Cenozoic climate changes: A review based on time series analysis of marine benthic $\delta^{18}$O records

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Abstract The climate during the Cenozoic era changed in several steps from ice-free poles and warm conditions to ice-covered poles and cold conditions. Since the 1950s, a body of information on ice volume and temperature changes has been built up predominantly on the basis of measurements of the oxygen isotopic composition of shells of benthic foraminifera collected from marine sediment cores. The statistical methodology of time series analysis has also evolved, allowing more information to be extracted from these records. Here we provide a comprehensive view of Cenozoic climate evolution by means of a coherent and systematic application of time series analytical tools to each record from a compilation spanning the interval from 4 to 61 Myr ago. We quantitatively describe several prominent features of the oxygen isotope record, taking into account the various sources of uncertainty (including measurement, proxy noise, and dating errors). The estimated transition times and amplitudes allow us to assess causal climatological-tectonic influences on the following known features of the Cenozoic oxygen isotopic record: Paleocene-Eocene Thermal Maximum, Eocene-Oligocene Transition, Oligocene-Miocene Boundary, and the Middle Miocene Climate Optimum. We further describe and causally interpret the following features: Paleocene-Eocene warming trend, the two-step, long-term Eocene cooling, and the changes within the most recent interval (Miocene-Pliocene). We review the scope and methods of constructing Cenozoic stacks of benthic oxygen isotope records and present two new latitudinal stacks, which capture besides global ice volume also bottom water temperatures at low (less than 30$^\circ$) and high latitudes. This review concludes with an identification of future directions for data collection, statistical method development, and climate modeling.

1. Introduction

Although life changed dramatically at the end of the Cretaceous period/Mesozoic era [Stanley, 1989] around 65 Myr ago [Gradstein et al., 2004], global climate during the beginning of the Cenozoic era continued in the warm mode that had persisted before [Press and Siever, 1986]. The early Cenozoic was characterized by higher global temperatures than today, smaller temperature gradients between low and high latitudes, an almost complete absence of continental ice, and levels of atmospheric carbon dioxide concentration perhaps as high as 1500 parts per million by volume (ppmv) [Zachos et al., 2001a, and references therein]. Since then, the variables describing the Earth’s atmosphere, climate system, and cryosphere, experienced substantial changes. The long-term climate change during the Cenozoic corresponds at first order to a cooling, which drove Earth from a state without ice caps to one with two poles glaciated [Fischer, 1981].

To a second, more detailed order the overall Cenozoic climate trend can be considered as a succession of smaller changes: more gradual transitions, such as the cooling during the Eocene epoch [Broecker, 1995; Seibold and Berger, 1996], which is roughly the interval [34 Ma; 56 Ma] [Gradstein et al., 2004], and more abrupt, event-like changes, such as the Eocene-Oligocene Transition (EOT) [Miller et al., 1987, 1991], roughly 34 Myr ago. This succession of long- and short-term climate transitions was not a monotonic series of cooling: warming also occurred [Kennett, 1982; Cronin, 2010].

Achieving a comprehensive understanding of the driving actions and reactions requires an assessment of both long-term causal influences, such as tectonic changes [Crowley and Burke, 1998], and short-term causes, such as changes in atmospheric greenhouse gas concentrations at the Paleocene-Eocene Thermal Maximum (PETM) event [DeConto et al., 2012], about 56 Myr ago, or at the Miocene-Pliocene boundary, about 5 Myr ago. This is the dynamical approach, which takes the timescales of changes into account. It is
the basis for obtaining insight into Cenozoic climate physics, its various processes, and their interactions that led to the recorded climate history.

Quantitative analysis of climate data should take all uncertainties into account in order to obtain results with realistic error bars [Mudelsee, 2010] and hence allow more rigorous testing of scientific hypotheses [Popper, 1935]. The quantitative statistical approach helps also with testing and comparing paleoclimate model variants [Saltzman, 2002; Schmidt et al., 2014], for stimulating new model developments.

Several books or book chapters exist on Cenozoic climate changes. Kennett [1982] offers a marine perspective, while Crowley and North [1991] focus on the development of computer models of Cenozoic climate. Broecker [1995] and Seibold and Berger [1996] consider basic ideas and conceptual models. Crowley and Burke [1998] deal with the slow, tectonic causal actions, and Cronin [2010] gives a recent overview of the various empirical findings. Various review articles are concerned with certain points regarding Cenozoic climate changes, such as sea level and continental margin erosion [Miller et al., 1987], modeling onset of glaciation [Crowley and North, 1990], plateau uplift [Ruddiman and Kutzbach, 1990; Ruddiman et al., 1997], the “snow gun hypothesis” [Prentice and Matthews, 1991], changes in Antarctica [Ehrmann et al., 1992; Shevenell and Kennett, 2007] and South America [Le Roux, 2012a, 2012b], changes in the Pacific [Lyle et al., 2008], the carbon cycle [Zachos et al., 2008], and deep-sea temperatures and global ice volume [Lear et al., 2000]. Marine sediment cores represent the climate archive most commonly used in the studies described above, with oxygen isotope measurements being the climate proxy variable predominantly considered, indicating changes in global ice volume and ocean water temperature.

However, there are, up to date, a limited number of studies that include a rigorous statistical treatment of these data, although the benefits of this approach have long been acknowledged. Shackleton [1982, p. 199 therein] demonstrates the “feasibility of gathering a data base for examining climatic variability without [the] usual bias toward the recent” and considerable timescale uncertainties, which, at the time of writing, were seldom better than 1 Myr. More recently, Zachos et al. [2001a] focused on the periodic and anomalous components of variability over the early Cenozoic portion, for which they compiled a large data set, and Cramer et al. [2009], concerned with ocean overturning since the late Cretaceous, used an even larger data set and employed advanced statistical bootstrap simulation methods to obtain climate trend estimates with error bars.

It is desirable to have a curve representing global climate over the Cenozoic since this gives orientation and allows records of regional climate to be put into context. It also facilitates comparison with output from conceptual and higher-resolved global climate models [Crowley and North, 1990, 1991; Zhisheng et al., 2001; DeConto and Pollard, 2003; Nisancioglu et al., 2003], which are constructed to test hypotheses about Cenozoic climate mechanisms. For achieving this objective, stacks of foraminiferal oxygen isotope records were constructed from a multitude of marine sediment cores, with the rationale that regional temperature variations are attenuated and a global signal, representing global ice volume and temperature, emerges. (In paleoclimatology, a stack is a summary curve made from several individual curves by means of an averaging procedure.) Of particular relevance for stack construction has been the usage of shells of benthic dwelling foraminifera [Miller et al., 1987; Prentice and Matthews, 1988], although tropical, planktic foraminifera have also been used [Prentice and Matthews, 1988]. An important step has been the construction of Zachos et al.’s stack of benthic oxygen isotopes [Zachos et al., 2001a], which is based on data compiled from more than 40 marine drilling sites. This record (Figure 1) shows Cenozoic climate evolution at high precision, owing to the large number of sites. However, due to the uneven spatial and temporal distribution of the individual data, even that stack may not be free of bias [Zachos et al., 2001a, 2008]. A more recent benthic oxygen isotope stack [Cramer et al., 2009] is based on an even larger and more recent data compilation. A principal interpretative challenge arises from the time-dependent mixing of the ice volume and temperature signals in the oxygen isotope values. This is evidently less problematic for the earlier part of the Cenozoic, prior to about 34 Myr ago, when only small [Miller et al., 1987; Zachos et al., 2001a; Tripati et al., 2008] or no ice sheets are thought to have existed (Figure 1), but it is more problematic for the later part. Attempts [Lear et al., 2000; Cramer et al., 2011] have been made to reconcile both signal contributions by means of other records, such as the Mg/Ca elemental ratio as a proxy for temperature or sea level as an equivalent for ice volume, but those corrections, currently based on a sparse data set, may introduce considerable new uncertainties.

Another recent line of development regards methods of statistical time series analysis, developed, adapted, and tested by one of us [Mudelsee, 2010]. These methods are specifically tailored to meet the analytical...
Figure 1. Cenozoic climate evolution. The proxy of oxygen isotopic composition (δ¹⁸O) obtained from benthic foraminiferal shells found in marine sediment cores [Zachos et al., 2001a] indicates changes in global ice volume (from around 34 Ma) and bottom water temperature, heavier δ¹⁸O values reflecting more ice/lower temperatures. The δ¹⁸O values are adjusted to the genus Cibicidoides and 0.64‰ added [Shackleton and Hall, 1984; Zachos et al., 2001a]. The record is a stack based on data from several drill sites and smoothed by means of a five-point running mean. The timescale of the δ¹⁸O stack is after Berggren et al. [1995], while the boundaries of the Cenozoic epochs are after Gradstein et al. [2004]. The temperature axis [Zachos et al., 2001a] was computed on the basis of an ice-free ocean; it applies to the time before the start of the EOT glaciation (about 34 Ma; see Table 8). The analyzed transitions and events are Paleocene-Eocene warming trend (PE-Trend), Paleocene-Eocene Thermal Maximum (PETM), Long-term Eocene Cooling I (LTEC-I), Long-term Eocene Cooling II (LTEC-II), Eocene-Oligocene Transition (EOT), the Oligocene “swinging” trends (O-Swings), Oligocene-Miocene Boundary (OMB), Mid-Miocene Climatic Optimum (MMCO), and interval from 4 to 10 Ma (4-to-10-Ma); the Early Eocene Climatic Optimum (EECO) is not analyzed separately but in its relation with PE-Trend and LTEC-I; likewise, the Mid-Eocene Climatic Optimum (MECO) is analyzed in its relation with LTEC-I and LTEC-II. Some transitions and events (asterisks) have not yet been explicitly named in the literature.

needs of climatologists, who are concerned with quantifying climate transitions and constructing composites and who wish to provide estimation results with realistic error bars. To appreciate the statistical approach, consider the task of quantifying a climate transition. A statistical regression model comprises a trend component (“signal”), corresponding to the “true” climate change, and a noise component, summarizing the unknown influences. For oxygen isotope records, the trend may correspond to a long-term ice volume change, and the noise may correspond to short-term influences, such as diagenesis, local water temperature fluctuations, measurement error, and so forth. While the statistical approach would correctly extract the trend component, another approach could incorrectly look just on the extreme values and infer a too large climate transition amplitude. This overestimation would thus result from wrongly interpreting noise effects. In the present review, we employ our methods of climate time series analysis [Mudelsee, 2010] and utilize the recent, large data compilation of marine benthic oxygen isotope records [Cramer et al., 2009]. We put our analysis into context with existing results and overviews. This “quantitative reanalysis review” is aimed at advancing the quantitative and causal understanding of Cenozoic climate changes.

Within the database, we distinguish between low and high latitudes to accommodate bottom water temperature differences contained in benthic foraminiferal oxygen isotope records. In the quantitative analytical approach, we are confronted with various sources of uncertainty (measurement, proxy) and also the “challenging properties” of real-world paleoclimatic time series: nonnormal distributional shape, autocorrelation (also called persistence or serial dependence), and uneven time spacing. We meet these challenges by performing computing-intensive bootstrap simulations [Mudelsee, 2010]. We further take into account another uncertainty source, dating errors, and uncertain timescales. The bootstrap approach is employed for enhancing two time series procedures. First, with parametric regression we fit change-point models [Mudelsee, 2000, 2009] to the records. This yields change-point times and amplitudes of changes with realistic error bars. Such knowledge is indispensable for assessing causes of Cenozoic climate transitions. We compare our transition parameter estimates with those from previous papers. Second, with nonparametric regression [Mudelsee et al., 2012] we smooth the pooled data set to obtain stacks that are not parametrically restricted. The resulting two stacks (low and high latitudes) with uncertainty band are compared with the existing benthic stack [Zachos et al., 2001a, 2008], for which no uncertainty band has been published. We note two caveats. First, although our stacks are based on a larger data set [Cramer et al., 2009] than Zachos et al.’s stack, the uneven spatiotemporal data distribution may introduce bias also in our stacks.
Second, our approach of comparing low with high latitudes may yield biased results for time intervals of strong latitudinal dependent evolutionary processes.

The presented work on quantifying transitions and events in Cenozoic climate evolution draws on previously identified features, such as the glaciation at the EOT, or the PETM, but it also suggests features that have, to the best of our knowledge, not yet been explicitly named and/or quantified in the literature (Figure 1).

In this review we first describe the data material (section 2), thereby heavily relying on the extensive work carried out by the compilers [Cramer et al., 2009]. We list the employed seafloor drilling sites, introduce the notation for oxygen isotopes, and evaluate the precision of the timescales. The time series analysis methods (section 3) comprise the parametric regression models and stack construction via nonparametric regression. This section also explains in two parts the error analytical methods. In section 4, where we discuss the results, we first consider the statistical estimates (section 4.1), going from older to younger epochs. We start in the middle of the Paleocene, at 61 Ma, and end in the Pliocene, at 4 Ma. For older time intervals, the database becomes too sparse to allow meaningful application of the advanced time series methods. For younger time intervals, which include the major part of the Northern Hemisphere Glaciation (NHG), the abundance of material and the achieved temporal resolution of records is a magnitude better than for the [4 Ma; 61 Ma] interval; our quantitative statistical approach has already been applied to the [2 Ma; 4 Ma] interval and used for assessing causal explanations of the NHG [Mudelsee and Raymo, 2005]. In section 4.1, for each individual climate transition and event, we also compare our own estimation results with previous results from the literature and assess the causal explanations brought forward. Section 4.3 presents the new oxygen isotope stacks. We conclude this review by summarizing the essential results and causal interpretations of the Cenozoic climate evolution (section 5). Therein, we further identify future research directions for studying Cenozoic climate changes regarding the sampling of data, the adaptation of statistical analytical tools, and the formulation of climate models.

2. Data

Scientific drilling into the ocean floor (Deep Sea Drilling Project [1969–1986], Ocean Drilling Program [1986–2004, 1988–2007], and Integrated Ocean Drilling Program [2005]; see also the publications from the successor International Ocean Discovery Program) has over the past four decades led to an impressive climate archive of marine sediment cores. Cramer et al. [2009] tapped this archive to produce a large database (34,479 data entries) of Cenozoic δ18O records. In this review, we employ their database, perform some initial data checks, and select the records suitable for our purpose of time series analysis.

The checks and preliminary modifications of the database [Cramer et al., 2009] (see the supporting information) consist in removing missing values (time given but not δ18O), testing for strictly monotonically increasing time values (per record), and averaging δ18O values for which identical time values exist. Since the statistical time series analysis methods (section 3) are applied on a site-by-site basis for each transition, a certain minimum sampling density is required. On the other hand, it is preferable to maximize the number of records analyzed to achieve a fuller spatial (global) coverage. Table 1 shows our database for the low latitudes and Table 2 for the high latitudes. We solved this dilemma problem by setting the minimum sample size per record to 17.

These data sets accompany this review as supporting information for helping the readers who wish to replicate the results.

2.1. Seafloor Drilling Sites

The employed low-latitude sites amount to 16, and they cover the oceanic area reasonably well; the high-latitude sites amount to 32, and they show a better coverage. The division between low and high latitudes at 30°N and 30°S is not followed strictly since geographical positions changed during the Cenozoic. For example, the position of ODP Site 1209 moved from south of 30°N during the recorded interval [37.7 Ma; 60.5 Ma] to north of 30°N at present; hence, we considered ODP 1209 as indicative of low latitudes. Cramer et al. [2009, Figure 1 therein] show the paleogeographic positions of many sites.

