This is a repository copy of *On the causes of mid-Pliocene warmth and polar amplification*.

White Rose Research Online URL for this paper:
http://eprints.whiterose.ac.uk/88182/

Version: Accepted Version

**Article:**


© 2012, Elsevier. Licensed under the Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International
http://creativecommons.org/licenses/by-nc-nd/4.0/

---

**Reuse**
Unless indicated otherwise, fulltext items are protected by copyright with all rights reserved. The copyright exception in section 29 of the Copyright, Designs and Patents Act 1988 allows the making of a single copy solely for the purpose of non-commercial research or private study within the limits of fair dealing. The publisher or other rights-holder may allow further reproduction and re-use of this version - refer to the White Rose Research Online record for this item. Where records identify the publisher as the copyright holder, users can verify any specific terms of use on the publisher's website.

**Takedown**
If you consider content in White Rose Research Online to be in breach of UK law, please notify us by emailing eprints@whiterose.ac.uk including the URL of the record and the reason for the withdrawal request.
On the causes of mid-Pliocene warmth and polar amplification

Daniel J. Lunt(1,*), Alan M. Haywood(2), Gavin A. Schmidt(3), Ulrich Salzmann(4), Paul J. Valdes(1), Harry J. Dowsett(5), Claire A. Loptson(1)

(1) School of Geographical Sciences
University of Bristol
University Road
Bristol BS8 1SS, UK

(2) School of Earth and Environment
Woodhouse Lane
University of Leeds
Leeds LS2 9JT, UK

(3) NASA/Goddard Institute for Space Studies
2880 Broadway
New York
NY 10025, USA

(4) School of the Built and Natural Environment
Northumbria University
Newcastle upon Tyne NE1 8ST, UK

(5) U.S. Geological Survey
12201 Sunrise Valley Drive MS926A
Reston
VA 20192, USA

(*) Corresponding author:
Email: d.j.lunt@bristol.ac.uk
Tel: +44 (0) 117 3317483
Fax: +44 (0) 117 9287878
1 Abstract

The mid-Pliocene (∼3 to 3.3 million years ago), is a period of sustained global warmth in comparison to the late Quaternary (0 to ∼1 million years 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.

2 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+ metres) meaning that global ice volume was reduced (e.g. Dowsett and Cronin, 1990; Naish and Wilson, 2009; Dwyer and Chandler, 2009). There were large fluctuations in ice cover on Greenland and West Antarctica, and during the interglacials they were probably largely free of ice (Lunt et al., 2008; Pollard and DeConto, 2009; Hill et al., 2010;
Dolan et al., 2011). 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 CO$_2$ 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 (Jansen et al., 2007; Haywood et al., 2000a), (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 (Fedorov et al., 2010; Bonham et al., 2009; 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).

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 Wm⁻²), surface albedo (2.3 Wm⁻²) and cloud cover (1.8 Wm⁻²) changes. They further partitioned the surface albedo component between land ice and snow (55%) and sea ice (45%) changes. From interrogating the ocean streamfunction 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 year 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).

3 Experimental Design

3.1 Model Description - HadCM3

All the General Circulation Model (GCM) simulations described in this paper are carried out using the UK Met Office coupled ocean-atmosphere GCM HadCM3, version 4.5 (Gordon et al., 2000).
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 Palaeoclimate 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).

3.2 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 et al., 1999; Dowsett, 2007), which are described below.

The PRISM2 orography reconstruction was based on palaeobotanical evidence suggesting that the
East African rift areas were 500 m higher during the mid-Pliocene relative to today (Thompson and Fleming, 1996). In contrast, palaeoelevation 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 et al., 1994; Dowsett, 2007). 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 foram 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., 2010b,a). 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 years, which is considerably more than could be achieved with new boundary conditions in a reasonable timeframe. In section 5.1 we discuss the implications for this study of using PRISM2 compared to PRISM3 boundary conditions.
3.3 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, $\Delta T$, into four components, each due to the change in one of the boundary conditions $CO_2$, 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.

