On the causes of mid-Pliocene warmth and polar amplification

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The mid-Pliocene (~3 to 3.3 Ma ago), is a period of sustained global warmth in comparison to the late Quaternary (0 to ~1 Ma ago), and has potential to inform predictions of long-term future climate change. However, given that several processes potentially contributed, relatively little is understood about the reasons for the observed warmth, or the associated polar amplification. Here, using a modelling approach and a novel factorisation method, we assess the relative contributions to mid-Pliocene warmth from: elevated CO₂, lowered orography, and vegetation and ice sheet changes. The results show that on a global scale, the largest contributor to mid-Pliocene warmth is elevated CO₂. However, in terms of polar amplification, changes to ice sheets contribute significantly in the Southern Hemisphere, and orographic changes contribute significantly in the Northern Hemisphere. We also carry out an energy balance analysis which indicates that that on a global scale, surface albedo and atmospheric emissivity changes dominate over cloud changes. We investigate the sensitivity of our results to uncertainties in the prescribed CO₂ and orographic changes, to derive uncertainty ranges for the various contributing processes.

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1. Introduction

The most recent palaeoclimate reconstructions (Dowsett et al., 2009) suggest that during warm ‘interglacials’ of the Pliocene epoch (~5.3 to 2.6 Ma), global annual mean sea surface temperatures were 2 to 3 °C higher than the pre-industrial era. During these warm interglacials sea levels were higher than today (estimated to be 10 to 30 m) meaning that global ice volume was reduced (e.g. Dowsett and Cronin, 1990; Dwyer and Chandler, 2009; Naish and Wilson, 2009). There were large fluctuations in ice cover on Greenland and West Antarctica, and during the interglacials they were probably largely free of ice (Dolan et al., 2011; Hill et al., 2010; Lunt et al., 2008; Pollard and DeConto, 2009). Some ice may also have been lost from around the margins of East Antarctica especially in the Aurora and Wilkes sub-glacial basins (Hill et al., 2007). Coniferous forests replaced tundra in the high latitudes of the Northern Hemisphere (Salzmann et al., 2008), and the Arctic Ocean may have been seasonally free of sea-ice (e.g. Cronin et al., 1993). The most recent estimates of Pliocene atmospheric CO₂ concentrations range between 280 and 450 ppmv (Pagani et al., 2010; Seki et al., 2010). The Mid-Piacenzian

Warm Period (henceforth ‘mid-Pliocene’; 3.26 to 3.025 Ma BP; timescale of Lisiecki and Raymo (2005)) is a particularly well documented interval of warmth during the Pliocene, with global data sets of multi-proxy sea surface temperatures, bottom water temperatures, vegetation cover, topography and ice volume readily available as boundary conditions and/or evaluation datasets for global climate models (Dowsett et al., 2010b; Haywood et al., 2010).

Many parallels have been drawn between the apparent similarities in climate between warm intervals of the Pliocene and the end of the 21st Century, particularly in terms of (relative to pre-industrial) (a) the change in annual mean global temperature (Haywood et al., 2000a; Jansen et al., 2007), (b) changing meridional surface temperature profiles showing a strong polar amplification of the warming (Dowsett et al., 1992; Robinson, 2009), (c) changing precipitation patterns and storm tracks (Haywood et al., 2000b) and even (d) Hurricane intensity and ENSO-event frequency/extra-tropical teleconnections (Bonham et al., 2009; Fedorov et al., 2010; Scroxton et al., 2011; Watanabe et al., 2011). This attraction is made more intense by the fact the continents had essentially reached their modern position, and due to its relative youth, geologically speaking, inferences about the environmental tolerances of many of the biological proxies used to reconstruct Pliocene environments and climates can be made with far greater confidence than further back in Earth history (Dowsett and Poore, 1996; Salzmann et al., 2008).