2.2. Oxygen Isotopes

Oxygen isotopic composition is usually expressed in delta notation [Bradley, 1999]:

$$\delta^{18}O = 1000 \frac{R_{\text{Sample}} - R_{\text{Standard}}}{R_{\text{Standard}}} \cdot (\%\text{o})$$  (1)
Table 1. Database: Low-Latitude Records

<table>
<thead>
<tr>
<th>Record</th>
<th>Geographical Position</th>
<th>Water Depth (m)</th>
<th>Time Interval (Ma)</th>
<th>Average (kyr)</th>
<th>Maximum (kyr)</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSDP 77</td>
<td>0°N 133°W</td>
<td>4291</td>
<td>5.1; 34.1</td>
<td>207</td>
<td>1851</td>
</tr>
<tr>
<td>DSDP 289</td>
<td>0°N 159°E</td>
<td>2206</td>
<td>6.2; 20.6</td>
<td>55</td>
<td>529</td>
</tr>
<tr>
<td>DSDP 317</td>
<td>11°S 162°W</td>
<td>2560</td>
<td>5.6; 23.3</td>
<td>274</td>
<td>2289</td>
</tr>
<tr>
<td>DSDP 366</td>
<td>6°N 20°W</td>
<td>2853</td>
<td>9.0; 35.7</td>
<td>281</td>
<td>2381</td>
</tr>
<tr>
<td>DSDP 574</td>
<td>4°N 133°W</td>
<td>4561</td>
<td>10.2; 34.7</td>
<td>48</td>
<td>4747</td>
</tr>
<tr>
<td>DSDP 577</td>
<td>32°N 158°E</td>
<td>2675</td>
<td>38.0; 60.9</td>
<td>246</td>
<td>5907</td>
</tr>
<tr>
<td>ODP 667</td>
<td>5°N 22°W</td>
<td>3529</td>
<td>6.0; 30.2</td>
<td>171</td>
<td>1950</td>
</tr>
<tr>
<td>ODP 803</td>
<td>2°N 161°E</td>
<td>3422</td>
<td>16.3; 34.5</td>
<td>198</td>
<td>2981</td>
</tr>
<tr>
<td>ODP 806</td>
<td>0°N 159°E</td>
<td>2520</td>
<td>4.3; 19.2</td>
<td>127</td>
<td>847</td>
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<tr>
<td>ODP 865</td>
<td>18°N 180°W</td>
<td>1517</td>
<td>36.0; 60.9</td>
<td>444</td>
<td>2816</td>
</tr>
<tr>
<td>ODP 926</td>
<td>4°N 43°W</td>
<td>3598</td>
<td>4.0; 26.9</td>
<td>7</td>
<td>1434</td>
</tr>
<tr>
<td>ODP 929</td>
<td>6°N 44°W</td>
<td>4356</td>
<td>15.1; 25.1</td>
<td>17</td>
<td>2180</td>
</tr>
<tr>
<td>ODP 1209</td>
<td>33°N 159°E</td>
<td>2387</td>
<td>37.7; 60.9</td>
<td>401</td>
<td>1428</td>
</tr>
<tr>
<td>ODP 1218</td>
<td>9°N 135°W</td>
<td>4828</td>
<td>19.0; 41.0</td>
<td>7</td>
<td>373</td>
</tr>
<tr>
<td>IODP 1258</td>
<td>9°N 55°W</td>
<td>3192</td>
<td>44.6; 53.8</td>
<td>64</td>
<td>697</td>
</tr>
<tr>
<td>IODP 1260</td>
<td>9°N 55°W</td>
<td>2549</td>
<td>39.4; 47.8</td>
<td>70</td>
<td>1470</td>
</tr>
</tbody>
</table>

*DSDP, Deep Sea Drilling Project; ODP, Ocean Drilling Program; IODP, Integrated Ocean Drilling Program.

a For paleogeographical position and paleo water depth, see Cramer et al. [2009]. For references of δ¹⁸O data, see Cramer et al.[2009] (supporting information); for references and methods of timescale construction, see section 2.3.

where $R$ is the number ratio of $^{18}$O to $^{16}$O isotopes and the index refers to the sample or a standard; for the employed database, VPDB isthe standard material against which the sample is compared. Vital effects can produce δ¹⁸O offsets in different foraminiferal genera and species; this was corrected for [Cramer et al., 2009] by adjusting [Shackleton and Hall, 1984] isotope values to a common genus (Cibicidoides). Diagenetic effects on the δ¹⁸O value of shells of benthic foraminifera are thought to be small [Edgar et al., 2013].

2.3. Dating and Timescale Construction

The timescales of the originally published δ¹⁸O records used biostratigraphic and magnetostratigraphic events identified in the sediment records. Since the assumed age values for those events have been updated over the years, Cramer et al. [2009] adjusted the dates by linear interpolation to two currently accepted Cenozoic timescales; in this review we employ their adjustment to the Gradstein et al. [2004] timescale. Cramer et al. [2009] then readjusted age models for the middle to late Eocene portions, between approximately 40 and 34 Ma, by linear interpolation. It should therefore be kept in mind that agreement of estimated middle to late Eocene transition times among records may be partly due to that readjustment.

On the basis of the adjustments to the common timescale [Gradstein et al., 2004] and the readjustment for the middle to late Eocene, Cramer et al. [2011] concluded that the relative precision of dates (i.e., among records) is less than $s_{date} = 0.1$ Myr. We use that $s_{date}$ value in a conservative approach to including timescale errors in the statistical estimations. The absolute precision of dates (i.e., with respect to true time) may be larger.

All of the analyzed records (Tables 1 and 2) have an average time spacing or resolution of better than 900 kyr; several records have an average resolution of a few tens of kiloyears. However, none of the records covers the whole interval [4 Ma; 61 Ma]. Many records exhibit large hiatuses, that is, data gaps for which no meaningful statistical analysis can be performed. Tables 1 and 2 give not only the average but also the maximum time spacing. If the maximum is clearly larger than the average (e.g., for DSDP 525), then this may indicate the presence of larger gaps.

3. Time Series Analysis Methods

Following the convention in the statistical analysis of time series [Priestley, 1981; von Storch and Zwiers, 1999], we denote the measured values (δ¹⁸O) of a record as $x(i)$ and the measured time values (ages) as $t(i)$. 
Table 2. Database: High-Latitude Records

<table>
<thead>
<tr>
<th>Record</th>
<th>Geographical Position(^a)</th>
<th>Water Depth(^a) (m)</th>
<th>Time Interval (Ma)</th>
<th>Average Interval (kyr)</th>
<th>Maximum Interval (kyr)</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSDP 281</td>
<td>48°S 148°E</td>
<td>1591</td>
<td>[7.9; 20.6]</td>
<td>236</td>
<td>3180</td>
</tr>
<tr>
<td>DSDP 334</td>
<td>37°N 34°W</td>
<td>2619</td>
<td>[5.3; 11.1]</td>
<td>251</td>
<td>986</td>
</tr>
<tr>
<td>DSDP 357</td>
<td>30°S 36°W</td>
<td>2086</td>
<td>[8.6; 21.5]</td>
<td>807</td>
<td>1610</td>
</tr>
<tr>
<td>DSDP 360</td>
<td>36°S 18°E</td>
<td>2949</td>
<td>[6.0; 23.8]</td>
<td>194</td>
<td>3111</td>
</tr>
<tr>
<td>DSDP 397</td>
<td>27°N 15°W</td>
<td>2900</td>
<td>[4; 9.7]</td>
<td>178</td>
<td>583</td>
</tr>
<tr>
<td>DSDP 401</td>
<td>47°N 9°W</td>
<td>2495</td>
<td>[34.9; 56.0]</td>
<td>351</td>
<td>1870</td>
</tr>
<tr>
<td>DSDP 410</td>
<td>46°N 20°W</td>
<td>2975</td>
<td>[5.2; 12.0]</td>
<td>148</td>
<td>542</td>
</tr>
<tr>
<td>DSDP 516</td>
<td>30°S 35°W</td>
<td>1313</td>
<td>[4.0; 20.9]</td>
<td>392</td>
<td>6539</td>
</tr>
<tr>
<td>DSDP 522</td>
<td>26°S 5°W</td>
<td>4441</td>
<td>[23.0; 34.6]</td>
<td>26</td>
<td>867</td>
</tr>
<tr>
<td>DSDP 525</td>
<td>29°S 3°E</td>
<td>2467</td>
<td>[4; 61.0]</td>
<td>227</td>
<td>19,950</td>
</tr>
<tr>
<td>DSDP 527</td>
<td>28°S 2°E</td>
<td>4428</td>
<td>[4; 56.9]</td>
<td>894</td>
<td>22,987</td>
</tr>
<tr>
<td>DSDP 529</td>
<td>29°S 3°E</td>
<td>3035</td>
<td>[21.5; 44.1]</td>
<td>279</td>
<td>7113</td>
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<tr>
<td>DSDP 549</td>
<td>49°N 13°W</td>
<td>2513</td>
<td>[24.0; 38.6]</td>
<td>768</td>
<td>1613</td>
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<tr>
<td>DSDP 552</td>
<td>56°N 23°W</td>
<td>2301</td>
<td>[4.0; 10.2]</td>
<td>21</td>
<td>348</td>
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<tr>
<td>DSDP 553</td>
<td>58°N 23°W</td>
<td>2328</td>
<td>[4.0; 23.6]</td>
<td>213</td>
<td>691</td>
</tr>
<tr>
<td>DSDP 555</td>
<td>57°N 21°W</td>
<td>1659</td>
<td>[6.4; 18.4]</td>
<td>226</td>
<td>2683</td>
</tr>
<tr>
<td>DSDP 563</td>
<td>34°N 44°W</td>
<td>3786</td>
<td>[9.3; 32.7]</td>
<td>113</td>
<td>2047</td>
</tr>
<tr>
<td>DSDP 608</td>
<td>43°N 23°W</td>
<td>3526</td>
<td>[5.6; 24.2]</td>
<td>119</td>
<td>593</td>
</tr>
<tr>
<td>ODP 689</td>
<td>65°S 3°E</td>
<td>2080</td>
<td>[23.8; 57.0]</td>
<td>54</td>
<td>7283</td>
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<tr>
<td>ODP 690</td>
<td>65°S 7°E</td>
<td>2914</td>
<td>[24.5; 58.1]</td>
<td>131</td>
<td>4837</td>
</tr>
<tr>
<td>ODP 698</td>
<td>51°S 33°W</td>
<td>2138</td>
<td>[48.7; 57.2]</td>
<td>529</td>
<td>1210</td>
</tr>
<tr>
<td>ODP 702</td>
<td>51°S 26°W</td>
<td>3083</td>
<td>[35.8; 58.3]</td>
<td>382</td>
<td>2176</td>
</tr>
<tr>
<td>ODP 703</td>
<td>47°S 8°E</td>
<td>1796</td>
<td>[22.5; 37.0]</td>
<td>268</td>
<td>1475</td>
</tr>
<tr>
<td>ODP 704</td>
<td>47°S 7°E</td>
<td>2532</td>
<td>[4; 29.5]</td>
<td>74</td>
<td>5447</td>
</tr>
<tr>
<td>ODP 738</td>
<td>63°S 83°E</td>
<td>2253</td>
<td>[33.4; 60.8]</td>
<td>184</td>
<td>2125</td>
</tr>
<tr>
<td>ODP 744</td>
<td>62°S 81°E</td>
<td>2307</td>
<td>[10.0; 36.7]</td>
<td>40</td>
<td>1670</td>
</tr>
<tr>
<td>ODP 747</td>
<td>55°S 77°E</td>
<td>1697</td>
<td>[40; 26.7]</td>
<td>130</td>
<td>2495</td>
</tr>
<tr>
<td>ODP 748</td>
<td>58°S 79°E</td>
<td>1288</td>
<td>[23.7; 40.8]</td>
<td>63</td>
<td>1401</td>
</tr>
<tr>
<td>ODP 1051</td>
<td>30°N 76°W</td>
<td>1981</td>
<td>[36.6; 55.9]</td>
<td>85</td>
<td>18,337</td>
</tr>
<tr>
<td>ODP 1088</td>
<td>41°S 14°E</td>
<td>2082</td>
<td>[4.0; 8.7]</td>
<td>54</td>
<td>296</td>
</tr>
<tr>
<td>ODP 1090</td>
<td>43°S 5°E</td>
<td>3702</td>
<td>[15.9; 24.2]</td>
<td>11</td>
<td>110</td>
</tr>
<tr>
<td>ODP 1171</td>
<td>48°S 149°E</td>
<td>2148</td>
<td>[11.1; 16.9]</td>
<td>10</td>
<td>292</td>
</tr>
</tbody>
</table>

\(^a\)For paleogeographical position and paleo water depth, see Cramer et al. [2009]. For references of \(\delta^{18}O\) data, see Cramer et al. [2009] (supporting information); for references and methods of timescale construction, see section 2.3.

The index \(i\) runs from 1 to \(n\) (sample size), and we denote this measured sample as \(\{T(i), X(i)\}_{i=1}^{n}\). From this level of measured values, statistical science [Priestley, 1981; Wasserman, 2004] distinguishes the level of the process, \(\{T(i), X(i)\}_{i=1}^{n}\), that generated the sample. The task of statistical inference is to guess the properties of the process on the basis of the sample. The type of inference employed for this review is regression estimation, where we estimate the trend, that is, the long-term systematic relationship between time, \(T(i)\), and climate, \(X(i)\). The convention uses the “hat notation” for distinguishing between the true but unknown trend parameter (e.g., the slope, \(\beta_1\), in a linear model) and its estimate (slope estimate, \(\hat{\beta}_1\)).

First, we give the motivation for and the concepts of the regression models we employ to quantify Cenozoic climate trends (section 3.1). Then we explain how we determined the uncertainties (1σ errors) associated with the estimations (section 3.2), the typical size of the deviation between true value (\(\beta_1\)) and estimate (\(\hat{\beta}_1\)).

The mathematical algorithms of the presented regression models [Mudelsee, 2010] contain more details (e.g., on numerical tools). The uncertainty determination methods have been tested by means of Monte Carlo experiments [Mudelsee, 2010], where one generates many artificial series with known (prescribed) properties and studies how well the estimation method infers what has been prescribed.

Whereas regression models are applied to each of the 48 original, unsmoothed \(\delta^{18}O\) time series (section 2) separately, the stacking procedure provides synoptic views. We examine both the high and the low
Figure 2. Regression models. (a) The linear model has two parameters (intercept and slope), (b) the ramp model has four parameters, and (c) the break model has four parameters (t1 is constrained as left, t3 as right bound of the time interval).

Ordinary least squares (OLS) yields a straightforward estimation of regression parameters. The estimators \( \hat{\beta}_0 \) and \( \hat{\beta}_1 \) minimize the sum of squares of differences between data and model,

\[
SSQ_{\text{lin}}(\hat{\beta}_0, \hat{\beta}_1) = \sum_{i=1}^{n} [x(i) - x_{\text{lin}}(i)]^2,
\]

where \( x_{\text{lin}}(i) \) is given by \( x_{\text{lin}}(i) = X_{\text{lin}}(i) + S \cdot X_{\text{noise}}(i) \). The solutions \( \hat{\beta}_0 \) and \( \hat{\beta}_1 \) can be found in textbooks [Montgomery and Peck, 1992; von Storch and Zwiers, 1999; Mudelsee, 2010].

3.1.2. Ramp Regression

The ramp regression [Mudelsee, 2000] employs a nonlinear model with two change points (Figure 2b),

\[
X(i) = X_{\text{ramp}}(i) + S \cdot X_{\text{noise}}(i),
\]

\[
X_{\text{ramp}}(i) = \begin{cases} 
  x_1 & \text{for } T(i) \leq t_1, \\
  x_1 + (T(i) - t_1)(x_2 - x_1)/(t_2 - t_1) & \text{for } t_1 < T(i) \leq t_2, \\
  x_2 & \text{for } T(i) > t_2.
\end{cases}
\]
This is the most straightforward parametric approach for analyzing climate change questions such as: when did the transition start (answer: t2), when did it end (t1), and what was the amplitude of the change (x2−x1)? An OLS fit criterion minimizes
\[
SSQ_{\text{ramp}}(t1, x1, t2, x2) = \sum_{i=1}^{n} [x(i) - \hat{x}_{\text{ramp}}(i)]^2 .
\]
where \( \hat{x}_{\text{ramp}}(i) \) is the sample version of \( X_{\text{ramp}}(i) \). If \( \hat{t1} \) and \( \hat{t2} \) were known, then the solutions \( \hat{x1} \) and \( \hat{x2} \) followed directly from analytical minimization of \( SSQ_{\text{ramp}} \) [Mudelsee, 2000]. Since \( \hat{t1} \) and \( \hat{t2} \) are unknown, one uses a brute force search over all combinations of \( \hat{t1} \) and \( \hat{t2} \) from the set \{t(i)\}_{i=1}^{n} [Mudelsee, 2000]; for the data sizes encountered (section 2), such minimization costs are insignificant.