$$\Delta T = dT_{CO_2} + dT_{orog} + dT_{ice} + dT_{veg}$$  \hspace{1cm} (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 $dT_{CO_2}$, $dT_{orog}$, $dT_{veg}$, and $dT_{ice}$. 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 ($i$). 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 sim-
plified example, this could involve an ensemble of 3 GCM simulations: $E$, $E_c$, and $E_{oc}$. The total
temperature anomaly, $\Delta T$ (equal to $T_{oc} - T$ in this simplified example), could be separated into 2
components:

$$dT_{CO_2} = T_c - T$$
$$dT_{orog} = T_{oc} - T_c,$$

(2)

This method, illustrated in Figure 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 ($N + 1$, where $N$ 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

$$dT_{CO_2} = T_{oc} - T_o$$
$$dT_{orog} = T_o - T,$$

(3)

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

Stein and Alpert (1993) (henceforth S&A93) recognised this and instead suggested that, consider-
ing 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

$$\Delta T = \frac{\partial T}{\partial CO_2} \Delta CO_2 + \frac{\partial T}{\partial orog} \Delta orog + \text{nonlinear terms}$$

(4)

They suggested that the nonlinear terms could be considered as ‘synergy’, $S$, between the two for-
ing variables, and that the partial derivatives be estimated from the GCM simulations relative to the
control, so that

$$dT_{CO_2} = T_c - T$$
This method, illustrated in Figure 1b, has been used in several previous studies (e.g. for the mid-Holocene and LGM see Wohlfahrt et al., 2004; Jahn et al., 2005). It has the advantage that it takes into account the non-linear interactions between the different boundary conditions. However, it requires a larger number of simulations ($2^N$) than the linear approach. Perhaps more importantly, it has the problem that it is not symmetric: one could equally carry out the Taylor expansion about the perturbed climate, and write

\[ -dT_{CO_2} = T_o - T_{oc} \]

\[ -dT_{orog} = T_c - T_{oc} \]

\[ -S = T - T_o - T_c + T_{oc} \]  

(6)

i.e. it would in general give a different answer if one asked “why is the mid-Pliocene warmer than pre-industrial” than if one asked “why is the pre-industrial cooler than the mid-Pliocene” (although the synergy term, $S$, would have the same magnitude in both cases).

In order to obtain a symmetric and unique factorisation, we instead estimate the partial derivatives in equation 4 with their average values over the domain considered, and write for our simplified case:

\[ dT_{CO_2} = \frac{1}{2}((T_c - T) + (T_{oc} - T_o)) \]

\[ dT_{orog} = \frac{1}{2}((T_o - T) + (T_{oc} - T_c)) \].  

(7)

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

$$dT_{CO_2} = T_c - T + S/2$$
$$dT_{orog} = T_o - T + S/2$$

$$S = T_{oc} - T_o - T_c + T$$  \hspace{1cm} (8)$$

It is also equivalent to averaging the two linear formulations in Equations 2 and 3.

Our formulation has the advantage that it takes into account non-linear 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:

$$dT_{CO_2} = \frac{1}{8}((T_c - T) + (T_{oc} - T_o) + (T_{ic} - T_i) + (T_{vc} - T_v) +$$
$$ (T_{ocv} - T_{ov}) + (T_{oci} - T_{oi}) + (T_{civ} - T_{iv}) + (T_{ociv} - T_{oiv})), \hspace{1cm} (9)$$

$$dT_{orog} = \frac{1}{8}((T_o - T) + (T_{co} - T_c) + (T_{io} - T_i) + (T_{vo} - T_v) +$$
$$ (T_{cov} - T_{cv}) + (T_{coi} - T_{ci}) + (T_{oiv} - T_{iv}) + (T_{coiv} - T_{civ})), \hspace{1cm} (10)$$

$$dT_{veg} = \frac{1}{8}((T_v - T) + (T_{cv} - T_c) + (T_{iv} - T_i) + (T_{vo} - T_o) +$$
$$ (T_{cvo} - T_{co}) + (T_{cvi} - T_{ci}) + (T_{vio} - T_{io}) + (T_{cvio} - T_{cio})), \hspace{1cm} (11)$$

$$dT_{ice} = \frac{1}{8}((T_i - T) + (T_{ci} - T_c) + (T_{vi} - T_v) + (T_{oi} - T_o) +$$
$$ (T_{cio} - T_{co}) + (T_{civ} - T_{ce}) + (T_{vie} - T_{vo}) + (T_{ciev} - T_{cvo})). \hspace{1cm} (12)$$