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As such, it is particularly important to understand why the mid-Pliocene was warmer than pre-industrial. Up until now, the most comprehensive attempt to answer this question was carried out by Haywood and Valdes (2004), henceforth H&V04. Using the UK Met Office coupled atmosphere–ocean General Circulation model, HadCM3, they carried out a model simulation of the mid-Pliocene, and compared it to a pre-industrial simulation. They found a global mean surface air temperature difference of 3.1 °C. From the assumed CO₂ radiative forcing in the model and consideration of top-of-the-atmosphere radiative fluxes, they partitioned the causes of this temperature difference between CO₂ (1.9 W m⁻²), surface albedo (2.3 W m⁻²) and cloud cover (1.8 W m⁻²) changes. They further partitioned the surface albedo component between land ice and snow (55%) and sea ice (45%) changes. From interrogating the ocean stream function and net heat transports, they also concluded that ocean circulation changes did not lead to significant surface temperature warming. Given the considerable computational constraints at the time (the 300 yr simulation took 9 months to complete), the H&V04 study contributed significantly to our understanding of the causes of mid-Pliocene warmth. However, the fact that further sensitivity studies could not be carried out meant that cause and effect was not easily partitioned. For example, the albedo change due to sea ice was itself a result of the imposed CO₂ (and orography, and vegetation, and land-ice) changes. Similarly for clouds — some of the cloud changes would be due to the land ice (and other) changes. In this paper we address this issue, by describing a new methodology for a robust, self-consistent partitioning of climate change between several causal factors. We then apply it to the warm periods of the mid-Pliocene, resulting in a partitioning of temperature changes between changes in the prescribed CO₂, orography, vegetation and ice sheet boundary conditions. We also carry out an analysis of the pre-industrial and mid-Pliocene results using an energy balance method described by Heinemann et al. (2009).

2. Experimental design

2.1. Model description — HadCM3

All the General Circulation Model (GCM) simulations described in this paper are carried out using the UK Met Office coupled atmosphere GCM HadCM3, version 4.5 (Gordon et al., 2000). The resolution of the atmospheric and land components is 3.75° in longitude by 2.5° in latitude, with 19 vertical levels in the atmosphere. The resolution of the ocean model is 1.25° by 1.25° with 20 levels in the vertical. Parameterisations include the radiation scheme of Edwards and Slingo (1996), the convection scheme of Gregory et al. (1997), and the MOSES-1 land-surface scheme, whose representation of evaporation includes the dependence of stomatal resistance on temperature, vapour pressure and CO₂ concentration (Cox et al., 1999). The ocean model uses the Gent and McWilliams (1990) mixing scheme. There is no explicit horizontal tracer diffusion in the model. The horizontal resolution allows the use of a smaller coefficient of horizontal momentum viscosity leading to an improved simulation of ocean velocities compared to earlier versions of the model. The sea ice model uses a simple thermodynamic scheme and contains parameterisations of ice concentration (Hibler, 1979) and ice drift and leads (Cattle and Crossley, 1995). In simulations of the present-day climate, the model has been shown to simulate SST in good agreement with modern observations, without the need for flux corrections (Gregory and Mitchell, 1997). Future climate predictions from the model were presented in the latest IPCC report (Solomon et al., 2007), and it has been used in the Paleoclimate Modelling Intercomparison Project to simulate Last Glacial Maximum and Mid-Holocene climates (Braconnot et al., 2007). The model will also be used in the forthcoming PlioMIP project (Haywood et al., 2010, 2011b).

2.2. Boundary conditions

The PRISM project (http://geology.er.usgs.gov/eespteam/prism/) has as its main aim the characterisation of the palaeoenvironment of the mid-Pliocene warm period (3.26–3.025 Ma) on a global scale. In this paper, we simulate the mid-Pliocene climate by making use of the PRISM2 reconstruction of orography, vegetation, and ice sheet extent (Dowsett, 2007; Dowsett et al., 1999), which are described below.

The PRISM2 orography reconstruction was based on palaeobotanical evidence suggesting that the East African rift area were 500 m higher during the mid-Pliocene relative to today (Thompson and Fleming, 1996). In contrast, palaeoelavation of the Western Cordillera of North America and northern South America was reduced by 50%. Large elevation differences are noted in both Greenland and Antarctica due to significant removal of continental ice (Dowsett, 2007; Dowsett et al., 1994). PRISM2 land ice distribution and volume was closely associated with sea level estimates from several sources (see Dowsett, 2007), which indicate a eustatic sea level rise of around 25 m compared to modern. These estimates have recently been confirmed by independent studies based on the depth palaeoecology of foraminiferal assemblages from New Zealand (Naish and Wilson, 2009) and benthic Mg/Ca and oxygen isotopes (Dwyer and Chandler, 2009). Antarctic ice distribution was based upon a modelled stable ice sheet configuration (see Dowsett et al., 1999), strongly constrained by the sea-level reconstructions. The PRISM2 vegetation reconstruction (Dowsett et al., 1999) was compiled from fossil pollen and plant macrofossil data from 74 sites covering all continents. PRISM2 vegetation is identical to PRISM1 (see Thompson and Fleming, 1996). PRISM2 uses seven land cover categories (desert, tundra, grassland, deciduous forest, coniferous forest, rainforest, and land ice) that are a simplification of the 22 land cover types of Matthews (1985). From the PRISM2 vegetation, orography, and ice-sheet extent, we derive all the boundary conditions necessary to run the GCM in mid-Pliocene mode (a total of 23 variables different to those of the pre-industrial, such as heat capacity of the soil, albedo, moisture holding capacity etc.).