3.1.3. Break Regression
The break regression [Mudelsee, 2009] employs a nonlinear model with one change point (Figure 2c),
\[
X(i) = X_{\text{break}}(i) + S \cdot X_{\text{noise}}(i).
\]
\[
X_{\text{break}}(i) = \begin{cases} 
  x1 + [T(i) - t1](x2 - x1)/(t2 - t1) & \text{for } T(i) \leq t2, \\
  x2 + [T(i) - t2](x3 - x2)/(t3 - t2) & \text{for } T(i) > t2.
\end{cases}
\]
An alternative formulation would comprise the four parameters \( t2, x2, \beta_1 = (x2 - x1)/(t2 - t1), \) and \( \beta_2 = (x3 - x2)/(t3 - t2) \). The break can be useful for describing a change in linear trend at one point \( (t2, x2) \), from slope \( \beta_1 \) to \( \beta_2 \).

An OLS fit criterion minimizes
\[
SSQ_{\text{break}}(x1, t2, x2, x3) = \sum_{i=1}^{n} [x(i) - \hat{x}_{\text{break}}(i)]^2 .
\]
where \( \hat{x}_{\text{break}}(i) \) is the sample version of \( X_{\text{break}}(i) \). Analogous to the ramp: if \( \hat{t2} \) were known, then the solutions \( \hat{x1}, \hat{x2}, \) and \( \hat{x3} \) followed directly from analytical minimization [Mudelsee, 2009]. One uses a brute force search over all \( \hat{t2} \) from \{t(i)\}_{i=1}^{n} [Mudelsee, 2009].

3.2. Uncertainties I
The nonzero noise component introduces uncertainty to the estimation. For simple forms of the noise component, such as a normal distributional shape and absent autocorrelation, and additionally simple estimation problems, such as the linear regression, the estimation uncertainty can be analytically determined from the curvature of the SSQ function [Montgomery and Peck, 1992]. However, climate noise is usually more complex, regarding the shape and also the autocorrelation [von Storch and Zwiern, 1999; Mudelsee, 2010], and the change-point estimation problems encountered here (ramp, break) are more complex than the linear model. An additional source of uncertainty comes from dating errors (section 2.3). The complexities shift the uncertainty determination toward analytical intractability. This situation requires usage of computational tools of uncertainty estimation, that is, bootstrap resampling, which we explain in section 3.2.1. We also show how to combine a number of estimates with (bootstrap-determined) error bars into a weighted mean to obtain a summary estimate (section 3.2.2).

3.2.1. Bootstrap Resampling
The bootstrap computational approach [Efron and Tibshirani, 1993] resamples randomly, with replacement, from the regression residuals,
\[
e(i) = x(i) - \hat{x}_{\text{ns}}(i), \quad i = 1, \ldots, n.
\]
The \( \hat{x}_{\text{ns}}(i) \) denotes the fitted regressions: \( \hat{x}_{\text{ns}}(i), \hat{x}_{\text{ramp}}(i), \) or \( \hat{x}_{\text{break}}(i) \). This random sample is written as \( \{e^*(i)\}_{i=1}^{n} \).

The resample is formed as
\[
X^*(i) = \hat{x}_{\text{ns}}(i) + e^*(i), \quad i = 1, \ldots, n.
\]
The estimation (linear, ramp, or break) is repeated on the resample, yielding new estimates (e.g., \( \hat{t2}^* \)). The procedure resampling-estimation is repeated until \( B = 400 \) copies of \( \hat{t2}^* \) are available. The bootstrap standard error is the standard deviation over the \( B \) copies; it serves to measure the estimation uncertainty [Efron and Tibshirani, 1993].
3.2.1. Nonnormal Distributions
Not all climate variables follow the theoretically tractable situation of normal shape. Resampling from the data (residuals) preserves the distributional shape in the resample. Using the bootstrap is therefore more robust than assuming a specific distributional shape [Efron, 1979].

3.2.1.2. Autocorrelation
No climate variable follows the theoretically simple situation of absent autocorrelation. Instead, climate shows persistence, it “memorizes” past values over a range of timescales [Gilman et al., 1963; Hasselmann, 1976; Briskin and Harrell, 1980; Wunsch, 2003; Mudelsee, 2010]. To preserve autocorrelation in the resample requires not resampling point-wise from the residuals but instead doing this differently, for example, block-wise [Künsch, 1989]. The blocks should be long enough to capture the climate variable’s persistence time [Mudelsee, 2002]; see also section 3.3.1.1. The employed block-length selector [Mudelsee, 2010, equation (3.28) therein] considers also the data size, n; more data points allow to use longer blocks.

3.2.2. Weighted Mean
The climate transitions are recorded by a number, m, of benthic δ18O records (section 4). This means that m transition-time estimates are available, for example, \( \{ \hat{\tau}_j \}^m_{j=1} \). Henceforth in section 3.2.2, for brevity we omit to write the index j, and we illustrate the concept using \( \hat{\tau}_2 \).

Also, m bootstrap standard errors, \( s_{\hat{\tau}_j} \), are available. Estimates and standard errors can be accurately combined in a summary estimate, the weighted mean [Birge, 1932; Bevington and Robinson, 1992],

\[
\langle \hat{\tau}_2 \rangle = \frac{\sum \hat{\tau}_j / (s_{\hat{\tau}_j})^2}{\left[ \sum 1 / (s_{\hat{\tau}_j})^2 \right]^{1/2}}.
\]

The sums are over \( j = 1, \ldots, m \).

The internal error of the weighted mean is given by

\[
s_{\text{int}, \langle \hat{\tau}_2 \rangle} = \frac{1}{\left[ \sum 1 / (s_{\hat{\tau}_j})^2 \right]^{1/2}}\!
\]

The external error of the weighted mean is given by

\[
s_{\text{ext}, \langle \hat{\tau}_2 \rangle} = \left\{ \left[ \sum \left( \hat{\tau}_j - \langle \hat{\tau}_2 \rangle \right) / (s_{\hat{\tau}_j})^2 \right] \right\}^{1/2} / \left\{ \left[ \sum 1 / (s_{\hat{\tau}_j})^2 \right] \right\}^{1/2}.
\]

The internal error measures via the average statistical error from the individual bootstrap standard errors. The external error measures via the spread of the individual estimates. A deviation between internal and external errors indicates violated assumptions; a smaller external error may point to overestimated individual standard errors, and a larger external error may point to hidden systematic influences that are not included in the individual standard errors. We report both internal and external errors and, adopting a conservative approach [Birge, 1932], we consider the maximum of both for interpretation of results (section 4).

3.2.2.1. Dating-Error Effects
Due to dating uncertainties, the timescales of the records are not exact but exhibit a random error component with a standard deviation of \( s_{\text{date}} = 0.1 \) Myr (section 2.3). This timescale uncertainty is taken into account by means of a correction of the internal/external error values of averaged transition parameters that involve time.

For the change-point times start \( \langle \hat{\tau}_2 \rangle \) and end \( \langle \hat{\tau}_1 \rangle \) of the ramp model (Figure 2b), the correction (e.g., for \( \hat{\tau}_2 \)) is via error propagation:

\[
s_{\text{cor}, \langle \hat{\tau}_2 \rangle} = \left[ (s_{\hat{\tau}_2})^2 + (s_{\text{date}})^2 \right]^{1/2},
\]

where the prime denotes the correction. The corrected individual errors \( s'_{\hat{\tau}_j} \) enter then the weighted averaging.

The correction is applied also to the midpoint, \( \langle \hat{\tau}_1 + \hat{\tau}_2 \rangle / 2 \), of the ramp (Figure 2b) and the change-point time, \( \hat{\tau}_2 \), of the break (Figure 2c). However, it is not applied to the duration, \( \hat{\tau}_2 - \hat{\tau}_1 \), of the ramp. This is because one may expect a strong correlation of the dating-error effects on \( \hat{\tau}_1 \) and \( \hat{\tau}_2 \): If \( \hat{\tau}_2 \) has to be shifted to earlier ages, then also \( \hat{\tau}_1 \) and vice versa. (Relative dates are more accurate than absolute dates.)

Amplitude estimates are hardly affected by dating errors (no correction). Regarding estimates of the slope (i.e., amplitude/duration), we assume that the amplitude error dominates the slope error and that dating-error effects via the duration are negligible (no correction).
3.3. Stack Construction

For building the stacks of benthic δ¹⁸O across the Cenozoic (4 to 61 Ma), we pool the data points, following various predecessors in stack construction [Imbrie et al., 1984; Martinson et al., 1987; Zachos et al., 2001a; Lisiecki and Raymo, 2005]. The pooling is done into two groups: the high latitudes with 32 records (poled data size, \(n = 6360\)) and the low latitudes with 16 records (\(n = 8706\)). The two data pools are analyzed by means of non-parametric regression (section 3.3.1), also denoted as smoothing, which yields the benthic δ¹⁸O stacks. The stacks are the deterministic long-term trends; the short-term noise components are smoothed away. Uncertainty bands around the stacks are constructed using a specific adaptation of bootstrap resampling and taking dating errors into account (section 3.3.1.1). An alternative procedure of stack construction, not explored here and, to the best of our knowledge, neither in previous work, would consist in smoothing records individually and then averaging them.

### 3.3.1. Nonparametric Regression

Instead of identifying the trend component, \(X_{\text{trend}}\), with a specific linear \((X_{\text{lin}})\) or change-point function \((X_{\text{ramp}}, X_{\text{break}})\) with parameters to be estimated, the smoothing method estimates the trend at a time point \(T^*\) by, loosely speaking, averaging the data points \(X(i)\) within a neighborhood around \(T^*\). (A simple example is the running mean, where the points inside a window are averaged and the window runs along the time axis.) Better estimation properties than of the running mean can be achieved by replacing the non-smooth weighting window (points inside receive constant, positive weight and points outside zero weight) by a smooth kernel function, \(K\). We base the estimation on the kernel estimator after Gasser and Müller [1979, 1984],

\[
\hat{X}_{\text{trend}}(T^*) = h^{-1} \sum_{i=1}^{n} \int_{s(i-1)}^{s(i)} K \left( \frac{T^* - y}{h} \right) X(i) \, dy,
\]

where \(K\) is a parabola (with negative curvature), \(h\) is the bandwidth, and the sequence \(s\) satisfies \(T(i - 1) \leq s(i - 1) \leq T(i)\); we take \(s(i - 1) = (T(i - 1) + T(i))/2\) with \(s(0) = 4\) Ma and \(s(n) = 61\) Ma.

We further perform the smoothing in an adaptive manner by allowing for time-dependent bandwidth, \(h(T)\). This has the advantage that (1) the uneven time spacing and (2) heteroscedasticity or time-dependent variance can be taken into account. For example, a smaller spacing (higher resolution) or a reduced variance of the noise around the trend enables a smaller bandwidth to be used and hence finer details to be resolved. A bandwidth optimized in that manner yields more accurate trend estimates than nonoptimized smoothing. Determination of \(h(T)\) is done iteratively [Brockmann et al., 1993; Herrmann, 1997]: assume variance, calculate trend, estimate variance by means of the regression residuals, recalculate trend, and so forth. The optimized bandwidths vary between about 0.5 and 2.0 Myr (Figure 3).

### 3.3.1.1. Uncertainties II

Construction of an uncertainty band around the nonparametric trend estimate is, analogously to parametric estimation (section 3.2), based on the residuals,

\[
e(i) = x(i) - \hat{X}_{\text{trend}}(i), \quad i = 1, \ldots, n,
\]

where \(\hat{X}_{\text{trend}}(i)\) is the fitted nonparametric regression (equation (17)) at time point \(T = t(i)\). In the following part of this section we refer to the index \(i = 1, \ldots, n\) and the data size, \(n\), in a “record-wise” manner, since estimation of persistence, resampling, and timescale simulation is performed for each record separately.

A simple model of red noise persistence of climatic fluctuations for discrete time and uneven spacing is the AR(1) process, \(X_{\text{non}}(i) = \exp\{-[(T(i) - (T(i-1)))/\tau]\} X_{\text{non}}(i-1) + \text{a random innovation}\). The persistence time, \(\tau\), can be estimated from data by numerical minimization of a least squares cost function [Mudelsee, 2002]. For each record, the persistence model is fitted to the kernel regression residuals. The resulting persistence time values (Table 3) are in the order of a few kiloyears to a few tens of kiloyears. Climatological interpretation is deferred to section 4.2. We note that for uncertainty band construction, the bootstrap resampling
adopts a persistence time of $r = 41$ kyr because Cenozoic climate noise may show signs of Milankovitch’s obliquity variations, which act on this timescale [Berger, 1978]. The value of 41 kyr is somewhere on the upper limit of estimates (Table 3). The effective data size is the number of statistically independent data points. It determines the size of the estimation error; the smaller the effective data size, the larger is the estimation error. In the case of AR(1) serial dependence, the effective data size is less than the sample size; the larger the AR(1) persistence time, the smaller is the effective data size [Mudelsee, 2010, Chapter 2 therein]. Adopting the upper limit of the persistence time thus means calculating with the lower limit of the effective data size, which leads to error bars on the upper limit. It is therefore unlikely that the error bars and the constructed uncertainty bands are too narrow. We call this approach conservative.

Uncertainty band construction of stacks uses pooled resamples, $\{x^i, x^*\}_{i=1}^{n_{\text{res}}}$, on which kernel estimation is repeated. Resampling the oxygen isotope values, that is, generating $x^*(i)$, is done record-wise via a parametric AR(1) persistence model. The algorithm, denoted as autoregressive bootstrap or ARB resampling [Mudelsee, 2010, Chapter 3 therein], is built upon the idea to (1) calculate the white noise residuals, $e(i) = \exp(-\{\text{abs}(-[t(i) - t(i-1)]/r)\} \cdot e(i-1))$, (2) scale them to variance unity by dividing by $(1 - \exp(-2[t(i)-t(i-1)]/r))^{1/2}$, (3) resample point-by-point with replacement from the scaled white noise residuals, and (4) “add the redness” as an inverse of step (1).

Resampling the time values, that is, generating $t^*(i)$, is done record-wise via a simple parametric timescale model. We overtake the reported age error of 0.1 Myr (section 2.3), plug it as standard deviation into a Gaussian (normal) random number generator, and shift by that random amount all time points of a record simultaneously. Different records, and different copies of a record’s resample, have independent timescale errors, but one resample of a record has completely dependent timescale errors. This solution, dictated by the absence of more advanced timescale models from, for example, Bayesian methods [Buck and Millard,
Figure 4. Stack construction, bootstrap resampling. Kernel regression is fitted (black line) to the data points \((t, x)\), which are shown for two hypothetical records, A (blue symbols) and B (red symbols). ARB resampling is applied to the residuals (vertical lines connecting data and fit), yielding a first version of the resamples \((t^*, x^*)\). Timescale simulation is applied to all values of a record simultaneously but independent between records, yielding the final version of the resamples \((t^{**}, x^{**})\).

2004 or frequentist tools used in speleothem dating [Scholz and Hoffmann, 2011], is a conservative uncertainty approach since the simultaneous, completely dependent time shift should generate a higher timescale variability for the time points of the pools compared to using less dependent errors from more advanced models.

The procedure of record-wise ARB resampling with timescale errors (Figure 4) and reestimating the nonparametric kernel regression on the resamples (adopting each time the optimized bandwidths from Figure 3) is repeated until \(B = 400\) copies of simulated nonparametric trends are available. For each of the time points \(T\) (discretized over the 4–61 Ma interval with a spacing of 1 kyr), the standard deviation over the \(B\) copies is determined. The resulting point-wise, standard error uncertainty band is, owing to the twofold conservative approach taken, very likely not an underestimation of the full uncertainties (measurement, proxy, and dating) influencing the estimation.