Given the computational expense of carrying out 16 fully-coupled GCM simulations, we choose instead to consider CO$_2$/orography, and vegetation/ice sheets separately, and carry out two $N = 2$
factor separations (as in Equation 13), requiring only 7 simulations (illustrated in Figure 2).

\[
\begin{align*}
    dT_{CO_2} &= \frac{1}{2}((T_c - T) + (T_{oc} - T_o)), \\
    dT_{orog} &= \frac{1}{2}((T_o - T) + (T_{oc} - T_c)), \\
    dT_{veg} &= \frac{1}{2}((T_{ocv} - T_{oc}) + (T_{ociv} - T_{oci})), \\
    dT_{ice} &= \frac{1}{2}((T_{oci} - T_{oc}) + (T_{ocio} - T_{ocv})).
\end{align*}
\] (13)

This factorisation is more computationally efficient than the full factorisation in Equation 12, but is not fully symmetric.

Five of these simulations (\(E, E_o, E_c, E_{oc}, E_{ociv}\)) were used in the study of Lunt et al. (2010a) in the context of deriving estimates of Earth system sensitivity, and the orography and snow-free surface albedo of these simulations are shown in their Table 1 of their Supplementary Information. The orography and snow-free albedo (an indicator of the land ice and vegetation distributions) for the 2 new simulations (\(E_{oci}, E_{ocv}\)), along with those for \(E\) and \(E_{ociv}\) for comparison, are shown in Figure 3.

It is worth noting that because the ice sheets and vegetation are mutually exclusive in any one model grid cell, it is not possible to uniquely define boundary conditions for simulations \(E_{oci}\) and \(E_{ocv}\). For the simulation with modern vegetation but Pliocene ice sheets (\(E_{oci}\)), in the regions which are ice sheet-free in the Pliocene but have ice sheets in the modern (e.g. the West Antarctic peninsula), it is not clear what albedo should be prescribed as there is no modern vegetation defined in these regions. Similarly, for the simulation with modern ice but Pliocene vegetation (\(E_{ocv}\)), in the same regions it is unclear whether to use the albedo of the Pliocene vegetation or of the modern ice. In other words, it is not well defined whether the albedo-induced warming associated with reduced ice sheets during the Pliocene is due to the reduction of ice *per se*, or due to the vegetation which replaces it. Here, we make the decision to attribute this warming to the vegetation that replaces it. As such, both simulations \(E_{oci}\) and \(E_{ocv}\) have the albedo of ice in regions which are ice-free in the Pliocene but have ice in the modern (Figure 3).
3.4 Mid-Pliocene model-data comparison

Before presenting and discussing our results, it is first important to have some confidence that the mid-Pliocene simulation, \( E_{\text{ociv}} \), 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\(^\circ\)C in the model and 1.67\(^\circ\)C in the observations). However, they also found that the latitudinal distribution of temperature change was not well simulated (their Figure 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 \( E_{\text{ociv}} \) and the data on a global scale, with significantly improved skill at high latitudes in the \( E_{\text{ociv}} \) simulation compared with the pre-industrial \( E \) simulation.