Since the development of the PRISM2 dataset, the USGS have now released an updated version — PRISM3 (Dowsett et al., 2010a, 2010b). We use the PRISM2 dataset; firstly, to maintain consistency with previous modelling studies, in particular H&V04 and Lunt et al. (2010a); secondly, the mid-Pliocene simulation with PRISM2 boundary conditions has been spun up for a total of over 1000 yrs, which is considerably more than could be achieved with new boundary conditions in a reasonable timeframe. In Section 4.1 we discuss the implications for this study of using PRISM2 compared to PRISM3 boundary conditions.

2.3. Factorisation methodology

The primary aim of this study is to assess the relative importance of various boundary condition changes which contribute to mid-Pliocene warmth. Therefore, we are aiming to partition the total mid-Pliocene warming, ΔT, into four components, each due to the change in one of the boundary conditions CO₂, orography, ice sheet, and vegetation. The assumption here is that other palaeogeographic changes not currently captured by the PRISM dataset, such as soils or lakes, have a negligible impact on the global mean temperature change.

\[ \Delta T = \Delta T_{CO_2} + \Delta T_{orog} + \Delta T_{ice} + \Delta T_{veg} \]  

(1)

‘Factor separation’ techniques (e.g. Stein and Alpert, 1993) can be used to determine these components of the mid-Pliocene surface air temperature change \( \Delta T_{CO_2} \), \( \Delta T_{orog} \), \( \Delta T_{ice} \), and \( \Delta T_{veg} \). Typically, this involves carrying out an ensemble of GCM simulations with various
combinations of boundary conditions. Here we present a new factorisation methodology, which we believe improves on previous work.

We name a GCM simulation which has boundary conditions x and y modified from pre-industrial to mid-Pliocene as \( E_{xy} \). The four boundary conditions considered are atmospheric \( CO_2 \) (c), orography (o), vegetation (v), and ice sheets (l). Thus, a pre-industrial simulation is \( E \), a mid-Pliocene simulation is \( E_{ociv} \), and e.g. a simulation with pre-industrial ice sheets and vegetation but mid-Pliocene orography and \( CO_2 \) is \( E_{oc} \). The corresponding surface air temperature distributions in these simulations we name \( T \), \( T_{ociv} \), and \( T_{oc} \) respectively.

For simplicity, we first describe our factorisation methodology by considering a simpler example, where only two boundary conditions \((CO_2 \) and orography\) are changed instead of four. The simplest factor separation technique is the incremental application of the boundary conditions. For our simplified example, this could involve an ensemble of 3 GCM simulations: \( E \), \( E_{o} \), and \( E_{oc} \). This has the advantage that a limited number of simulations \((N + 1, \text{where } N \text{ is the number of processes investigated}) \) need be carried out. It has the disadvantage that it results in a non-unique solution: one could equally define

\[
\begin{align*}
\delta T_{CO_2} & = T_{oc} - T_o, \\
\delta T_{orog} & = T_{oc} - T_c,
\end{align*}
\]

which, due to non-linearity, would in general result in a different partitioning.

Stein and Alpert (1993) (henceforth S&A93) recognised this and instead suggested that, considering the temperature response as a continuous function of two variables (in our simplified example orography and \( CO_2 \)), and carrying out a Taylor expansion about the control climate, one can write

\[
\Delta T = \frac{\partial T}{\partial CO_2} \Delta CO_2 + \frac{\partial T}{\partial orog} \Delta orog + \text{nonlinear terms}. \tag{4}
\]

They suggested that the nonlinear terms could be considered as ‘synergy’, \( S \), between the two forcing variables, and that the partial derivatives be estimated from the GCM simulations relative to the control, so that

\[
\begin{align*}
\delta T_{CO_2} & = T_{oc} - T_o, \\
\delta T_{orog} & = T_{oc} - T_c, \\
S & = T_{oc} - T_o - T_c + T.
\end{align*}
\]

This method, illustrated in Fig. 1a, has been used extensively in the climate literature (e.g., for the LGM see Broccoli and Manabe, 1987; von Deimling et al., 2006). It has the advantage that a limited number of simulations \((N + 1, \text{where } N \text{ is the number of processes investigated}) \) need be carried out. It has the disadvantage that it results in a non-unique solution: one could equally define

\[
\begin{align*}
\delta T_{CO_2} & = T_{oc} - T_o, \\
\delta T_{orog} & = T_{oc} - T_c, \\
S & = T_{oc} - T_o - T_c + T.
\end{align*}
\]

This is equivalent to averaging the two different formulations of the S&A93 approach in Eqs. (5) and (6). An alternative, but identical, interpretation is that our technique uses the S&A93 formulation of Eq. (5) but attributes the synergy term, \( S \), equally between the two forcings:

\[
\begin{align*}
\delta T_{CO_2} & = T_{oc} - T_o + S/2, \\
\delta T_{orog} & = T_{oc} - T_c + S/2, \\
S & = T_{oc} - T_o - T_c + T. \tag{8}
\end{align*}
\]

It is also equivalent to averaging the two linear formulations in Eqs. (2) and (3).