4. Results and Discussion

4.1. Cenozoic Climate Transitions and Events

4.1.1. Paleocene-Eocene

Although the Cenozoic witnessed to first order a cooling, the transition from a greenhouse to an icehouse climate, its earliest phase saw a warming trend. This began in the middle of the Paleocene and culminated in the Early Eocene Climatic Optimum (EECO). This climatic warming has been recognized in previous work [Miller et al., 1987; Shackleton et al., 1984; Kennett and Stott, 1990]; we call it Paleocene-Eocene Trend (PE-Trend). Superimposed on the PE-Trend were short-term warmings termed hyperthermals [Zachos et al., 2008; Sexton et al., 2011]. The most prominent of those events was the PETM [Kennett and Stott, 1991].

4.1.1.1. Climate Transition PE-Trend

Seven records allow quantification of the PE-Trend transition (Figure 5 and Table 4). The warming set in \(\sim 57.5\) Ma. Within error bars low and high latitudes were coeval. The end was \(\sim 54.5\) Ma (low latitudes) and \(\sim 53.5\) Ma (high latitudes), but those two estimates do not strongly deviate statistically from each other. The \(\delta^{18}O\) amplitude was 0.6 to 0.7‰, indistinguishable for low and high (southern) latitudes.

The decrease in benthic \(\delta^{18}O\) should be interpreted as a warming (of bottom waters) since for that time the existing ice volume (and its changes) was negligible. How much did it warm?

We reestimated the calibration ratio between temperature and \(\delta^{18}O\) changes from the classic paper by Epstein et al. [1953]. We analyzed the laboratory-generated data given in
late hereinafter the temperatures with a larger uncertainty, 4.3 ± 0.1°C per %. In addition to the statistical bootstrap uncertainty, there is systematic uncertainty stemming from violations of (1) the assumed linear form (Epstein et al. [1953] adopted a parabolic form and determined a small second-order term) and (2) the actualism that must inevitably be assumed when applying calibration formula to paleoclimatic problems. We therefore conservatively calculate hereinafter the temperatures with a larger uncertainty, 4.3 ± 0.4°C per %. This value accommodates also the ratio of 3.9°C per %, which Zachos et al. [2001a] employed for the ice-free ocean (Figure 1), the ratio of 4.5°C per %, which Barras et al. [2010] determined on cultured benthic foraminiferal calcite, and the ratio of 4.6°C per %, which Marchitto et al. [2014, equations (5) to (7) therein], presented based on core top benthic foraminifera measurements.

Adopting the above calibration transforms the δ18O decrease into a warming of 2.9 ± 0.4°C. This gradual warming during PE-Trend led to the EECO [Zachos et al., 2001a], the warmest longer phase during the entire Cenozoic. The warming did not change the equator-South Pole bottom water temperature gradient (no data are available on the equator-North Pole gradient at that time). Kennett and Stott [1990] obtained a larger PE-Trend estimate (~5°C warming) on the basis of data from ODP 690. The deviation of their result from ours (Table 4) might be ascribed in part to differences in age models used, but it also likely reflects the existence of a hiatus in the ODP 690 record ~52 Ma and the scatter of between-record results (larger external errors).

### 4.1.1.2. Climate Event PETM

Eight benthic δ18O records have high enough temporal resolution to allow us to quantify the PETM (Figure 6 and Table 5). We statistically model the PETM as a peak event, consisting of an earlier warming start, which peaked at a certain time, and a cooling trend; the warming and cooling changes need not necessarily have the same amplitude (Figure 6).

The five high-latitude records, four from the Southern Hemisphere (DSDP 525, DSDP 527, ODP 689, and ODP 690) and one from the Northern Hemisphere (ODP 1051), show a remarkable agreement in amplitudes: a warming of 1.21 ± 0.11‰ or 5.2 ± 0.7°C and a cooling of 0.96 ± 0.06‰ or 4.1 ± 0.5°C. They also agree in peak timing (55.76 Ma), although this agreement may be partly due to bringing them onto the common timescale [Gradstein et al., 2004]. The systematic error for each of these estimates is not considerably larger than the statistical error (Table 5). Two of the three low-latitude records (DSDP 577 and ODP 865) agree well with the high-latitude results regarding peak timing; the third low-latitude record (ODP 1209) has a slightly later age estimate for the PETM peak (56.7 Ma). This likely reflects a combination of the lower temporal resolution of the ODP 1209 record (0.24 Myr at around the time of the PETM) and the ODP 1209 age model lacking magnetostratigraphy [Dutton et al., 2005].

On the other hand, the three low-latitude records exhibit on average smaller temperature amplitudes than their high-latitude counterparts (a warming of 0.54 ± 0.17‰ or 2.3 ± 0.8°C followed by a cooling of 0.46 ± 0.12‰ or 2.0 ± 0.6°C). The deviating result from DSDP 577 has little weight due to large estimation uncertainty.

---

### Table 4. Resultsa: Climate Transition PE-Trend (cf. Figure 5)

<table>
<thead>
<tr>
<th>Record</th>
<th>Start (Ma)</th>
<th>End (Ma)</th>
<th>Duration (Myr)</th>
<th>Midpoint (Ma)</th>
<th>δ18O Amplitudeb (%o)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Low-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DSDP 577</td>
<td>59.47 ± 0.78</td>
<td>52.95 ± 0.93</td>
<td>6.52 ± 1.40</td>
<td>56.21 ± 0.50</td>
<td>+0.92 ± 0.10</td>
</tr>
<tr>
<td>ODP 865</td>
<td>56.18 ± 0.62</td>
<td>55.31 ± 0.49</td>
<td>0.87 ± 1.01</td>
<td>55.74 ± 0.23</td>
<td>+0.43 ± 0.07</td>
</tr>
<tr>
<td>ODP 1209</td>
<td>58.16 ± 1.35</td>
<td>53.88 ± 1.07</td>
<td>4.28 ± 2.11</td>
<td>56.02 ± 0.61</td>
<td>+0.76 ± 0.12</td>
</tr>
<tr>
<td>Averagec</td>
<td>57.54±0.18</td>
<td>54.65±0.68</td>
<td>3.01±2.00</td>
<td>55.86±0.13</td>
<td>+0.63±0.15</td>
</tr>
<tr>
<td><strong>High-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ODP 690</td>
<td>58.09 ± 0.34</td>
<td>53.89 ± 0.27</td>
<td>4.20 ± 0.48</td>
<td>55.99 ± 0.20</td>
<td>+0.79 ± 0.08</td>
</tr>
<tr>
<td>ODP 698</td>
<td>56.08 ± 0.84</td>
<td>51.62 ± 0.91</td>
<td>4.46 ± 1.46</td>
<td>53.85 ± 0.48</td>
<td>+0.90 ± 0.13</td>
</tr>
<tr>
<td>ODP 702</td>
<td>57.27 ± 1.14</td>
<td>51.31 ± 1.42</td>
<td>5.96 ± 2.18</td>
<td>54.29 ± 0.69</td>
<td>+0.59 ± 0.08</td>
</tr>
<tr>
<td>ODP 738</td>
<td>54.99 ± 0.81</td>
<td>53.23 ± 0.59</td>
<td>1.76 ± 1.27</td>
<td>54.11 ± 0.31</td>
<td>+0.61 ± 0.07</td>
</tr>
<tr>
<td>Averagec</td>
<td>57.39±0.66</td>
<td>53.53±0.42</td>
<td>4.02±2.00</td>
<td>55.15±0.55</td>
<td>+0.68±0.06</td>
</tr>
</tbody>
</table>

aValues are rounded.
bNegative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.
cWeighted average with external error (superscript) and internal error (subscript); start, end, and midpoint averages include dating-error effects.
Previous work by others on the timing of the PETM peak reflects also the preference for certain age models [Cronin, 2010]. Published dates include 57.33 Ma [Kennett and Stott, 1991] (who, however, acknowledged in their paper that this estimate would be revised) and 54.95 Ma [Zachos et al., 2001a]. A recent chronology [Westerhold et al., 2007], based on countable eccentricity cycles of 405 kyr period in deep ocean sedimentary records from the Walvis Ridge, suggests a PETM peak timing of either 55.53 or 55.93 Ma (depending on the currently undecided counting solution)—our estimate of 55.76 Ma would perfectly fit into the middle.

Previous work by others on the amplitude of the PETM deep water temperature signal can be compared with our results (Table 5). Kennett and Stott [1991] analyzed the same isotope data (ODP 690) as us and found an amplitude of around 2‰ or 8.6°C, which we think is too high. Kennett and Stott [1991] further noted that the cooling amplitude was smaller by ~1 to 2°C (equivalent to 0.2 to 0.3‰) than the warming amplitude of the PETM, with which we agree. Zachos et al. [2001a, 2008] analyzed δ¹⁸O records from DSDP 525, DSDP 527, ODP 690, and ODP 865 and found a PETM amplitude of more than 5°C. Tripati and Elderfield [2005] measured via Mg/Ca paleothermometry the bottom water temperatures across the PETM from Sites DSDP 527, ODP 865, and ODP 1209, using three different foraminifera genera (yielding different estimates). Their

![Figure 6](image-url)

**Figure 6.** Results, climate event PETM ((a) low latitudes; (b) high latitudes); cf. Figure 5. This climate event was described by a combination of regression models (ramp-ramp for DSDP 527, DSDP 577, ODP 690, ODP 1051, and ODP 1209, ramp-break-ramp for DSDP 525 and ODP 689, and break-break-break for ODP 865). See Table 5 for numerical results. Note that for DSDP 577 and ODP 1209, the fit range is larger than the shown time axis range.

<table>
<thead>
<tr>
<th>Table 5. Results*: Climate Event PETM (cf. Figure 6)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Record</strong></td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td><strong>Low-Latitude Records</strong></td>
</tr>
<tr>
<td>DSDP 577</td>
</tr>
<tr>
<td>ODP 865</td>
</tr>
<tr>
<td>ODP 1209</td>
</tr>
<tr>
<td>Average c</td>
</tr>
<tr>
<td><strong>High-Latitude Records</strong></td>
</tr>
<tr>
<td>DSDP 525</td>
</tr>
<tr>
<td>DSDP 527</td>
</tr>
<tr>
<td>ODP 689</td>
</tr>
<tr>
<td>ODP 690</td>
</tr>
<tr>
<td>ODP 1051</td>
</tr>
<tr>
<td>Average c</td>
</tr>
</tbody>
</table>

*aValues are rounded.*

*bNegative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.*

*cWeighted average with external error (superscript) and internal error (subscript); time averages include dating-error effects.*
finding of a PETM amplitude of 4 to 5°C warming agrees with our finding (Table 5) for the high latitudes. They also detected a PETM warming of similar magnitude for the low-latitude site of ODP 865, while we detect via δ¹⁸O an amplitude of 0.87 ± 0.25‰ or 3.7 ± 1.1°C—which is very compatible. However, for ODP 1209 we find a PETM warming of only 0.48 ± 0.04‰ or 2.1 ± 0.3°C—which is clearly smaller than 4 to 5°C. An explanation could be that the lower temporal resolution of the ODP 1209 δ¹⁸O record at the PETM (Figure 6) did not sample the extreme PETM values.

Regarding polar temperature amplification, it is mathematically possible to calculate a polar amplification factor of deep water amplitudes from the statistical results (Table 5). The warming at the beginning of the PETM yields a factor of

\[
\frac{1.21 ± 0.11}{0.54 ± 0.17} ≈ 2.2 ± 0.7,
\]

and the cooling at the end of the PETM yields a factor of

\[
\frac{0.96 ± 0.06}{0.46 ± 0.12} ≈ 2.1 ± 0.6,
\]

which is indistinguishable from the factor for the beginning (note that in a conservative approach we have used the larger systematic error bars). This calculation suggests that amplification did occur. However, if the result from ODP 1209 is ignored (because of too low resolution and missed extremes), then there is no evidence for polar deep water temperature amplification. In addition, the “climatological uncertainties” of the estimated amplification factors are likely larger than the statistical uncertainties since the representativeness of the selected low- and high-latitude records for the respective geographical regions is limited.

The duration of the warming phase of the PETM, and also of its cooling or “recovery” phase, is an important climate-dynamical parameter. Since the duration may be rather short, as we shall see, the PETM parameters amplitude and duration may contain information about fast climate feedback and short-term climate sensitivity, which could help to put the current anthropogenically induced greenhouse gas emissions into a quantitative climatic context [Sexton et al., 2011; DeConto et al., 2012; PALAEONSENS Project Members, 2012; Masson-Delmotte et al., 2013; Zeebe and Zachos, 2013]. Since different compartments in the climate system are characterized by their specific response timescales, we focus here on previous literature that describes changes in the temperature of the deep ocean.

Our results on the durations of the PETM warming and cooling phases (Table 5) are—rightly—dominated by the high-accuracy estimates from ODP 690 and ODP 1051: the warming was accomplished within 6 ± 3 kyr, and the cooling was accomplished within 25 ± 12 kyr (conservative error bounds). These high-accuracy results are owing to relatively high average temporal resolutions of these records around the PETM (ODP 690, 6.7 kyr; ODP 1051, 3.5 kyr). The next coarser resolved series (DSDP 527, 15 kyr) still shows relatively short durations of 36 kyr (warming) and 30 kyr (cooling). The short-duration estimates of the initial warming phase, all from high-latitude records, are in agreement with previous estimates, obtained partly on identical records under different timescales and from per-eye inspection [Kennett and Stott, 1991; Zachos et al., 2001a, 2008; Cronin, 2010]. At face value, our estimate of a short duration of also the second, cooling phase of the PETM seems to disagree with a previous, detailed study [Röhl et al., 2007] on ODP 690 and IODP 1263, finding a duration of the whole PETM of ~170 kyr. However, the following points may help to reconcile this apparent disagreement.

1. The study by Röhl et al. [2007] was based on precession-cycle counting of the Ba elemental records and defining the PETM in the conventional way, via carbon isotopes (δ¹³C).
2. As Röhl et al. [2007, p. 6 therein] noted, the “location of the termination of the recovery phase [called cooling phase by us] is somewhat subjective because of the asymptotic shape of the carbon isotope excursion.” Our adopted regression models (section 3.1) do explicitly allow for a termination of the recovery phase of the PETM at a warmer level than before the PETM—which we think is realistic. Recovery to the identical level, if one wishes to adopt such a definition, would have taken longer.

Extraterrestrial ³He-based timescales for the PETM sections in marine sedimentary records [Farley and Eltgroth, 2003; Murphy et al., 2010] provide additional information. To study the influence on estimated PETM parameters of the selection of the timescale, we brought the ODP 690 and ODP 1051 δ¹⁸O records [Cramer et al., 2009] onto the ³He-based timescales [Farley and Eltgroth, 2003] by means of linear interpolation and utilizing the sediment-depth points. The ³He-based timescales are relative to the timing of the
PETM peak; hence, we studied only durations and amplitudes. The results (not shown) attest to the robustness of estimates from ODP 1051; all entries (Table 5) are only minimally affected. Also, both amplitude estimates (warming and cooling) from ODP 690 are robust. On the other hand, the duration estimates from ODP 690 (Table 5) should be interpreted with caution. While the value for the end (cooling phase of the PETM) changed from $35 \pm 16$ kyr (Table 5) to $16 \pm 11$ kyr ($^4$He-based), which is still compatible with the summary estimate of a few tens of kiloyears duration, the value for the start (warming phase) changed from $14 \pm 4$ kyr (Table 5) to $33 \pm 8$ kyr ($^4$He-based). However, still valid are the conclusions that (1) both warming and cooling phases of the PETM were relatively fast (i.e., within a few of tens of kiloyears) and (2) the amplitude of the warming was larger than that of the cooling.

