into account) calls for caution on the exact timing of the increase.

4.4 Impact of interannual variability

As noted before, our results are sensitive to the sample size used in the study. Indeed, increasing the sample size comparatively reduces the effect of noise (variability) in the model on the definition of the mean temperature of the sample. A large sample is thus less affected by a series of years with temperature above the mean than a smaller sample. Using a small sample size (e.g. 25 yr), our results therefore emphasize the potential for interannual variability anomalies to be significant at a long-term scale (deglaciation scale). Using different sample size (25, 100, and 200 yr in the following), we may truly assess what is the timing of climate change in the model and decipher regions where the interannual variability is too large to allow significant climatic anomalies on those longer timescales. Figure 9 compares the timing of first significant warming for four different samples of increasing size.

An evident feature arising from Fig. 9 is that a shorter sample yields generally a later significant warming. This results from the fact that the value of the Welch’s test depends strongly on the sample size to determine the significance of the anomaly: If the sample is relatively small and the variability within the sample is large or larger than the reference period, then a larger temperature anomaly is needed to stand out compared to the local variability (cf. Welch’s test equation). Increasing the sample size thus decreases the importance of the internal variability over the signal and enables a more accurate determination of the first significant, externally forced, warming. In simpler terms, this can be interpreted as looking at climate compared to looking at internal high-frequency variability: With a small sample having a large variability, one needs a very different sample mean to be significantly different from the reference.

Most interestingly, the size of the sample needed to discuss the climate anomaly versus the reference climate is variable.
spatially. Indeed, both the absolute temperature anomaly and local temperature variability vary in space. Two examples can be taken from Fig. 9 to highlight this feature.

1. In the northern tropical regions over the Pacific and southwestern North America, the total temperature anomalies from 21 ka BP to 9 ka BP (cf. Fig. 5) are relatively small in our model, below 2°C. Distinguishing those small anomalies from a larger interannual variability (that is in a sample with a large variance) is therefore difficult and requires a larger sample. One can note that even with a 200-year sample, not every location in those areas is significantly warmer at 9 ka BP than at LGM.

2. Most regions of central Asia become significantly warmer only late in deglaciation using a 25-year sample. In this case, this is not solely the effect of a small LGM to 9 ka BP temperature difference (some areas have a temperature anomaly of about 10°C) but because of very large variance within the sample, related to high interannual to centennial variability. In fact, the variance of the samples during deglaciation are systematically higher than those of the reference run, making it harder to decipher a climate change from internal high-frequency variability (noise). It should be noted also that the sample size strongly affects the date of first significant anomaly in this area: Central Asia is significant in the 25-year sample only at 14 ka BP, is significant between 17 and 15 ka BP in the 100-year sample and around 18 ka BP in the 200-year sample.

The analysis of Fig. 9 confirms our previous inferences that there are three main areas with leading temperature changes: the northern North Atlantic, a north Equatorial band and the Southern Ocean between 45 and 60° S.

5 Discussion

Our study so far focusses on a single 12,000 yr run with slow forcings included. To reproduce the effects of millennial-scale climate variability, the modeling study would require the use of higher frequency forcings, such as freshwater fluxes from melting ice-sheets, or understanding how the response to slow forcings can act to produce abrupt events through the non-linearities of the climate system as is recorded in many different proxies (North Greenland Ice Core Project members, 2004; Shackleton et al., 2000; Wang et al., 2001; von Grafenstein et al., 1999, for example). How could we proceed to better determine the response of our climate model to the (imposed) slow forcings? One often-used method (Goosse et al., 2005, for example) is to perform ensemble simulations with identical forcings, varying only the initial conditions. The different expression of the internal variability of the model in the different ensemble members would then cancel out in the mean, leading to a more robust response of the forced response. However, changing our
approach of temporal samples to ensemble samples would require 100 simulations of the full deglaciation period. This is difficult to obtain due to computational constraints. We are thus limited to a single run for the time being.