Our formulation has the advantage that it takes into account nonlinear interactions, and is symmetric. In common with the S&A93 approach, it requires \( 2^N \) GCM simulations, and so is more computationally demanding than the linear approach.

For our mid-Pliocene study, where we actually have 4 variables (\( CO_2 \), orography, vegetation, and ice sheets), this would require \( 2^4 = 16 \) simulations. The factorisation would be as follows:

\[
\begin{align*}
\delta T_{CO_2} & = \frac{1}{8} \left( (T_{oc} - T_o) + (T_{oc} - T_o) + (T_{oc} - T_c) + (T_{oc} - T_c) + (T_{oc} - T_o) + (T_{oc} - T_o) + (T_{oc} - T_c) + (T_{oc} - T_c) \right), \\
\delta T_{orog} & = \frac{1}{8} \left( (T_{oc} - T_o) + (T_{oc} - T_c) + (T_{oc} - T_o) + (T_{oc} - T_c) + (T_{oc} - T_o) + (T_{oc} - T_c) + (T_{oc} - T_o) + (T_{oc} - T_c) \right). \tag{9}
\end{align*}
\]
this case CO2, orography, vegetation, and ice (Eq. 13).

\[ \Delta T_{\text{veg}} = \frac{1}{2}(T_{\text{veg}} - T_{\text{ocv}}), \]

\[ \Delta T_{\text{ice}} = \frac{1}{2}(T_{\text{ice}} - T_{\text{ocv}}). \]

2.4. Mid-Pliocene model-data comparison

Before presenting and discussing our results, it is first important to have some confidence that the mid-Pliocene simulation, \( E_{\text{ocv}} \), is consistent with observations of that period.

The SSTs in our mid-Pliocene simulation were evaluated relative to reconstructions of mid-Pliocene SST in Lunt et al. (2010a). They showed that the global mean SST change, mid-Pliocene minus pre-industrial, was well simulated (1.83 °C in the model and 1.67 °C in the observations). However, they also found that the latitudinal distribution of temperature change was not well simulated (their Fig. 3c): the modelled mid-Pliocene warming being too great in the tropics and too small towards the poles. These discrepancies were investigated and discussed further in Dowsett et al. (2011).

A model-data comparison for the terrestrial climate, using a database of Pliocene palaeobotanical data (Salzmann et al., 2008, 2009) was presented in the Supplementary Information of Lunt et al. (2010a) They found a fair agreement between \( E_{\text{ocv}} \) and the data on a global scale, with significantly improved skill at high latitudes in the \( E_{\text{ocv}} \) simulation compared with the pre-industrial E simulation.

3. Results

The temperature changes due to the CO2 (\( \Delta T_{\text{CO2}} \)), orography (\( \Delta T_{\text{orog}} \)), vegetation (\( \Delta T_{\text{veg}} \)) and ice sheet (\( \Delta T_{\text{ice}} \)) boundary condition changes, as calculated from Eq. (13), as well as the total change, \( \Delta T \), are illustrated in Fig. 4. As a global average, of the total mid-Pliocene 3.3 °C temperature change, 1.6 °C (48%) is from the CO2 (\( \Delta T_{\text{CO2}} \)), 0.7 °C (21%) is from the orography (\( \Delta T_{\text{orog}} \)), 0.7 °C (21%) is from the vegetation (\( \Delta T_{\text{veg}} \)), and 0.3 °C (10%) is from the ice sheets (\( \Delta T_{\text{ice}} \)).