One may ask whether the warmer PETM recovery temperature is due to the long-term background warming trend (PE-Trend). The $\delta^{18}O$ slope of the PE-Trend is (weighted average of the entries in Table 4)

$$(0.67 \pm 0.05)%/ (3.95 \pm 0.50 \text{ Myr}) \approx 0.17 \pm 0.02/\text{Myr}.$$
Médoc, along a longer-term “event” within the interval from 39 to 44 Ma. The Médoc was geographically rather heterogeneous:

The apparently nonsynchronized end of the first cooling phase LTEC-I (section 4.1.2.1) was followed by the Médoc, a longer-term “event” within the interval from 39 to 44 Ma. The Médoc was geographically rather heterogeneous: some records (e.g., ODP 689 and ODP 748) display a strong $\delta^{18}O$ minimum, while other records do not (Figures 7 and 8). (Bohaty and Zachos [2003] report warming amplitudes equivalent to 1‰ $\delta^{18}O$ for a compilation of records: ODP 689, ODP 690, ODP 738, ODP 744, and ODP 748.) Cronin [2010, pp. 105–106 therein] relates the Médoc heterogeneity to heterogeneous changes in calcium-carbonate compensation depth (CCD) and productivity [van Andel, 1975; see also more recent quantifications, Bohaty and Zachos, 2003; Coxall et al., 2005; Lyle et al., 2005; Pälike et al., 2012]. Information from high-resolution $\delta^{18}O$ records on the end of the Médoc and the second cooling phase LTEC-II is more sparse (five records) than for LTEC-I. However, it is certain that both cooling phases were long-term, over several million years [Ehrmann et al., 1992; Kennett and Stott, 1990; Zachos et al., 2001a]. It is interesting to note that the $\delta^{18}O$ amplitude of the LTEC-II transition was significantly smaller for the low-latitude record ODP 1218, which has 0.35‰, than for each of the four high-latitude records (Figure 8 and Table 7), which average 0.68‰. Although this is just a

---

**Table 6. Results**: Climate Transition LTEC-I (cf. Figure 7)

<table>
<thead>
<tr>
<th>Record</th>
<th>Start (Ma)</th>
<th>End (Ma)</th>
<th>Duration (Myr)</th>
<th>Midpoint (Ma)</th>
<th>$\delta^{18}O$ Amplitude ($%$)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Low-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ODP 865</td>
<td>49.17 ± 0.68</td>
<td>44.44 ± 0.53</td>
<td>4.73 ± 1.02</td>
<td>46.81 ± 0.34</td>
<td>-0.86 ± 0.05</td>
</tr>
<tr>
<td>ODP 1209</td>
<td>47.81 ± 1.39</td>
<td>42.51 ± 1.49</td>
<td>5.29 ± 2.47</td>
<td>45.16 ± 0.74</td>
<td>-1.05 ± 0.21</td>
</tr>
<tr>
<td>IODP 1258</td>
<td>49.08 ± 0.29</td>
<td>44.94 ± 0.37</td>
<td>4.15 ± 0.54</td>
<td>47.01 ± 0.19</td>
<td>-1.06 ± 0.08</td>
</tr>
<tr>
<td>IODP 1260</td>
<td>47.06 ± 0.90</td>
<td>43.76 ± 0.75</td>
<td>3.30 ± 1.48</td>
<td>45.41 ± 0.37</td>
<td>-0.47 ± 0.07</td>
</tr>
<tr>
<td><strong>Average</strong></td>
<td>48.88 ± 0.35</td>
<td>44.55 ± 0.33</td>
<td>4.22 ± 0.24</td>
<td>46.59 ± 0.39</td>
<td>-0.82 ± 0.12</td>
</tr>
<tr>
<td><strong>High-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DSDP 401</td>
<td>49.20 ± 0.26</td>
<td>48.50 ± 0.25</td>
<td>0.71 ± 0.46</td>
<td>48.85 ± 0.11</td>
<td>-0.97 ± 0.10</td>
</tr>
<tr>
<td>DSDP 527</td>
<td>47.07 ± 0.57</td>
<td>44.26 ± 0.55</td>
<td>2.61 ± 0.92</td>
<td>45.66 ± 0.31</td>
<td>-0.58 ± 0.11</td>
</tr>
<tr>
<td>ODP 690</td>
<td>49.31 ± 0.75</td>
<td>45.58 ± 0.81</td>
<td>3.73 ± 1.34</td>
<td>47.44 ± 0.40</td>
<td>-0.54 ± 0.08</td>
</tr>
<tr>
<td>ODP 702</td>
<td>50.49 ± 0.51</td>
<td>44.50 ± 0.31</td>
<td>5.99 ± 0.67</td>
<td>47.50 ± 0.26</td>
<td>-0.97 ± 0.07</td>
</tr>
<tr>
<td>ODP 738</td>
<td>49.77 ± 0.94</td>
<td>41.02 ± 0.67</td>
<td>8.75 ± 1.25</td>
<td>45.39 ± 0.52</td>
<td>-1.18 ± 0.11</td>
</tr>
<tr>
<td><strong>Average</strong></td>
<td>49.17 ± 0.47</td>
<td>46.08 ± 1.19</td>
<td>2.96 ± 1.33</td>
<td>47.96 ± 0.60</td>
<td>-0.84 ± 0.12</td>
</tr>
</tbody>
</table>

*a* Values are rounded.

*b* Negative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.

*c* Weighted average with external error (superscript) and internal error (subscript); start, end, and midpoint averages include dating-error effects.
single record, ODP 1218 covers the full fit interval rather homogeneously at a high temporal resolution, and this low-latitude record agrees in change-point times excellently with the weighted averages from the four high-latitude records. We thus conclude that ODP 1218 gives rather reliable estimates. Under the assumption of still negligible ice volume changes [Zachos et al., 2001a], it follows that the LTEC-II transition may have been associated with a changing deep ocean circulation pattern or latitudinal temperature gradients.

### 4.1.3. Eocene-Oligocene

After the long-term Eocene cooling phases, which ended ∼38 Ma (LTEC-II, section 4.1.2.2), the global climate system seems to have remained relatively stable for several million years until the EOT. The EOT spans the Eocene-Oligocene boundary at ∼34 Ma and is marked by a rather abrupt transition toward heavier $\delta^{18}O$. This shift is widely interpreted as reflecting the glaciation of Antarctica [Miller et al., 1987, 1991; Prothero et al., 2003]. This interpretation is ultimately supported by direct geological [Ivany et al., 2006] and sedimentological [Barrera and Huber, 1991; Ehrmann et al., 1992; Zachos et al., 1996] evidence. With the appearance of significant ice, the partition problem of the $\delta^{18}O$ amplitudes (temperature versus ice volume signal) intensifies. We estimate the EOT glaciation signals on the basis of 12 records and also quantify the “overshoot” behavior at the end of the transition, previously termed the Eocene-Oligocene Glacial Maximum or EOGM [Zachos et al., 1996].

#### 4.1.3.1. EOT

Our estimates for the timing of the EOT start (Figure 9 and Table 8) show some agreement (within systematic error bars) between low and high latitudes, with a combined weighted average of $34.04 \pm 0.09$ Ma. The result from ODP

---

**Table 7. Results a: Climate Transition LTEC-II (cf. Figure 8)**

<table>
<thead>
<tr>
<th>Record</th>
<th>Start (Ma)</th>
<th>End (Ma)</th>
<th>Duration (Myr)</th>
<th>Midpoint (Ma)</th>
<th>$\delta^{18}O$ Amplitudeb (‰)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Low-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ODP 1218</td>
<td>38.89 ± 0.37</td>
<td>38.23 ± 0.36</td>
<td>0.66 ± 0.67</td>
<td>38.56 ± 0.15</td>
<td>−0.35 ± 0.04</td>
</tr>
<tr>
<td><strong>High-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ODP 689</td>
<td>39.55 ± 0.20</td>
<td>38.45 ± 0.26</td>
<td>1.10 ± 0.41</td>
<td>39.00 ± 0.11</td>
<td>−0.74 ± 0.06</td>
</tr>
<tr>
<td>ODP 702</td>
<td>39.05 ± 0.18</td>
<td>37.51 ± 0.29</td>
<td>1.54 ± 0.36</td>
<td>38.28 ± 0.16</td>
<td>−0.63 ± 0.04</td>
</tr>
<tr>
<td>ODP 738</td>
<td>39.70 ± 0.87</td>
<td>37.10 ± 0.87</td>
<td>2.60 ± 1.47</td>
<td>38.40 ± 0.46</td>
<td>−0.68 ± 0.12</td>
</tr>
<tr>
<td>ODP 748</td>
<td>39.58 ± 0.30</td>
<td>37.71 ± 0.54</td>
<td>1.87 ± 0.72</td>
<td>38.65 ± 0.25</td>
<td>−1.04 ± 0.16</td>
</tr>
<tr>
<td>Average</td>
<td>39.34±0.15</td>
<td>38.00±0.22</td>
<td>1.44±0.19</td>
<td>38.70±0.19</td>
<td>−0.68±0.05</td>
</tr>
</tbody>
</table>

---

*a* Values are rounded.

*b* Negative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.

*Weighted average with external error (superscript) and internal error (subscript); start, end, and midpoint averages include dating-error effects.

---

**Figure 9.** Results, Eocene-Oligocene Transition (EOT) ((a) low latitudes; (b) high latitudes); cf. Figure 5. For ODP 1218, the transition is statistically modeled as a two-step change [Coxall et al., 2005]. Arrows indicate overshoot behavior. See Table 8 for numerical results.
Table 8. Results*: Eocene-Oligocene Transition (EOT) (cf. Figure 9)

<table>
<thead>
<tr>
<th>Record</th>
<th>Start (Ma)</th>
<th>End (Ma)</th>
<th>Duration (Myr)</th>
<th>Midpoint (Ma)</th>
<th>$\delta^{18}O$ Amplitudeb (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSDP 77</td>
<td>33.99 ± 0.05</td>
<td>33.84 ± 0.09</td>
<td>0.14 ± 0.11</td>
<td>33.92 ± 0.04</td>
<td>−0.75 ± 0.12</td>
</tr>
<tr>
<td>DSDP 566</td>
<td>34.13 ± 0.16</td>
<td>33.82 ± 0.14</td>
<td>0.30 ± 0.28</td>
<td>33.97 ± 0.06</td>
<td>−0.84 ± 0.07</td>
</tr>
<tr>
<td>DSDP 574</td>
<td>33.69 ± 0.34</td>
<td>33.61 ± 0.49</td>
<td>0.08 ± 0.57</td>
<td>33.65 ± 0.31</td>
<td>−0.48 ± 0.16</td>
</tr>
<tr>
<td>ODP 803</td>
<td>34.45 ± 0.64</td>
<td>33.76 ± 0.25</td>
<td>0.69 ± 0.25</td>
<td>34.11 ± 0.12</td>
<td>−1.11 ± 0.18</td>
</tr>
<tr>
<td>ODP 1218c</td>
<td>33.93 ± 0.06</td>
<td>33.65 ± 0.02</td>
<td>0.28 ± 0.06</td>
<td>33.79 ± 0.03</td>
<td>−0.96 ± 0.04</td>
</tr>
<tr>
<td>Averaged</td>
<td>34.19±0.12</td>
<td>33.66±0.02</td>
<td>0.27±0.05</td>
<td>33.86±0.04</td>
<td>−0.91±0.06</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Low-Latitude Records</th>
<th>High-Latitude Records</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSDP 522</td>
<td>34.33 ± 0.08</td>
</tr>
<tr>
<td>DSDP 529</td>
<td>33.93 ± 0.15</td>
</tr>
<tr>
<td>DSDP 549</td>
<td>34.15 ± 0.15</td>
</tr>
<tr>
<td>ODP 689</td>
<td>33.85 ± 0.05</td>
</tr>
<tr>
<td>ODP 703</td>
<td>34.13 ± 0.06</td>
</tr>
<tr>
<td>ODP 744</td>
<td>33.91 ± 0.03</td>
</tr>
<tr>
<td>ODP 748</td>
<td>33.78 ± 0.04</td>
</tr>
<tr>
<td>Averaged</td>
<td>33.99±0.07</td>
</tr>
</tbody>
</table>

aValues are rounded.
bNegative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.
cFor ODP 1218, the transition is statistically modeled as a two-step change [Coxall et al., 2005]; the earlier step is estimated as from (33.93 ± 0.06 Ma, 0.82 ± 0.03‰) to (33.90 ± 0.02 Ma, 1.22 ± 0.02‰) and the later step as from (33.70 ± 0.03 Ma, 1.20 ± 0.06‰) to (33.65 ± 0.02 Ma, 1.78 ± 0.03‰).
dWeighted average with external error (superscript) and internal error (subscript); start, end, and midpoint averages include dating-error effects.

803, western equatorial Pacific, deviates. Rather than a regional climatological signal, this likely reflects the less-than-optimal statistical conditions (only short coverage of the earlier part of the transition, see Figure 9). The end of the isotope shift is around 33.67 ± 0.03 Ma (combined weighted average from roughly synchronous low and high latitudes).

Our estimates for the duration of the EOT isotope shift (Table 8) are around 0.2 to 0.3 Myr, in excellent agreement with Coxall et al. [2005]. (This range is also compatible with a calculation of the duration via \[34.04 ± 0.09\] Ma − [33.67 ± 0.03] Ma = [0.37 ± 0.09] Myr.) Some systematic uncertainties are apparent. These may stem from the less-than-optimal functional form of the ramp regression model (Figure 9). Some of the series, especially the high-resolution records ODP 744 and ODP 1218, but also others such as DSDP 529, DSDP 574, ODP 689, and ODP 748, show the clear overshoot behavior at the EOT end, where $\delta^{18}O$ does not remain constant but turns slightly back and recovers at lighter values (indicated by arrows in Figure 9), marking the end of the EOGM [Zachos et al., 1996]. Owing to the unavailability of an objective statistical regression model for the overshoot, the eye may be better at finding the overshoot present in other records. (At least two additional parameters, one for the size, the other for the duration of the overshoot would need to be invoked, making fitting rather difficult. We attempt to quantify the overshoot for the high-resolution record ODP 1218 in a subsequent paragraph.)

The comparison of our time estimates with previous estimates from the literature is more fruitful for the EOT duration than for its start or end timings since the latter reflect also the preference by researchers for adopting a certain geologic timescale. Barrera and Huber [1991] see a duration of 0.6 Myr in the benthic $\delta^{18}O$ record from ODP 744, which we think (see the individual estimation result in Table 8) is an overestimation, perhaps caused by an influence of the overshoot. Zachos et al. [1996] examine benthic $\delta^{18}O$ from ODP 744 and DSDP 522 and find a shorter duration, 0.35 Myr. The ODP 1218 $\delta^{18}O$ record, from the eastern equatorial Pacific, provides excellent statistical inference conditions (7 kyr resolution). The EOT in that record has been described as comprising two distinct steps, each of a duration of 0.04 Myr [Coxall et al., 2005; Coxall and Wilson, 2011]. Our fitting of two ramps to that record (Table 8) agrees well with their result. Previously, based on records from DSDP 522 and ODP 744, Zachos et al. [1996, p. 251 therein] reported that “more than half of the EOT isotope shift occurred over the final 40–50 kyr [from 350 kyr].”—a view that is compatible with a two-step change. Bohaty et al. [2012] also identify the two isotope steps in their Southern Ocean $\delta^{18}O$ records. One can easily fit two ramps instead of one to a smooth transition (Figure 9).
Coxall et al. [2005] further identified in the same record, ODP 1218, the overshoot behavior associated with the end of the EOGM. This feature was noted also by Pälike et al. [2006], who gave a duration of the recovery after the overshoot of 0.4 to 0.8 Ma. We fitted a ramp to the ODP 1218 $\delta^{18}$O record, regression interval from 32.50 to 33.65 Ma, in order to quantify the recovery from the overshoot. It turned out (results not shown), confirming Pälike et al. [2006], that the recovery was achieved within $0.5 \pm 0.2$ Myr, with an amplitude of $+0.26 \pm 0.04\%$ (deglaciation/warming). To summarize the findings on the timing of the EOT, (1) it was a fast transition (0.2 to 0.3 Myr duration), (2) likely in at least two faster steps, as seen in the high-resolution ODP 1218 record, (3) it was centered $\sim 33.86$ Ma (Table 8), and (4) it included at the end an overshoot behavior with a recovery to less glaciated and/or warmer conditions of $0.26\%$ $\delta^{18}$O amplitude within 0.5 Myr, a phenomenon that is quantified here for the ODP 1218 record but which likely has a larger regional, even global scale.