4 Results

The temperature changes due to the CO\(_2\) \( (dT_{CO_2}) \), orography \( (dT_{\text{orog}}) \), vegetation \( (dT_{\text{veg}}) \) and ice sheet \( (dT_{\text{ice}}) \) boundary condition changes, as calculated from equations 13, as well as the total change, \( \Delta T \), are illustrated in Figure 4. As a global average, of the total mid-Pliocene 3.3\(^\circ\)C temperature change, 1.6\(^\circ\)C (48\%) is from the CO\(_2\) \( (dT_{CO_2}) \), 0.7\(^\circ\)C (21\%) is from the orography \( (dT_{\text{orog}}) \), 0.7\(^\circ\)C (21\%) is from the vegetation \( (dT_{\text{veg}}) \), and 0.3\(^\circ\)C (10\%) is from the ice sheets \( (dT_{\text{ice}}) \).

\( dT_{CO_2} \) (Figure 4b) represents the temperature change due to CO\(_2\) alone. It shares much in common with similar (CO\(_2\) 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). $dT_{orog}$ (Figure 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. Kutzbach
et al., 1989; Foster et al., 2010). 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). $dT_{veg}$ (Figure 4c) 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 PRISM2 reconstruction indicates a shift from
desert to grassland vegetation (based on pollen data (Van Campo, 1991)), resulting in a lower albedo
in the mid-Pliocene than in the modern (see Figure 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. $dT_{ice}$ (Figure 4d) shows warming in Greenland and
parts of Antarctica due to a combination of lapse-rate, due to a lower mid-Pliocene ice sheet height,
and albedo, due to the less reflective mid-Pliocene surface. The regions of Antarctic cooling are due to the fact that the PRISM ice sheet is higher in the Pliocene than in the modern in these regions. This is consistent with increased precipitation in the interior of the East Antarctic ice sheet in the warmer climate, and with modelled predictions for the future evolution of the Antarctic ice sheet under greenhouse gas forcing (e.g. Huybrechts et al., 2004). The cooling in the Barents Sea is also consistent with previous work investigating the climatic effects of the removal of the Greenland ice sheet (Toniazzo et al., 2004; Lunt et al., 2004). However, apart from in this region, the signal due to the removal of the ice is very localised.

The results also allow us to ascertain the contribution to polar amplification of the four factors. We define polar amplification in this case to be any warming in the polar regions which is greater than the global mean warming. Figure 5(a) shows the same results as in Figure 4, but as zonal means. It is clear that the polar amplification in the Southern Hemisphere is due primarily to the ice sheet changes, whereas in the Northern Hemisphere it is due primarily to a combination of CO$_2$ and orography changes, with some contribution from vegetation around 60-70°N. Figures 1-3 in Supplementary Information illustrate the seasonality of the factorisation and polar amplification. It is clear that in the Northern Hemisphere, the polar amplification is dominated by an autumn and winter signal; in JJA there is almost no Northern Hemisphere polar amplification. In the Southern Hemisphere the seasonality 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 CO$_2$ change depend on the climate base state. Or, in other terms, how large is the ‘synergy’ term ($S$ in Equation 5) in the S&A93 formulation? Figure 6 shows the two terms ($T_c - T$ and $T_{oc} - T_o$) which make up $dT_{CO_2}$ in Equation 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 CO$_2$ 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 CO$_2$ 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). Figure 7a shows the difference in mid-Pliocene surface air temperature between our simulation and that of H&V04, and Figure 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 ($E_{ociv}$) and pre-industrial ($E$) simulations. The method gives latitudinal distributions of the contribution to the surface temperature change, $E_{ociv} - E$, 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 Figure 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 Figure 5(b)). On a global scale, the
contribution of heat transports to the total change is by definition zero, but in the Northern Hemi-
sphere 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 lat-
itudes. Changes in emissivity (due to the increase in greenhouse gas from 280 to 400ppmv, and the
associated water vapour forcing) contributes 61% of the total surface temperature change, with great-
est 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
short-wave 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.

5 Discussion

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

In section 3.2 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., 2003). 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:

\[ \Delta T_{\text{no orogy}} = \Delta T - dT_{\text{orogy}} = dT_{CO_2} + dT_{\text{veg}} + dT_{\text{ice}} \]  

which is 2.6 °C. Then, the partitioning (Table 1) is 1.6°C (61%) from the CO₂ \( dT_{CO_2} \), 0.7°C (27%) is from the vegetation \( dT_{\text{veg}} \), and 0.3°C (13%) from the ice sheets \( dT_{\text{ice}} \).