\( \Delta T_{\text{CO2}} \) (Fig. 4b) represents the temperature change due to CO2 alone. It shares much in common with similar (CO2 doubling as opposed to 280–400 ppmv here) results presented in the most recent report of the IPCC (Solomon et al., 2007). For example, there is polar amplification due to snow and sea ice feedbacks, and greater temperature change on land compared to ocean due to reduced latent cooling and lower heat capacity. The North Atlantic shows reduced temperature increase due to ocean mixing and reduced northward heat transport in the Atlantic due to an increase in the intensity of the hydrological cycle. The increase of 1.6 °C implies a climate sensitivity due to a doubling of CO2 of ~3.2 °C, which is close to the middle of the IPCC range (Solomon et al., 2007). \( \Delta T_{\text{orog}} \) (Fig. 4c) highlights the local lapse-rate warming effect of the lower mid-Pliocene Rocky Mountain range. There is also a cooling to the west of the mid-Pliocene Canadian Rockies, associated with reduced precipitation and cloud cover, due to reduced ascent over the mountain range. There is a significant non-local effect of the lower Rockies — there is a large Arctic warming, in particular in the Barents Sea, which is amplified by reduced sea ice cover. This is due to a modification of the Rossby wave pattern, which is more zonally symmetric with the lower Rockies, indicated by a reduced trough over Greenland in the 500 mbar geopotential height field, consistent with previous work (e.g. Foster et al., 2010; Kutzbach et al., 1989). Very localised cooling associated with topographic effects are seen in the Andes, Himalayas, and East African rift valley regions. The surface ocean warming east of Japan is consistent with previous work showing this to be a region sensitive to orographic change in this model (Lunt et al., 2010b).
Fig. 3. Orography and snow-free albedo for the $E$, $E_{oci}$, $E_{ocv}$, and $E_{ociv}$ GCM simulations. For equivalent figures of the other GCM simulations ($E_{ne}$, $E_n$, and $E_{nv}$), see Table 1 of Supplementary Information of Lunt et al. (2010a).
(b) shows that the largest vegetation-related temperature changes are in the Canadian Arctic, in particular Greenland (change from ice sheet to boreal forest), the Canadian archipelago (change from bare soil and glaciers to boreal forest), and Siberia (change from bare soil to boreal forest). This warming can be attributed to the relatively low albedo of boreal forest in the model, even when there is snow-cover on the ground. There are also large changes in the tropics, in particular in the Arabian peninsula, where the transition from desert to grassland vegetation (based on pollen data (Van Campo, 1991)) resulted in a lower albedo in the mid-Pliocene than in the modern (see Fig. 3). Some of the temperature changes attributed to vegetation will also be due to modifications to the roughness length, potential evapotranspiration, and other vegetation-specific model parameters.

\( dT_{\text{veg}} \) (Fig. 4c) shows the largest vegetation-related temperature changes are in the Canadian Arctic, in particular Greenland (change from ice sheet to boreal forest), the Canadian archipelago (change from bare soil and glaciers to boreal forest), and Siberia (change from bare soil to boreal forest). This warming can be attributed to the relatively low albedo of boreal forest in the model, even when there is snow-cover on the ground. There are also large changes in the tropics, in particular in the Arabian peninsula, where the transition from desert to grassland vegetation (based on pollen data (Van Campo, 1991)) resulted in a lower albedo in the mid-Pliocene than in the modern (see Fig. 3). Some of the temperature changes attributed to vegetation will also be due to modifications to the roughness length, potential evapotranspiration, and other vegetation-specific model parameters. \( dT_{\text{ice}} \) (Fig. 4d) shows warming in Greenland and parts of Antarctica due to a combination of lapse-rate, due to a lower mid-Pliocene ice
It is clear that in the Northern Hemisphere, the polar amplification around 60°N is much more muted. These features are consistent with sea-ice and snow being the main causes of the seasonality.

It is interesting to assess the linearity of the climate system to these changes in boundary conditions. For example, to what extent does the temperature response of the system to a CO₂ change depend on the climate base state. Or, in other terms, how large is the ‘synergy’ term (S in Eq. 5) in the S&K93 formulation? Fig. 6 shows the two terms (TC−T and Tco−Tc) which make up dTco₂ in Eq. (8), and the difference between them (S). The non-linearity is small compared to the temperature change itself, showing that in this case, the temperature response to an increase in CO₂ is largely independent of the orographic configuration. Similarly, the vegetation and ice sheet changes exhibit relatively small non-linearity (not shown). This implies that in this case, similar results could be obtained with a simple linear factorisation. However, it is not possible to know this a priori. The subtle non-linearities of the response of the system to changes in CO₂ alone are discussed in more detail in Haywood et al. (2011a).

It is also instructive to compare our results with those of H&V04. Our mid-Pliocene simulation differs from that of H&V04 for two reasons. Firstly, our simulation is a continuation of that of H&V04, and so is further spun-up and closer to equilibrium. Secondly, our simulation has been carried out over a number of ‘real-world’ years, and over this time has been migrated across several computers and Fortran compilers. Both hardware and compiler changes can affect the mean equilibrium climate of a model, due at least in part to non-standard programming practice, for example multiple ‘data’ statements in Fortran subroutines (Steenman-Clark, 2009). Fig. 7a shows the difference in mid-Pliocene surface air temperature between our simulation and that of H&V04, and Fig. 7b shows the difference in mid-Pliocene surface air temperature anomaly, mid-Pliocene minus pre-industrial, between our simulation and that of H&V04. Our mid-Pliocene simulation is significantly cooler than that of H&V04 (~0.8 °C in the global annual mean), but the difference in anomalies is smaller (0.3 °C). Examination of the temporal evolution of these differences indicates that the effect of hardware and compiler change is more important than the effect of increased spinup time. This underlines the importance of always carrying out sensitivity simulations on the same machine, and with the same compiler, as any control simulation.