In a recent paper, Westerhold et al. [2014] developed an astronomically tuned timescale for the middle Eocene to early Oligocene and determined the Eocene–Oligocene boundary to be at 33.89 Ma, which is in close agreement with our EOT midpoint estimate of 33.86 ± 0.04 Ma (Table 8; low and high latitudes, external error).

Our estimates for the benthic $\delta^{18}$O amplitude of the EOT (Table 8) show an excellent agreement between low (0.91 ± 0.06%) and high (0.98 ± 0.07%) latitudes. Averaging all individual values, the overall glaciation/cooling is found to be 0.94% with a systematic error of 0.05% and a statistical error of 0.02%. Our estimate for the amplitude is similar whether we assume the isotopeshift was achieved in one or in two steps. A subsequent recovery from that “glacial overshoot” seems to be a phenomenon of a large regional, perhaps even global scale.

Previous amplitude estimates, based on (1) analyzing subsets of the same $\delta^{18}$O database as ours (section 2), (2) using the same Gradstein et al., 2004 or (the earlier papers) slightly other timescales, and (3) quantifying the amplitudes in the curves mostly per eye, are mainly comparable to our results. Barrera and Huber [1991] find a glaciation/cooling of 1.15% on the ODP 744 record, an almost perfect agreement with our individual estimate (Table 8). In their review paper on the Cenozoic Antarctic cryosphere evolution, Shevenell andKennett [2007] see an overall amplitude of $\sim 1\%$. Billups and Schrag [2003] examine paired proxies (benthic $\delta^{18}$O and Mg/Ca ratios) on records from ODP 689 and ODP 757, finding no evidence for temperature changes across the EOT and, hence, assessing the $\delta^{18}$O amplitude of $\sim 1\%$ as ice volume related. Only Tripati et al. [2005, p. 341 therein] overestimate in our opinion the EOT amplitude when they constitute a benthic $\delta^{18}$O amplitude of “up to” 1.5%. We think this overestimation stems from putting too much emphasis on the extremes (i.e., the warm extremes at the EOT start and the cold extremes at the EOT end) rather than on the mean trend (as the regression does it).

Before turning to the application of Mg/Ca paleothermometry with respect to the EOT, let us consider the simple interpretation of the observed $\delta^{18}$O amplitude as a pure ice volume signal. First, current Antarctic ice volume corresponds to a sea level change of 58 m [Fretwell et al., 2013]. For the EOT, there exists an independent estimate of the ice volume change based on sequence stratigraphy from the coastal area of New Jersey. Pekar et al. [2002] found that the apparent sea level during the earliest Oligocene fell by 80 ± 15 m. Since the temporal resolution of such studies is inevitably rather coarse, one has to compare that sea level amplitude with the EOT amplitude in marine $\delta^{18}$O that was attained following the EOGM in the recovery state. Our best quantitative estimate for this comes from marine record ODP 1218, where the initial (two-step) glaciation with an amplitude of 0.96 ± 0.04% (Table 8) was followed by a recovery deglaciation amplitude of 0.26 ± 0.04%. The net amplitude for longer-term ice volume changes (glaciation of Antarctica) would be 0.70 ± 0.06%. Adopting a relation of 1.1% per 100 m between changes in $\delta^{18}$O and sea level [Fairbanks and Matthews, 1978; de Boer et al., 2012], the net amplitude in oxygen isotopes would correspond to 64 ± 5 m sea level change, in good agreement with the New Jersey observation. However, the major source of uncertainty in bringing $\delta^{18}$O and ice volume together in this manner is the $\delta^{18}$O-sea level relation, which Fairbanks andMatthews [1978] established for a considerably different climatological-geographical situation, the Pleistocene. Since the assumed relation is violated to some degree when applied to the earliest Oligocene, for example, because of the theoretical nonlinear functional form [Mix and Ruddiman, 1984], the true error bars would be a little larger, and the “excellent agreement” would be somewhat spurious; “some,” “little,” “somewhat”: model results [de Boer et al., 2012] show that the violation of the assumed actualism is not strong.
From a physics viewpoint, the true $\delta^{18}O$ signal stored during ice buildup on Antarctica should depend on the travel distance of the precipitation (Rayleigh distillation) and the source values [Oeschger and Langway, 1989]. Uncertainties in the knowledge about past travel distances and source regions propagate ultimately into the uncertainty about the signal partitioning (ice volume versus temperature). Bohaty et al. [2012] offer a quantitative discussion of the relation between ice volume and $\delta^{18}O$ at around the EOT, and Gasson et al. [2012] explore the uncertainties in the relationship between temperature, ice volume, and sea level over the past 50 Myr. Compare also the work by Katz et al. [2008], Lear et al. [2008], and de Boer et al. [2012].

In an early study, Lear et al. [2000] found no evidence for a decrease (cooling) in the benthic foraminiferal Mg/Ca record from DSDP Site 522. This surprising result was later reproduced for ODP Sites 689 and 757 [Billups and Schrag, 2003], ODP Site 1218 [Lear et al., 2004], IODP Site 1263 [Peck et al., 2010], ODP Site 1090 and IODP Site 1265 [Pusz et al., 2011], and ODP Sites 689 and 748 [Bohaty et al., 2012]. Identification of a second carbonate saturation state control on benthic foraminiferal Mg/Ca [Elderfield et al., 2006] has enabled these results to be reconciled with a deep-sea cooling event, as the EOT was also marked by a $\sim 1$ km deepening of the CCD [Lear et al., 2004; Coxall et al., 2005; Lear et al., 2008, 2010; Peck et al., 2010; Pusz et al., 2011; Bohaty et al., 2012]. Regional variations in the CCD deepening likely produced a variable impact on benthic foraminiferal Mg/Ca at different sites. Furthermore, the carbonate saturation state effect on benthic foraminiferal Mg/Ca may be nonlinear, perhaps operating only below a saturation state threshold [Rosenthal et al., 2006], although see the useful discussion in the paper by Elderfield et al. [2006].

The two-step CCD deepening might therefore be expected to produce an offset also in the timing of Mg/Ca signals between sites. Current work using combinations of new calibrations, independent proxies for carbonate saturation state and/or exploiting infaunal benthic foraminifera, shows promise in unraveling the temperature and saturation state controls on benthic foraminiferal Mg/Ca records [Elderfield et al., 2010; Lear et al., 2010; Yu et al., 2010]. However, until this secondary control on benthic foraminiferal Mg/Ca ratios is better quantified, it is not advisable to stack multisite benthic foraminiferal Mg/Ca records from intervals of significant change in carbonate saturation state for quantitative statistical analysis. In the deep-sea realm, the shelf-basin hypothesis implies that this precaution applies to most intervals of major changes in sea level [Berger and Winterer, 1974]. However, the CCD deepening at the EOT apparently did not affect the primary Mg/Ca of shallow-dwelling benthic or planktic foraminifera. The planktic foraminiferal Mg/Ca paleothermometry indicates a $\sim 2$ to $3^\circ$C cooling at both low [Bohaty et al., 2012] latitudes and suggests that approximately 0.6% of the overall EOT $\delta^{18}O$ shift can be ascribed to increased continental ice volume [Lear et al., 2008; Bohaty et al., 2012]. A shelf record of benthic foraminiferal $\delta^{18}O$ and Mg/Ca through an EOT section containing lithologic variations and hiatuses is understandably relatively noisy, yet the long-term shift in the $\delta^{18}O$ of seawater calculated from this record is also not inconsistent with this value [Katz et al., 2008]. This estimate of the change in the $\delta^{18}O$ of seawater across the EOT is also in good agreement with the results of a transient one-dimensional ice sheet model [de Boer et al., 2012].

Regarding physical-climatological causal explanations of the EOT glaciation, the timing of the EOT start, determined by us as 34.04 $\pm$ 0.09 Ma (Table 8), excludes an external astronomical influence in the form of the Popigai impact in Siberia that occurred 35.7 $\pm$ 0.1 Ma [Bottomley et al., 1997]; already, those authors had excluded that connection; this finding is robust also against uncertainties in the geologic timescale. We rather prefer declining levels of atmospheric CO$_2$ [DeConto and Pollard, 2003; Pagani et al., 2011; Egan et al., 2013] or the tectonic explanation via the opening of the Drake Passage and the Tasmanian Seaway and their impacts on Southern Ocean circulation, the thermal isolation of Antarctica, and various feedback links (e.g., ice albedo). This explanation that has chiefly been elaborated by James P. Kennett and his coworkers in a number of papers (see Kennett and Exon [2004], and references cited therein; see also Sijp et al. [2009] for a climate model analysis of the role of these forcing factors).

A model study using a fully coupled atmosphere-ocean-ice model is lacking up to date. By parameterizing the ocean heat transport in a coupled atmosphere-ice model, DeConto and Pollard [2003] concluded that the role of the Drake Passage is rather minor compared to greenhouse gas levels in conjunction with the Earth’s orbital parameters. A continental ice sheet can only be established if the orbital parameters favor cool austral summers. However, once the atmospheric CO$_2$ declines further, the Antarctic ice sheet becomes almost insensitive to the orbital forcing. Cristini et al. [2012] presented a model sensitivity study aimed to understand if and how the opening of the Drake Passage served as a forcing factor for the Antarctic climate transition. A reduced southward heat flux and a decrease of both water and air temperature are found around and over Antarctica when the Drake Passage is open. A more massive ice sheet develops on the
continent, in this case compared to the model configuration with closed Drake Passage. More recently, Wilson et al. [2013] suggested the possibility of substantial ice in the Antarctic interior before the Eocene-Oligocene boundary. As pointed out by these authors, the EOT glaciation likely depends on the distribution of the bedrock topography. Several long-term processes of landscape evolution, including glacial erosion, thermal subsidence, and tectonics, have likely lowered the topography in the West Antarctic region considerably, with Antarctic land area having decreased by approximately 20%. The ice sheet model, on which these reconstructions are based, shows (1) that the West Antarctic Ice Sheet first formed at around the EOT in concert with the continental-scale expansion of the East Antarctic Ice Sheet and (2) that the total volume of East and West Antarctic ice (33.4 to 35.9 million cubic kilometers) was more than 1.4 times greater than previously assumed.

4.1.4. Oligocene

Although the Oligocene seems to have had a relatively stable climate between the two glaciation steps EOT (section 4.1.3.1) and Oligocene-Miocene Boundary or OMB (section 4.1.5.1), we detect and quantify some statistically significant oscillations in benthic δ¹⁸O records (O-Swings).

4.1.4.1. O-Swings

The early part of the Oligocene swings started with a significant δ¹⁸O decrease (Figure 10 and Table 9). The high latitudes (DSDP 522, ODP 689, ODP 744, and ODP 748) exhibit a stronger deglacialiation/warming slope (∼0.46‰ Myr⁻¹) than the one low-latitude record ODP 1218 (∼0.23‰ Myr⁻¹), and that trend seems to have persisted longer for the high latitudes, up to ∼32 Ma. After that time the slopes stayed for a period at around zero, for low as well as high latitudes (Table 9).

Previous studies based on marine benthic δ¹⁸O [Miller et al., 1987; Barrera and Huber, 1991; Ehrmann et al., 1992; Zachos et al., 2001a; Lyle et al., 2008] have also found evidence for a relatively stable Oligocene climate, on Antarctica and also likely on a global scale. Previous studies based on Mg/Ca [Billups and Schrag, 2003; Lear et al., 2004] indicate for some locations that variations in temperature of deep waters (and their feeding surface sources) did exist but were not large.

The later part of the initial Oligocene swing brought a slight glaciation/cooling before ∼27 to 28 Ma and a slight deglaciation/warming thereafter (Figure 10 and Table 10). The associated δ¹⁸O slopes are small and seem not to deviate strongly between the six low-latitude sites and the seven high-latitude sites. Zachos et al. [2001a, p. 688 therein] had previously noted “a warming trend [that] reduced the extent of Antarctic ice” after 26 to 27 Ma.

An estimation of temperature versus ice volume changes across the Oligocene swings on the basis of quantified amplitudes would likely be rather inaccurate due to the small signal sizes and also the paucity of records (especially from low latitudes during the earlier part). However, the swings were to a considerable extent...
degree in concert (Tables 9 and 10), consistent with some contribution from fluctuating ice volumes [Lear et al., 2004]. These fluctuations likely affected the sheet established during the previous EOT on the Antarctic continent, although they were certainly too small for a complete melting. Possibly, also some parts of a minor ice sheet in the Northern Hemisphere grew and fluctuated, as there exists IRD evidence for that space-time point [Tripathi et al., 2008].

The timescale of fluctuations analyzed here for the swings of Oligocene climate are relatively long-term (several Myr). Short-term fluctuations (i.e., on timescales shorter than several Myr), of periods 405 kyr and 1.2 Myr, were identified in the high-resolution record from ODP 1218 [Pälike et al., 2006] and related to Earth’s orbital variations in eccentricity and obliquity, respectively. This short-term “heartbeat” of Oligocene climate was superimposed on the long-term swings we describe here. The availability of further high-resolution records would allow an improved understanding of the interplay of these variations of different timescales.

### Table 9. Results[^a]: Oligocene Swinging Trends (O-Swings), Early Part (cf. Figure 10)

<table>
<thead>
<tr>
<th>Record</th>
<th>Age (Ma)</th>
<th>Slope[^b], Start (‰Myr[^c])</th>
<th>Slope[^b], End (‰Myr[^c])</th>
</tr>
</thead>
<tbody>
<tr>
<td>ODP 1218</td>
<td>32.04 ± 0.32</td>
<td>0.23 ± 0.10</td>
<td>−0.06 ± 0.24</td>
</tr>
<tr>
<td>High-Latitude Records</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DSDP 522</td>
<td>32.14 ± 0.17</td>
<td>0.44 ± 0.08</td>
<td>−0.20 ± 0.13</td>
</tr>
<tr>
<td>ODP 689</td>
<td>32.58 ± 0.52</td>
<td>0.33 ± 3.50</td>
<td>+0.03 ± 2.09</td>
</tr>
<tr>
<td>ODP 744</td>
<td>32.99 ± 0.13</td>
<td>0.70 ± 0.26</td>
<td>−0.00 ± 0.03</td>
</tr>
<tr>
<td>ODP 748</td>
<td>32.14 ± 0.20</td>
<td>0.44 ± 0.12</td>
<td>−0.23 ± 0.11</td>
</tr>
</tbody>
</table>
| Average[^c] | 32.52 ±0.24 
±0.11 | 0.46 ±0.04 
±0.07 | −0.03 ±0.04 
±0.03 |

[^a]: Values are rounded.
[^b]: Negative and positive slopes indicate glaciation/cooling and deglaciation/warming, respectively.
[^c]: Weighted average with external error (superscript) and internal error (subscript); time average includes dating-error effects. Slope is given by amplitude/duration; dating-error effects via duration on slope are negligible against amplitude error effects.