Natural (observed) climate, on the other hand, is only one trajectory out of many possible solutions. Analysing a single simulation is therefore close to what is recorded by proxy data, albeit that we have a perfect recording of our simulated climate within our idealized “model world”, as opposed to the imperfect recording of the Earth’s climate in proxy data. We have shown that even with perfect recording of the simulated climate, there are regions where distinguishing between the deglaciation warming and local variability is problematic. Depending on the resolution of the signal recorded in the proxy, a similar issue may arise. What we show from our model simulations is that when the signal is very noisy, high temporal resolution is safer to determine variability and climate evolution. However, high resolution is practically limited by the type of proxy record chosen. For example, recording $\delta^{18}O$ in oceanic sediment cores from foraminifera have a maximal practical resolution of about ten years for glacial periods (depending on foraminifera abundance, sedimentation rates etc.). Furthermore, averaging the values of five specimens does not guarantee the consistence of five subsequent and equal periods of time within one sample of the core. Thus, analysing an oceanic sediment core at 100 yr resolution is not equivalent to obtaining the 100 yr mean of the signal. Similar examples could be taken from different compartments of the earth system, with e.g. pollens in terrestrial cores. The relationship between the mean of the recorded proxy and the local variability is complex. Our results are indicative of regions where the relationship between average climate change and variability is likely to be complicated by the amplitude of the latter.

Finally, the reader should not forget that the results presented have been obtained with one climate model and are only indicative of what is physically plausible within the framework of the given model. There is a need to repeat such approaches with different models to identify regions where it is likely that the high local variability will hamper our capability to record the mean climate changes and how such local climate variability is evolving through time. The regions (like the Pacific coast of Siberia) highlighted here are indicative with respect to the mechanisms occurring but are limited to the climate model used. Extension to the real climate system should be done with caution.

6 Conclusions

A number of conclusions arises from our analysis.

First, the first regions that are showing a significant temperature evolution during the deglaciation are sea-ice covered regions in both the Northern and Southern Hemispheres. This points to a crucial importance of sea-ice in setting the timing for deglaciation, as well as in constraining feedbacks mechanisms that will lead to further warming and deglaciation. The understanding of sea-ice evolution is most likely crucial in that sense, though probably more via the annual production of sea-ice (Paillard and Parrenin, 2004; Bouttes et al., 2010) than through the absolute sea-ice cover (Stephens and Keeling, 2000; Archer et al., 2003). The symmetry of the responses between the Northern and Southern Hemisphere points to the crucial role of obliquity in setting the deglacial timing.

Second, regions that are more “passively” responding to the deglaciation forcings and are remote to the ice-sheet locations are likely to respond with a time delay of $\approx 3000$ yr, that is when a significant global forcing such as greenhouse gases sets in. This delay is to be understood within a slowly varying forcing framework. In a simulation with abrupt climate change, the delay would still exist but the pattern would me more complicated to decipher due to the more complex deglaciation signal. Moreover, there is a large spatial variability in the first significant change during the last deglaciation even without abrupt climate changes. Therefore, caution on the spatial structure or robustness is needed when trying to infer leads and lags from existing deglaciation records or model results in these regions before any physical interpretation can be drawn.

Third, regions displaying few glacial to interglacial changes in the considered climatic variable (temperature here) and remote from the “centers of action” of the coupled climate system will not easily record a precise timing for the first change in the deglaciation. The interannual variability, whether in the climate model or in reality, will tend to cloud the true signal as in any noisy record. We have detailed this mechanism here for regions in tropical Asia. There is therefore a high dependence of first warming timing to local variability. In that respect, using long averages of about 100 to 200 yr to describe climate change is a requirement in analysing model results if one wants to avoid biases due to (modeled) variability at shorter timescales. This brings us to the question of what is to be understood as “climate change”: We infer from our simulations that it has to be a time long enough to be detected against background noise. How much precisely will vary spatially and in time, making it harder to decipher long-term climate changes from different climate model simulations – and/or – data proxies? Ultimately, it will vary both with the resolution of the proxy used to record the climate change and with the time window considered.

These conclusions will need to be first substantiated by other model studies to ascertain that the main results are not model dependent. Once the common pattern between different coupled model is established, there will be a need for model – data comparison at the global scale (e.g. Shakun and Carlson, 2010).
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