As stated in the Introduction, H&V04 estimated the contributions to mid-Pliocene warmth by considering aspects of the global energy balance. Heinemann et al. (2009), in the context of the Eocene, present a different method of energy-balance analysis which includes a meridional analysis. Here, we use the method of Heinemann et al. (2009) to analyse our mid-Pliocene (E_{oni}) and pre-industrial (E) simulations. The method gives latitudinal distributions of the contribution to the surface temperature change, E_{oni}−E, of: (a) emissivity changes due to changes in greenhouse gases, (b) emissivity changes due to changes in clouds, (c) albedo changes due to changes in the planetary surface, (d) albedo changes due to changes in clouds, and (e) heat transport changes. This latitudinal partitioning is shown in Fig. 5(b). The first thing to note is that this approach is based on zonal and seasonal means, and as such the total surface temperature change is slightly underestimated by the energy balance approach (compare the green line with the black line in Fig. 5(b)). On a global scale, the contribution of heat transport to the total change is by definition zero, but in the Northern Hemisphere there is a small positive contribution at high latitudes and a small negative contribution at low latitudes, consistent with a slight increase in poleward heat transport. The global mean contribution of clouds (both albedo and emissivity effects) is relatively small, but in the short-wave this results from a cancellation of a positive contribution in the tropics and a negative contribution at mid-high latitudes. Changes in emissivity (due to the increase in greenhouse gas from 280 to 400 ppmv, and the associated water vapour forcing) contribute 61% of the total surface temperature change, with greatest contribution in
mid-high latitudes. Surface albedo changes contribute 44%, due almost entirely to mid-high latitude changes; in the tropics the change in albedo contributes very little. Overall it can be seen that surface albedo and direct greenhouse-gas forcing are the greatest contributors to the total change, with the greenhouse gas forcing dominating in low latitudes, and the surface albedo changes dominating at mid-high latitudes. The polar amplification is significantly dampened by changes in shortwave cloud forcing. It should be noted that cloud processes are amongst the most uncertain in GCMs, and so these results are likely to be model dependent.

4. Discussion

Here we discuss some of the assumptions in this work, including quantitative estimates of how some of these assumptions could affect our results.

4.1. Palaeoenvironmental boundary conditions

In Section 2.2 we describe why we use the PRISM2 boundary conditions as opposed to the PRISM3 boundary conditions. The most significant effect of this is likely related to the different orography dataset in PRISM3 compared to PRISM2 (the ice sheets, although different, are similar in extent and height, and the PRISM3 vegetation is based on an extended dataset which includes PRISM2 as a subset). PRISM3 orography is based on the reconstruction of Markwick (2007). It differs from PRISM2 mainly in the high Eurasian latitudes and the Himalayas where the geological evidence is inconclusive and debated (e.g. Rowley and Garzione, 2007; Spicer et al., 2005). The Markwick (2007) reconstruction is actually much closer to modern than that of PRISM2. Therefore, using modern orography instead of PRISM2 provides an end-member approximation for the uncertainty in our results. In this case, given the linearity of the system highlighted in Section 4, the total mid-Pliocene temperature change can be approximated by:

$$\Delta T_{\text{model}} = \Delta T - dT_{\text{org}} = dT_{\text{CO}_2} + dT_{\text{veg}} + dT_{\text{wsc}}$$

which is 2.6 °C. Then, the partitioning (Table 1) is 1.6 °C (61%) from the CO2 ($dT_{\text{CO}_2}$), 0.7 °C (27%) is from the vegetation ($dT_{\text{veg}}$), and 0.3 °C (13%) from the ice sheets ($dT_{\text{wsc}}$).

There is no information given in either PRISM2 or PRISM3 on possible bathymetric differences between the mid-Pliocene and present. As such, we use modern bathymetry in the simulations presented here. However, geophysical records of mantle temperature beneath the North Atlantic indicate that the Greenland-Scotland ridge was about 300 m lower in the Pliocene than modern (Robinson et al., 2011). A recent modelling study (Robinson et al., 2011) has shown that, although this has negligible effect on the global mean temperature, it could lead to increased polar warmth (greater than 5 °C) in the mid-Pliocene due to increased oceanic northward heat transport in the North Atlantic. This has the effect of bringing the modelled SST to estimates of global surface air temperature using a scaling factor, Lunt et al. (2010a) estimated the true $\Delta T_{\text{CO}_2} = x$ to be about 0.27 °C greater than the $\Delta T$ predicted by the model. This allows us to estimate x, the ‘true’ value of mid-Pliocene CO2, to be 380 ppmv. It should be noted that this calculation assumes that our uncertainty in CO2 is much greater than uncertainties which arise due to model error, errors in the applied mid-Pliocene boundary conditions, and errors in the PRISM3 SSTs.