### Table 10. Results[^a]: Oligocene Swinging Trends (O-Swings), Late Part (cf. Figure 10)

<table>
<thead>
<tr>
<th>Record</th>
<th>Age (Ma)</th>
<th>Slope[^b], Start (‰Myr[^c])</th>
<th>Slope[^b], End (‰Myr[^c])</th>
</tr>
</thead>
<tbody>
<tr>
<td>DSDP 77</td>
<td>25.98 ± 0.77</td>
<td>−0.10 ± 0.03</td>
<td>+0.27 ± 0.25</td>
</tr>
<tr>
<td>DSDP 666</td>
<td>28.45 ± 0.51</td>
<td>−0.11 ± 0.04</td>
<td>+0.13 ± 0.07</td>
</tr>
<tr>
<td>DSDP 574</td>
<td>29.85 ± 1.15</td>
<td>−0.22 ± 0.25</td>
<td>+0.08 ± 0.08</td>
</tr>
<tr>
<td>ODP 667</td>
<td>28.23 ± 1.46</td>
<td>−0.13 ± 2.70</td>
<td>+0.06 ± 0.94</td>
</tr>
<tr>
<td>ODP 803</td>
<td>24.68 ± 2.74</td>
<td>−0.08 ± 1.01</td>
<td>+0.45 ± 0.48</td>
</tr>
<tr>
<td>ODP 1218</td>
<td>26.92 ± 0.22</td>
<td>−0.07 ± 0.01</td>
<td>+0.19 ± 0.03</td>
</tr>
</tbody>
</table>
| Average[^c] | 27.19 ±0.37 
±0.21 | −0.07±0.01 
±0.01 | +0.17±0.02 
±0.02 |
| High-Latitude Records |
| DSDP 522 | 29.33 ± 1.54 | +0.16 ± 3.56 | +0.03 ± 0.43 |
| DSDP 529 | 24.96 ± 2.13 | −0.04 ± 6.04 | +0.73 ± 1.00 |
| ODP 689 | 27.56 ± 0.77 | −0.09 ± 4.31 | +0.07 ± 0.06 |
| ODP 690 | 28.77 ± 2.12 | −0.13 ± 0.69 | −0.02 ± 0.39 |
| ODP 703 | 26.58 ± 1.11 | −0.07 ± 0.08 | +0.17 ± 0.21 |
| ODP 744 | 28.34 ± 0.37 | −0.14 ± 0.03 | +0.16 ± 0.02 |
| ODP 748 | 26.67 ± 2.04 | −0.04 ± 0.37 | +0.08 ± 0.15 |
| Average[^c] | 28.01 ±0.31 
±0.31 | −0.13±0.01 
±0.03 | +0.15±0.01 
±0.02 |

[^a]: Values are rounded.
[^b]: Negative and positive slopes indicate glaciation/cooling and deglaciation/warming, respectively.
[^c]: Weighted average with external error (superscript) and internal error (subscript); time averages include dating-error effects. Slope is given by amplitude/duration; dating-error effects via duration on slope are negligible against amplitude error effects.
already a conservative (large) choice. Adopting a sensitivity experiment to study the effects of increasing $s_{\text{date}}$ from 0.1 Myr [Cramer et al., 2011], which is already a conservative (large) choice. Adopting $s_{\text{date}} = 0.2$ Myr led to an OMB start of 23.24 ± 0.09 Ma (conservative error bounds) for the low latitudes and 23.64 ± 0.10 Ma for the high latitudes: still significantly different. Also, the conclusion that the OMB end was earlier for low than for high latitudes was not found to...
and also the shape of the typical course over time, points at a dynamical similarity: also, OMB may reflect a glaciation.

The low- and high-latitude OMB records display a remarkably close similarity in their amplitudes (Table 11). The combined overall weighted mean from the 15 records is a cooling/glaciation amplitude of 0.60‰, with a systematic error of 0.03‰ and a statistical error of 0.02‰. (We think that Miller et al. [1991] as well as Shevenell and Kennett [2007] overestimated the \( \delta^{18}O \) amplitude; both of them give a value of \( \sim 1.0 \)‰. This overestimation stems from employing too few records and/or putting too much emphasis on the extremes.)

The close similarity between low and high latitudes in the \( \delta^{18}O \) amplitude (an increase of 0.60 ± 0.03‰ across the OMB) suggests a more global signal origin: ice volume changes, in line with previous assessments.

The estimated duration of the OMB transition is with \( \sim 0.2 \) to 0.3 Myr about the same as that of the EOT glaciation (section 4.1.3.1). And also, the OMB transition leaves in the large majority of the analyzed records the imprint of an overshoot behavior followed by a recovery (indicated by arrows in Figure 11)—back to nearly as “warm” conditions as before the transition. This similarity between the EOT glaciation and the OMB transition in quantitative timing parameter values, as well as the implication that the OMB transition may reflect a glaciation, suggest a dynamical similarity in the processes.
[Miller et al., 1991; Shevenell and Kennett, 2007]. However, the temperature signal cannot be neglected: Billups and Schrag [2002] made Mg/Ca paleothermometry on the low-resolution, benthic record ODP 747 and found [Billups and Schrag, 2002, Figure 6 therein] a cooling of \(~1^\circ C\). Lear et al. [2004, p. 6 therein] applied the Mg/Ca method to high-resolution benthic ODP 1218 data, finding evidence for bottom water cooling-warming cycles of \(~2^\circ C\) amplitude, cycles that occurred before and around the OMB. Furthermore, Mawbey and Lear [2013] document an orbital component in the deep water temperature history at Ceara Rise sites ODP 926 and ODP 929, with amplitudes in the order of \(2^\circ C\). Interestingly, there may exist a time lag between temperature and ice volume changes. Shevenell and Kennett [2007, p. 2318 therein] noted that “cooling of deep-ocean waters may have played a role in triggering Mi-1.”

A temperature signal of \(1.5 \pm 0.5^\circ C\) or \(0.35 \pm 0.12\%\) \(\delta^{18}O\) implies a lower limit for the ice volume-related \(\delta^{18}O\) amplitude of \(0.25 \pm 0.12\%\) for the OMB glaciation. This value is less than, but in size comparable to, the \(\delta^{18}O\) amplitude of \(0.39 \pm 0.04\%\) associated with the NHG in the late Pliocene [Mudelsee and Raymo, 2005], but the geological evidence [Naish et al., 2001] suggests that at least part of the ice was stored on Antarctica. Shevenell and Kennett [2007] make the interesting observation that at around the OMB glaciation, the Antarctic ice sheet could have reached the continental shelf, with the resulting consequences of marine-based, fluctuating ice sheets, as described for the Northern Hemisphere in the Pleistocene [Berger and Jansen, 1994].

Regarding the causes of the OMB glaciation, note that prior to the glaciation, the temperatures of the deep ocean had changed (the lag behavior described in a previous paragraph). Further note the overshoot/recovery behavior, which previous work [Zachos et al., 2001b; Lear, 2007] had also described. Zachos et al. [2001b, p. 277 therein] use the insightful expression “overshoot of equilibrium.” This hints at considering elements of nonlinear dynamical systems theory [Stanley, 1971] for explaining abrupt paleoclimatic transitions. Abruptness requires that positive feedback loops set in, which could enhance and accelerate an ongoing transition by orders of magnitude.

In the case of the glaciation of Antarctica (or the Northern Hemisphere), it seems clear that two major, more or less independent atmospheric climate variables have to act together to yield rather fast changes in ice volume [Prentice and Matthews, 1991]: temperature (which needs to be low) and moisture (which needs to be high). One or the other variable may take the lead, driven externally by tectonic changes [Ruddiman et al., 1997; Crowley and Burke, 1998; Kennett and Exon, 2004] or astronomical pacing [Zachos et al., 1997; Naish et al., 2001; Zachos et al., 2001b; Pälike et al., 2006]. The other variable, influenced externally or by local climate noise (other, unknown influences) has to pass a certain threshold, a “material constant” of the paleoclimatic system, to let the abrupt change start. The change itself may then invoke negative feedback loops, which would bring an end at a new, intermediate equilibrium level [Stanley, 1971].

4.1.6. Miocene

The OMB transition (section 4.1.5.1) was a strong glaciation, which initiated the Miocene. It was not the only one in that epoch: “Isotope workers generally agree that there was a buildup of the Antarctic ice sheet in the middle Miocene” [Miller et al., 1991, p. 6840 therein], and these authors proposed that there were seven glaciation intervals (“Mi events”) in the early to middle Miocene. However, on a longer timescale, the Miocene climate trend contains a warming in its early half, which culminated in the climatic optimum (MMCO). Cronin [2010] notes that this warm interval was from 18 to 14 Ma. We statistically analyze the timing of this warm interval (finding that it was from \(~17 to 15 Ma\)) and also the amplitudes of the change points (start and end) in the middle Miocene climate. The later half of the Miocene brought, on a longer timescale, cooling and glaciation, driving climate away from the relatively warm MMCO.

We ask for attention of usage of words “start” and “end” in these sections on the Miocene. The MMCO-Start transition started and ended, then the MMCO was established. Afterward, the Middle Miocene Climate Transition (MMCT) [Flower and Kennett, 1993a; Shevenell et al., 2004] brought the MMCO to an end.

4.1.6.1. Climate Transition MMCO-Start

The start of the warming toward the MMCO (start of MMCO-Start) was at \(~17.5 Ma\); low and high latitudes were coeval within the considerably large error bars (Figure 12 and Table 12). The chief reason for the large estimation errors is the large climate variability around the trend, which is in its size even comparable to the \(\delta^{18}O\) amplitude of \(~0.2 to 0.4\%\) (Table 12). MMCO-Start was completed by \(~17.0 Ma\).
Some of the records documenting the MMCT reflect an almost abrupt $\delta^{18}O$ increase (DSDP 317, DSDP 366, and ODP 667 from the low latitudes; DSDP 281 and DSDP 555 from the high latitudes), while the other, which include some rather highly resolved series (e.g., DSDP 574 or ODP 1171), exhibit a longer-term transition that can be excellently statistically modeled by means of a ramp (Figure 13). The low resolution or the existence of hiatuses at around the MMCT may be responsible for the apparent abruptness at DSDP 317 and DSDP 366, but more data and improved age models are required before the possibility of an abrupt end to the MMCO in some places may be dismissed. For example, Holbourn et al. [2005, 2007] construct benthic $\delta^{18}O$ records from two low-latitudinal sites (ODP 1146, western Pacific; ODP 1237, eastern Pacific) that

![Figure 12](https://example.com/figure12.png)

**Figure 12.** Results, climate transition MMCO-Start (a) low latitudes; (b) high latitudes); cf. Figure 5. See Table 12 for numerical results.

The large climate variability imprinted on the $\delta^{18}O$ curves stems partly from the Mi events of ice volume fluctuations [Miller et al., 1991] and from deep water variability [Wright et al., 1992; Billups and Schrag, 2002].

### 4.1.6.2. MMCT

Although a few benthic $\delta^{18}O$ time series from the low latitudes and nine series from the high latitudes contribute to the estimation of the timing when the transition from the warm MMCO to colder conditions started and ended, the statistical results appear somewhat ambiguous (Figure 13 and Table 13). The start was at $\sim$15 Ma or slightly later, and the estimates for low and high latitudes are apart from each other by $0.32 \pm 0.26$ Myr (conservative error bounds propagated), that is, high latitudes started earlier. The end was clearly earlier ($\sim$14 Ma) for high compared to low latitudes ($\sim$13 Ma). For the other two major glaciation steps, EOT (Table 8) and OMB (Table 11), the high latitudes were either earlier (less likely) or coeval (more likely) with the low for both start and end.

Some of the records documenting the MMCT reflect an almost abrupt $\delta^{18}O$ increase (DSDP 317, DSDP 366, and ODP 667 from the low latitudes; DSDP 281 and DSDP 555 from the high latitudes), while the other, which include some rather highly resolved series (e.g., DSDP 574 or ODP 1171), exhibit a longer-term transition that can be excellently statistically modeled by means of a ramp (Figure 13). The low resolution or the existence of hiatuses at around the MMCT may be responsible for the apparent abruptness at DSDP 317 and DSDP 366, but more data and improved age models are required before the possibility of an abrupt end to the MMCO in some places may be dismissed. For example, Holbourn et al. [2005, 2007] construct benthic $\delta^{18}O$ records from two low-latitudinal sites (ODP 1146, western Pacific; ODP 1237, eastern Pacific) that

### Table 12. Results: Climate Transition MMCO-Start (cf. Figure 12)

<table>
<thead>
<tr>
<th>Record</th>
<th>Start (Ma)</th>
<th>End (Ma)</th>
<th>Duration (Myr)</th>
<th>Midpoint (Ma)</th>
<th>$\delta^{18}O$ Amplitude $^b$ (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Low-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DSDP 289</td>
<td>17.05 ± 1.02</td>
<td>16.92 ± 0.95</td>
<td>0.14 ± 0.64</td>
<td>16.98 ± 0.93</td>
<td>+0.23 ± 0.14</td>
</tr>
<tr>
<td>DSDP 317</td>
<td>17.01 ± 1.18</td>
<td>16.76 ± 0.77</td>
<td>0.24 ± 0.91</td>
<td>16.89 ± 0.88</td>
<td>+0.19 ± 0.14</td>
</tr>
<tr>
<td>DSDP 366</td>
<td>18.93 ± 0.94</td>
<td>18.77 ± 0.86</td>
<td>0.16 ± 1.02</td>
<td>18.85 ± 0.74</td>
<td>+0.22 ± 0.11</td>
</tr>
<tr>
<td>ODP 667</td>
<td>17.26 ± 1.25</td>
<td>16.92 ± 0.88</td>
<td>0.35 ± 1.02</td>
<td>17.09 ± 0.95</td>
<td>+0.19 ± 0.14</td>
</tr>
<tr>
<td>Average$^a$</td>
<td>17.70 ± 0.50</td>
<td>17.33 ± 0.48</td>
<td>0.20 ± 0.05</td>
<td>17.62 ± 0.51</td>
<td>+0.21 ± 0.01</td>
</tr>
<tr>
<td><strong>High-Latitude Records</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>DSDP 563</td>
<td>17.92 ± 0.86</td>
<td>16.58 ± 0.73</td>
<td>1.34 ± 1.04</td>
<td>17.25 ± 0.60</td>
<td>+0.33 ± 0.10</td>
</tr>
<tr>
<td>DSDP 608</td>
<td>17.19 ± 0.90</td>
<td>16.99 ± 0.84</td>
<td>0.20 ± 0.69</td>
<td>17.09 ± 0.80</td>
<td>+0.28 ± 0.10</td>
</tr>
<tr>
<td>ODP 704</td>
<td>20.03 ± 1.10</td>
<td>16.46 ± 0.94</td>
<td>3.56 ± 1.68</td>
<td>18.24 ± 0.58</td>
<td>+0.36 ± 0.09</td>
</tr>
<tr>
<td>ODP 744</td>
<td>17.79 ± 0.38</td>
<td>17.14 ± 0.28</td>
<td>0.65 ± 0.59</td>
<td>17.47 ± 0.16</td>
<td>+0.64 ± 0.11</td>
</tr>
<tr>
<td>ODP 1090</td>
<td>16.78 ± 0.30</td>
<td>16.75 ± 0.24</td>
<td>0.03 ± 0.47</td>
<td>16.76 ± 0.14</td>
<td>+0.33 ± 0.06</td>
</tr>
<tr>
<td>Average$^a$</td>
<td>17.32 ± 0.23</td>
<td>16.89 ± 0.18</td>
<td>0.46 ± 0.30</td>
<td>17.15 ± 0.20</td>
<td>+0.37 ± 0.05</td>
</tr>
</tbody>
</table>

$^a$Values are rounded.

$^b$Negative and positive amplitudes indicate glaciation/cooling and deglaciation/warming, respectively.

$^c$Weighted average with external error (superscript) and internal error (subscript); start, end, and midpoint averages include dating-error effects.
are both relatively short but bracket the MMCT at high temporal resolution: these records indicate a rather abrupt glaciation at around 13.9 Ma. Notably, the MMCT does not exhibit an overshoot/recovery signature.

While the 17 $\delta^{18}O$ records display scatter in the timing estimates, they render a rather close agreement in the $\delta^{18}O$ amplitude. The overall weighted mean of the entries (Table 13) is 0.88‰ with a systematic error of 0.04‰ and a statistical error of 0.02‰. This $\delta^{18}O$ increase is, within error bars, of the same magnitude as the increase across the EOT (section 4.1.3.1) and of $\sim$0.3‰ larger magnitude than the increase across the OMB (section 4.1.5.1). Already, the size of the amplitude, and also the coherency across results (Table 13), points to a strong ice volume component (glaciation) at the end of the MMCO.