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 SSTs in the mid-Pliocene \( E_{\text{ociv}} \) simulation into better agreement with the PRISM3 proxy estimates in this region.
5.2 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 δ¹³C 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 + - 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 \( x \), \( \Delta T^{CO_2=x} \) as:

\[
\Delta T^{CO_2=x} = dT_{orog} + dT_{veg} + dT_{ice} + \frac{\log(x/280)}{\log(400/280)} dT_{co2}
\]  

(15)

For a CO₂ level of 350 ppm this gives \( \Delta T^{CO_2=350} = 2.7 \) °C, and a partitioning (see Table 1) of 1.0°C (36%) from the CO₂ \( (dT_{CO_2}) \), 0.7°C (26%) from the orography, 0.7°C (26%) from the vegetation \( (dT_{veg}) \), and 0.3°C (12%) from the ice sheets \( (dT_{ice}) \). For a CO₂ level of 450 ppm this gives \( \Delta T^{CO_2=450} = 3.8 \) °C, and a partitioning (see Table 1) of 2.1°C (55%) from the CO₂ \( (dT_{CO_2}) \), 0.7°C (18%) from the orography, 0.7°C (18%), from the vegetation \( (dT_{veg}) \), and 0.3°C (9%) from the ice sheets \( (dT_{ice}) \).

Furthermore, given a ‘true’ mid-Pliocene global mean temperature change, \( \Delta T^{CO_2=x} \), we can solve Equation 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 \( \Delta T^{CO_2=x} \) 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 CO$_2$, to be 380 ppmv. It should be noted that this calculation assumes that our uncertainty in CO$_2$ 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.

5.3 Climate variability through the mid-Pliocene

The mid-Pliocene spans approximately 300,000 years, 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 glacial and interglacial phases (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 over much longer timescales than orbital fluctuations. However, the orbit, CO$_2$, 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 Wm$^{-2}$ compared to a maximum difference of +50 Wm$^{-2}$ during the mid-Pliocene (Lunt et al., 2008). For CO$_2$, we have used 400ppmv 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 25m compared to modern, whereas Dwyer and Chandler (2009) find variations in global sea-level of ±25m 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.

5.4 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 behavior, 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, while 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).

6 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 changes. Furthermore, we have carried out an energy balance analysis, and shown that surface albedo changes and direct greenhouse-gas forcing contribute significantly more than cloud feedbacks to the total mid-Pliocene warming, with the greenhouse gas forcing dominating in low latitudes, and the surface albedo changes dominating at mid-high latitudes.

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.

References


temperature reconstruction. *Climate of the Past, 5*, 769–783.


**Acknowledgments** DJL is funded by an RCUK fellowship. AMH acknowledges the Leverhulme Trust for the award of a Philip Leverhulme Prize. US received funding from the Natural Environment Research Council (NE/I016287/1). DJL and AMH received funding from the Natural Environment Research Council (NE/G009112/1).
Table 1: Total mid-Pliocene global mean warming compared to preindustrial ($\Delta T$), and the global mean partitioning between CO$_2$ ($dT_{CO_2}$), orography ($dT_{orog}$), vegetation ($dT_{veg}$), and ice ($dT_{ice}$). This is shown for the default case, and cases where the sensitivity to orography and CO$_2$ are tested, as described in Sections 5.1 and 5.2.