4.2. Mid-Pliocene CO2

Mid-Pliocene atmospheric CO2 has been reconstructed by a variety of proxies. A value of 400 ppmv has been used in this and several other previous modelling studies of the mid-Pliocene climate (including H&V04), but there are uncertainties in this figure. For example, based on measurements of $d^13C$ in ocean sediments, Raymo et al. (1996) cite a mean value of 380 ppmv with maxima as high as 425 ppmv. More recent data from Seki et al. (2010), using alkenones and boron isotope proxies, cite a mean of 360 ppmv with uncertainties $\pm$ 30 ppmv. Other recent data (Pagani et al., 2010) supports a mean of 380 ppmv. As such, for consistency with previous work, and to account for likely associated increases in non-CO2 greenhouse gases such as are observed in the ice core record (Siegenthaler et al., 2005), we consider here the effects of 350 and 450 ppmv as alternative CO2 concentrations. To first order, the temperature effects of elevated CO2 are expected to scale logarithmically with the CO2 concentration. Therefore, it is possible to estimate the total mid-Pliocene temperature change for an arbitrary CO2 level of $x$, $\Delta T_{\text{CO}_2} = x$ as:

$$\Delta T_{\text{CO}_2} = x = x - \Delta T_{\text{org}} = dT_{\text{CO}_2} + dT_{\text{veg}} + dT_{\text{wsc}}$$

For a CO2 level of 350 ppmv this gives $\Delta T_{\text{CO}_2} = 350 = 2.7$ °C, and a partitioning (see Table 1) of 1.0 °C (36%) from the CO2 ($dT_{\text{CO}_2}$), 0.7 °C (26%) from the orography, 0.7 °C (26%) from the vegetation ($dT_{\text{veg}}$), and 0.3 °C (13%) from the ice sheets ($dT_{\text{wsc}}$).

Furthermore, given a ‘true’ mid-Pliocene global mean temperature change, $\Delta T_{\text{CO}_2} = x$, we can solve Eq. (15) for $x$. By converting the PRISM3 estimates of global SST to estimates of global surface air temperature using a scaling factor, Lunt et al. (2010a) estimated the true $\Delta T_{\text{CO}_2} = x$ to be about 0.27 °C greater than the $\Delta T$ predicted by the model. This allows us to estimate x, the ‘true’ value of mid-Pliocene CO2, to be 380 ppmv. It should be noted that this calculation assumes that our uncertainty in CO2 is much greater than uncertainties which arise due to model error, errors in the applied mid-Pliocene boundary conditions, and errors in the PRISM3 SSTs.
4.3. Climate variability through the mid-Pliocene

The mid-Pliocene spans approximately 300,000 yrs, and, although relatively stable compared to the Quaternary, does display climate variability on orbital timescales (Lisiecki and Raymo, 2005), which can be interpreted as a series of glacialis and interglacials (albeit much smaller in magnitude than those of the Quaternary). By combining high resolution mid-Pliocene oxygen isotope and Mg/Ca measurements, Dwyer and Chandler (2009) identified six sea level highstands during the mid-Pliocene of between 10 and 30 m above modern, and several lowstands, including Marine Isotope Stage KM2 in the middle of the mid-Pliocene, estimated to be 40 m below modern. However, we carry out a single simulation to represent this entire time period.

For the orography, this is probably not an issue, as changes in orography occur on much longer timescales than orbital fluctuations. However, the orbit, CO2, ice sheets, and vegetation likely varied significantly through the mid-Pliocene. The orbital forcing in our simulations is that of modern. At 65°N in June, the modern forcing is close to the average forcing of the mid-Pliocene, the difference being 15 W m−2 compared to a maximum difference of +50 W m−2 during the mid-Pliocene (Lunt et al., 2008). For CO2, we have used 400 ppmv whereas the record of Raymo et al. (1996) varies between 330 and 425 ppmv. For ice sheets, the PRISM2 reconstruction is characterised by a sea-level increase of 25 m compared to modern, whereas Dwyer and Chandler (2009) find variations in global sea level of ±25 m compared to modern, encompassing glacial/interglacial variability. The PRISM3 SST evaluation dataset does consist of sub-orbitally dated sites. However, the PRISM SSTs do not represent average SSTs through the mid-Pliocene but have been filtered via a process of ‘warm peak averaging’ (Dowsett et al., 2009), which means that the PRISM3 SSTs represent average warm interglacial conditions in the mid-Pliocene. For vegetation, the data sites in the Thomson and Fleming reconstruction, upon which PRISM2 are based, are not dated to orbital timescale accuracy, and so each site could represent either glacial or interglacial-type conditions. The same is true of the Salzmann et al. (2008) vegetation dataset, with which our simulation has been evaluated. However, in locations where a number of possible biomisations were consistent with the data, Salzmann et al. (2008) chose the warmest, to maintain consistency with the SST warm peak averaging.

As such, our simulations are a hybrid representation of the mid-Pliocene: the orbit, vegetation and orography being close to mid-Pliocene average, and the CO2 and ice sheets being closer to interglacial values. The mid-Pliocene simulation has previously been compared with vegetation data which represent an average-to-warm mid-Pliocene palaeoenvironment (Lunt et al., 2010a), and SST data which represent interglacial values (Dowsett et al., 2011). These discrepancies may go some way to explaining some of the model-data disagreements. For example, the greater high-latitude warmth in the PRISM SST reconstruction compared to the model could be a result of the warm-peak averaging, which by definition biases the SST reconstructions to warm values. Future work will aim to carry out simulations more representative of specific time periods within the mid-Pliocene, and to compare these to orbitally-resolved versions of the PRISM SST dataset.

4.4. Model uncertainties

Uncertainties associated with the model itself (as opposed to the boundary condition uncertainties discussed above) can be broadly divided into ‘parametric uncertainty’ and ‘structural uncertainty’.

Parametric uncertainty relates to uncertainties in model parameters. These parameters are often associated with the representation of sub-gridscale processes and include, for example, the gridbox-average relative humidity at which clouds are assumed to start forming. They are generally poorly constrained by observations and so are essentially ‘tunable’. A single model simulation, as presented in this paper, can only represent one single point in the whole space of possible plausible parameter combinations, and as such undersamples...
the range of model possibilities. The full space can be explored by carrying out simulations in which these tunable parameters are perturbed. A preliminary study has been carried out with this model in the context of the mid-Pliocene (Pope et al., 2011). That study found a range of $\Delta T$ of 2.7 °C to 4.5 °C and could therefore be used to place approximate error bars on our $\Delta T$; however, it did not investigate the causes of such a change, so the impact of uncertain parameters on our factorisation is unclear, and is a focus of ongoing work.

Structural uncertainty relates to changes in the model which can not be made purely by modifying the values of tunable parameters. It relates to our uncertainty in the physical processes themselves which govern Earth System behaviour, and our inability to implement complex processes in a numerical model of a given resolution. Some information on the magnitude of this error can be obtained by considering other climate models. Haywood et al. (2009) compared two structurally different models, of the mid-Pliocene. They had a range of $\Delta T$ of 2.39 °C to 2.41 °C (this is very much a minimum uncertainty range, especially as those simulations were carried out with atmosphere-only models). Again, whilst putting some context to our results, it is not clear how this uncertainty would affect our factorisation or energy balance analysis. Ongoing work, in the framework of the project PlioMIP, is aiming to gain more information on the structural uncertainty by comparing many atmosphere-only and atmosphere–ocean Pliocene simulations produced by different models (Haywood et al., 2010, 2011b).

5. Conclusions

Using a novel form of factorisation, we have partitioned the causes of mid-Pliocene warmth between CO$_2$ (36%–61%), orography (0–26%), vegetation (21%–27%) and ice sheets (9–13%). The ranges are estimated by considering the sensitivity of the results to uncertainties in the mid-Pliocene CO$_2$ concentration and orography (summarised in Table 1). Despite the relatively small contribution of ice sheets on a global scale, it is responsible for the majority of Southern Hemisphere high latitude warming. Northern Hemisphere high-latitude warming is due mainly to a combination of CO$_2$ and orography warming. Hemisphere high latitude warming. Northern Hemisphere high-latitude warming is due mainly to a combination of CO$_2$ and orography warming. Northern Hemisphere high-latitude warming.

Future work should further assess the sensitivity of these results to the boundary conditions applied (for example by using the newer PRISM3 reconstructions compared with PRISM2 used here, and extending the datasets to include varying soil properties), to the model used, and to parameters within the models themselves. Both the modelling and data communities should start to investigate orbital-scale variability within the mid-Pliocene. This is particularly important for assessing the real relevance of the mid-Pliocene as an analogue for long-term future (sub-orbital timescale) climate change.

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Appendix A. Supplementary data

Supplementary data to this article can be found online at doi:10.1016/j.epsl.2011.12.042.

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