<table>
<thead>
<tr>
<th></th>
<th>$\Delta T$ [°C]</th>
<th>$dT_{CO_2}$ [°C]</th>
<th>$dT_{orog}$ [°C]</th>
<th>$dT_{ice}$ [°C]</th>
<th>$dT_{veg}$ [°C]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Default</td>
<td>3.30</td>
<td>1.58</td>
<td>0.70</td>
<td>0.70</td>
<td>0.33</td>
</tr>
<tr>
<td>orography = modern</td>
<td>2.60</td>
<td>1.58</td>
<td>0</td>
<td>0.70</td>
<td>0.33</td>
</tr>
<tr>
<td>CO$_2$ = 350ppmv</td>
<td>2.71</td>
<td>0.99</td>
<td>0.70</td>
<td>0.70</td>
<td>0.33</td>
</tr>
<tr>
<td>CO$_2$ = 450ppmv</td>
<td>3.83</td>
<td>2.10</td>
<td>0.70</td>
<td>0.70</td>
<td>0.33</td>
</tr>
</tbody>
</table>
**Figure Captions**

Figure 1: Factor separation for a function of two variables - in this case CO$_2$ and orography. (a) is the linear approach (Equation 2), (b) is the Stein and Alpert (1993) approach (Equation 5), and (c) is our approach (Equation 7 or Equation 8).

Figure 2: Factor separation used in our study for two functions of two variables each - in this case CO$_2$, orography, vegetation, and ice (Equation 13).

Figure 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 ($E_o$, $E_c$, and $E_{oc}$), see Table 1 of Supplementary Information of Lunt *et al.* (2010a).
Figure 4: (a) Simulated annual mean surface air temperature change, mid-Pliocene minus pre-industrial, $\Delta T$. (b-e) Surface air temperature changes due to (b) CO$_2$ ($dT_{CO_2}$), (c) orography ($dT_{orog}$), (d) vegetation ($dT_{veg}$), and (e) ice ($dT_{ice}$); as calculated from Equation 13.

Figure 5: Zonal annual mean surface air temperature changes due to CO$_2$ ($dT_{CO_2}$), orography ($dT_{orog}$), vegetation ($dT_{veg}$), and ice ($dT_{ice}$) [°C].

Figure 6: Surface air temperature change due to CO$_2$ alone calculated as (a) Equation 2 and (b) Equation 3. The difference between the two approaches (equal to the synergy, $S$ in Equation 5) is shown in (c).

Figure 7: (a) Difference in mid-Pliocene surface air temperature between our simulation and that of Haywood and Valdes (2004). (b) The same, but for the mid-Pliocene anomalies, mid-Pliocene minus pre-industrial.
First quantification of the relative influences on mid-Pliocene warmth and polar amplification of CO2, orography, vegetation, and ice sheets.

A new factorisation technique, an improvement on the traditional Stein+Alpert approach.

A quantitative assessment of the uncertainties in our results.
\begin{align*}
\text{dTco}_2 &= T_c - T \\
\text{dTorog} &= T_{oc} - T_c \\
S &= T_{oc} - T_o - T_c + T \\
\text{dTco}_2 &= 1/2((T_c - T) + (T_{oc} - T)) \\
\text{dTorog} &= 1/2((T_o - T) + (T_{oc} - T_c)) \\
\end{align*}
\[ \frac{dT_{\text{torog}}}{dt} = \frac{1}{2} \left( (T_{\text{oc}} - T_{\text{c}}) + (T_{\text{oc}} - T_{\text{c}}) \right) \]
\[ \frac{dT_{\text{ice}}}{dt} = \frac{1}{2} \left( (T_{\text{oci}} - T_{\text{oc}}) + (T_{\text{ocvi}} - T_{\text{ocv}}) \right) \]
\[ \frac{dT_{\text{co2}}}{dt} = \frac{1}{2} \left( (T_{\text{c}} - T_{\text{oc}}) + (T_{\text{oc}} - T_{\text{o}}) \right) \]
\[ \frac{dT_{\text{veg}}}{dt} = \frac{1}{2} \left( (T_{\text{ocvi}} - T_{\text{ocv}}) + (T_{\text{ocv}} - T_{\text{oci}}) \right) \]
Figure 3
Click here to download Figure: plio-causes-eps1_3.0_Fig3.pdf
Figure 4
Click here to download Figure: plio_causes_eps1_3.0_Fig4.pdf
Figure 7
Click here to download Figure: plio.causes.epsl.3.0_Fig7.pdf