Previous papers vary in their assessment of whether the MMCT was abrupt or not. Miller et al. [1987, p. 9 therein] mentioned a “sharp” $\delta^{18}O$ increase at $\sim$15 to 13 Ma, which they interpreted as a reestablishment or intensification of glacial conditions. Lear et al. [2000] as well as Billups and Schrag [2002] showed plots of Mg/Ca-derived deep water temperature changes and concluded that a gradual cooling by $\sim$2 to 3°C occurred. Using the advanced analytical technique via paired records of Mg/Ca and Li/Ca, indicating temperature and carbonate saturation variations, Lear et al. [2010] found a long-term change from 15.3 to 12.5 Ma and estimated an overall cooling of $\sim$1°C. In their review paper on Cenozoic climate evolution, Zachos et al. [2001a] interpreted previous work with oxygen and carbon isotopes on marine benthic records [Vincent et al., 1985; Flower and Kennett, 1995] and also concluded that the cooling was gradual and that a reestablishment of major Antarctic ice sheets occurred across the MMCT. On the other hand, Shevenell and Kennett [2007] relied on the work that generated the $\delta^{18}O$ record DSDP 281 [Shackleton and Kennett, 1975], which exhibits a very fast transition (Figure 13) and, thus, assessed the MMCT as abrupt, of $\sim$1‰ amplitude at 14 Ma. The variety of assessments underlines the importance of analyzing a multitude of records from distributed locations (Tables 1 and 2) to be able to distinguish between more regional and more global climate signals.

If we assume that besides Antarctic ice growth across the MMCT there occurred also “global cooling” [Shevenell and Kennett, 2007, p. 2319 therein], then we may start an ice volume estimation on the basis of a deep water cooling of 1.5 ± 0.5°C; the large error should accommodate for the fact that the cooling amplitude is inferred only from a few coarsely resolved [Lear et al., 2000; Billups and Schrag, 2002] and one
There is more abundant oxygen isotopic evidence for the MMCT cooling in the Miocene than for previous glaciations, and this evidence suggests that a complex causal network existed, spatially heterogeneous processes and feedback mechanisms acted, and the cooling was not globally homogeneous. Recently, higher-resolved [Lear et al., 2010] time series. The corresponding δ18O signal is 0.35 ± 0.12‰, and the ice volume-related δ18O signal from the MMCT is

\[(0.88 ± 0.04‰) - (0.35 ± 0.12‰) \approx 0.53 ± 0.13‰.\]

Thus, the signal proportion of ice volume is

\[1 - (0.35 ± 0.12)/(0.88 ± 0.04) \approx 0.60 ± 0.14.\]

Shevenell and Kennett [2007] assessed “direct” signal proportion determination (via Mg/Ca) and indirect determination and claimed a value of 80 and 70%, respectively. This agrees with our value of 60 ± 14%. We recommend using the calculated ice volume amplitude of 0.53 ± 0.13‰ for quantitative testing by means of concepts and models of sea level changes and the ice sheet geometry on Antarctica or the Northern Hemisphere [Saltzman, 2002].

Cronin [2010] briefly reviewed causal explanations of the end of the MMCO, the cooling, and the ice sheet growth after the middle of the Miocene. An initial cooling shift could have been caused by an enhanced biological pumping activity of the ocean, which led to atmospheric CO2 removal into organic sedimentary matter, as is documented by an increase in δ13C [Holbourn et al., 2005, 2007, 2013; Diester-Haass et al., 2009, 2013] and in organic-rich layers in the coastal Monterey Formation, California, and elsewhere [Vincent and Berger, 1985]. Alternatively to this “Monterey hypothesis,” CO2 removal could also have been accomplished by enhanced weathering, for example, in the uplifted Himalaya region [Raymo, 1994]. Regardless, there is recent evidence for reduced CO2 during the middle Miocene [Foster et al., 2012; Badger et al., 2013; Masson-Delmotte et al., 2013]. An essential element of the causal chain is Southern Ocean circulation and heat transport [Flower and Kennett, 1993a, 1993b, 1994, 1995]. Our result (Table 13) from time series analysis of an earlier start, and an earlier end as well, of the cooling at high latitudes compared to low, may allow the interpretation that polar cooling amplification was also involved.
Knorr and Lohmann [2014] have used a coupled atmosphere-ocean model to disentangle the effects of CO₂ and Antarctic ice sheet changes on the characteristic temperature evolution during the MMCT. Ocean circulation changes in response to Antarctic ice sheet growth can cause a sea-surface warming and concurrent deep water cooling in large parts of the Southern Ocean. In contrast, longer-term surface and deep water temperature trends are dominated by CO₂ changes, indicating that the feedback with the ice sheet changes in Antarctica might induce nonheterogeneous trends in temperatures of bottom waters, surface waters, and in other properties (e.g., salinity). These impacts provide a coherent explanation for the Southern Ocean temperature evolution during the Miocene ice sheet expansion [Shevenell et al., 2008] and suggest that the climate response and feedback to large ice sheet changes in Antarctica may be more complex than previously thought.

4.1.7. Miocene-Pliocene

The preceding sections analyzed the Miocene down to a bound of ∼10 Ma. We quantified the relatively warm MMCO and, at the end of it, ∼13 to 14 Ma, the third major Cenozoic glaciation step (MMCT). Previous own work [Mudelsee and Raymo, 2005] had examined the interval from 2 to 4 Ma in order to quantify the fourth glaciation step, the NHG, which started in the Pliocene at ∼3.6 Ma. For the range between 4 and 10 Ma, there is room for further analysis. We identify a longer-term, but minor cooling trend, which we call 4-to-10-Ma.

4.1.7.1. Climate Transition 4-to-10-Ma

There are no discernable, or by means of statistical techniques quantifiable, climate steps within the interval from 4 to 10 Ma (Figure 14). The numerical results show mostly cooling trends (four from five low-latitude and nine from 12 high-latitude records); and the few warming trends detected are statistically not, or hardly (ODP 806), significant (Table 14). The fact that systematic errors are about 2 to 3 times larger than the statistical errors indicates spatially heterogeneous slope values. However, no geographical pattern (low versus high latitudes or individual ocean basins) can be identified (Table 14). For example, even within a single region, the Ontong Java Plateau (west equatorial Pacific), the slopes have different signs (cooling at Site DSDP 289 and warming at
around 41 kyr [Mudelsee et al., 2014]. The premise that high-latitude climate fluctuations are more influenced by changes in obliquity (period $AR(1)$ inertia of deep water temperature fluctuations at low latitudes, compared to high latitudes, would be smaller) is supported by time estimates with small error bars, which dominate the weighted averaging (Table 3).

Mudelsee (Tables 1 and 2), which bring the “equivalent autocorrelation coefficient,” $a^d$, closer to zero, that means, the lower bound. All high-resolution records (i.e., with $d < 100$ kyr) yield meaningful persistence time estimates with small error bars, which dominate the weighted averaging (Table 3).

The average slope, calculated over all records from Table 14, is $0.039 \pm 0.008\%\text{Myr}^{-1}$; conservatively, the larger (systematic) error bar is given. Over the full 6 Myr span, this means a $\delta^{18}O$ change of $0.23 \pm 0.05\%$, a value that can further be tested (e.g., by means of climate models) and possibly split into contributions from changes in ice volume (which should be rather small), in temperature of bottom waters and, hence, the feeding surface waters, and in other properties (e.g., salinity).

Previous works utilizing Mg/Ca thermometry on deep waters support the conclusion of minor gradual cooling trends in the late Miocene and early Pliocene: Billups and Schrag [2002] examined ODP 747 from the Southern Ocean (however, there is a gap between ~5 and 7.5 Ma in their record), Lear et al. [2003] analyzed ODP 806 from the Ontong Java Plateau and ODP 926 from the equatorial Atlantic, and Billups and Schneiderich [2010] reviewed (Oligocene to Miocene paleotemperature trends from foraminiferal Mg/Ca records obtained from six ODP sites. Agreeably, more records analyzed in that manner would help to draw a clearer geographical picture. Decreasing trends in atmospheric CO$_2$ have been inferred [Tripati et al., 2009] and made responsible for the cooling [Cerling et al., 1997; Tripati et al., 2009].

### 4.2. Cenozoic Climate Persistence

The persistence time is an important dynamical parameter that quantifies the “memory” or “inertia” of random climate fluctuations. It corresponds in the frequency domain to the redness of a climate spectrum.

From a technological perspective, obtaining the estimates of the $AR(1)$ persistence time $\tau$ (Table 3) was a challenge due to the difficult numerical minimization of the least squares cost function [Mudelsee, 2002]. For 15 of the 48 analyzed records, no result could be obtained. These difficulties seem to originate mainly from the relatively coarse resolution (i.e., large average spacing, $\bar{d}$) of those records (Tables 1 and 2), which brings the “equivalent autocorrelation coefficient,” $a = \exp(-\bar{d}/\tau)$, close to zero, that means, the lower bound. All high-resolution records (i.e., with $\bar{d} < 100$ kyr) yield meaningful persistence time estimates with small error bars, which dominate the weighted averaging (Table 3).

The white noise residuals of the fitted $AR(1)$ persistence model [Mudelsee, 2010],

$$
\epsilon(i) = e(i) - \exp[-(\tau(i) - t(i - 1))/\tau] \cdot e(i - 1),
$$

(19)

attested in lag-1 scatterplots of $\epsilon(i - 1)$ versus $\epsilon(i)$ (not shown) graphically that the $AR(1)$ model may be a suitable description. The idea is that in a scatterplot, the $AR(1)$ (red noise) data display an orientation along the 1:1 line (“warm yesterday, warm today”), while white noise data do not display such an orientation and rather look like a cloud [Mudelsee, 2010]. Even when considering the large systematic errors (Table 3), it seems that the low-latitude records exhibit a shorter memory ($\tau \approx 4.0$ kyr) than the high-latitude records ($\tau \approx 8.5$ kyr). To give a physical-climatological explanation is difficult. A reduced $AR(1)$ inertia of deep water temperature fluctuations at low latitudes, compared to high latitudes, would have to be derived from climate-theoretical principles of long-term Cenozoic climate evolution. Alternatively, one may interpret the $AR(1)$ parameter by invoking noise-influenced Milankovitch variability under the premise that high-latitude climate fluctuations are more influenced by changes in obliquity (period around 41 kyr [Berger, 1978]) and low-latitude fluctuations by changes in precession (periods around 19 kyr [Berger, 1978]), but still the astronomical-climatological link has to be evaluated [Laapple and Lohmann, 2014].
Figure 15. Cenozoic climate stacks (marine benthic δ¹⁸O). Shown are Zachos et al.'s stack (gray line), which comprises data from low and high latitudes, the new stack from low-latitude data (red line) with uncertainty band (red shaded), and the new stack from high-latitude data (blue line) with uncertainty band (blue shaded). The δ¹⁸O values are adjusted to the genus *Cibicidoides* and 0.64‰ added [Shackleton and Hall, 1984; Zachos et al., 2001a]. For more details, see Figure 1 on Cenozoic climate, section 3.3 on stack construction, and section 4.3 on results interpretation.

2009, Figure 7 therein], including the nonlinearities in the system. When neglecting ice sheet feedback, the low-latitude fluctuations are dominated by changes in precession, semiprecession, and eccentricity. Such astronomical-physical-climatological concepts should be tested by means of climate models [Saltzman, 2002] and theoretical analyses [Laepple and Lohmann, 2009; Livina et al., 2011, 2013].

We further examined, by estimating persistence for time intervals before and after 34 Ma in a pairwise manner, whether the existence of larger ice sheets in the later interval could have influenced the memory of δ¹⁸O fluctuations in deep waters. Only four records, all from high latitudes (DSDP 525, ODP 689, ODP 690, and ODP 748), are of high resolution and cover the intervals before and after 34 Ma sufficiently long (Table 2) to be assessed as representative. Persistence time estimation on these records, however, did not give conclusive results; in two cases (ODP 690 and ODP 748) the later interval showed a longer memory (as one would assume from the existing larger ice volumes with larger “time constants” than the temperature of water bodies [Mudelsee and Raymo, 2005]); in the two other cases the results were opposite. Also, the results on low-latitude records were inconclusive.

The AR(1) is a rather simple model of persistence in the fluctuations around the long-term (~Myr) climate trend: just the immediate past value is “remembered.” An alternative model is that variations (shorter than ~Myr) in the geometry of Earth’s orbit are superimposed on the long-term climate trends. Although the arguments brought in favor of such Milankovitch variability are partly based on isotope and other climate proxy records that had previously been tuned to match the astronomical target, we assess the available evidence [Fischer, 1981; Hilgen et al., 1995; Zachos et al., 1996; Kent, 1999; Hinov, 2000; Kashiwaya et al., 2001; Zachos et al., 2001a, 2001b; Abdul Aziz et al., 2003; Coxall et al., 2005; Lourens et al., 2005; Pälike et al., 2006; Röhl et al., 2007; Shevenell and Kennett, 2007; Westerhold et al., 2007; Lyle et al., 2008; Sexton et al., 2011; DeConto et al., 2012; Westerhold et al., 2014] as convincing enough to consider that indeed Milankovitch variability cannot be ignored even on Cenozoic timescales. The upper bounds of the persistence time estimates (Table 3) are in the order of a few tens of kiloyears—which seems basically compatible with the result of fitting AR(1) models to noise-influenced Milankovitch time series (eccentricity, obliquity, and precession). Further evidence should be obtainable from spectrum estimation [Priestley, 1981; Percival and Walden, 1993; Bendat and Piersol, 2010; Mudelsee, 2010] on the residuals series, ei(t). This task is beyond the scope of this “time-domain” review. Researchers embarking on it should consider using spectrum estimation tools that (1) can directly process the unevenly spaced records [Schulz and Mudelsee, 2002] and (2) take timescale uncertainties into account [Mudelsee et al., 2009].

### 4.3. Cenozoic Benthic δ¹⁸O Stacks

Figure 15 shows the new stacks (for low and high latitudes) of benthic oxygen isotopic composition together with Zachos et al.’s stack [Zachos et al., 2001a]. For comparing the records, one should bear in mind the effects the smoothing bandwidth has on a resulting stack. Zachos et al. [2001a] employed a five-point...
running mean, and their database shows an average spacing of ∼5.2 kyr over the interval [4 Ma; 61 Ma], that is, their stack was produced with a bandwidth in the order of 26 kyr. On the other hand, the optimized bandwidth employed in our kernel smoothing for the new stacks is clearly larger, roughly between 500 and 2000 kyr (Figure 3). That means, our new stacks use a stronger smoothing and utilize also a larger database than Zachos et al. [2001a]. Consequently, one should expect narrower statistical uncertainty bands for our stacks than for Zachos et al.’s stack. In addition to these statistical prerequisites, Zachos et al. [2001a, 2008] remarked that portions of their stack may be biased due to the uneven spatial and temporal data distribution.

Considering the “theoretical” shortcomings of Zachos et al.’s stack mentioned in the previous paragraph, visually overlaying the three stacks reveals a surprisingly good agreement, over most time intervals, between Zachos et al.’s stack on the one hand and each of our new stacks on the other (Figure 15). Larger, systematic discrepancies seem to be restricted to the interval from about the end of the EOGM in the early Oligocene (∼31 Ma) to about the start of the MMCO in the middle Miocene (∼17 Ma); this interval is discussed in a subsequent paragraph. The old stack [Zachos et al., 2001a] appears to deviate from the two benthic stacks also regarding short-term excursions, such as the PETM (∼55 Ma for Zachos et al.’s stack) or the MECO (∼41 Ma). However, this cannot be interpreted as a deviation but rather as an inevitable away-smoothing of those short-term excursions by the large kernel bandwidths in our two new benthic kernel stacks.

Also, comparing the low with the high latitudes in the kernel stacks (Figure 15) reveals a rather close agreement. It seems that systematic deviations are restricted to roughly that interval between 17 and 31 Ma. This agreement is remarkable for the earlier part (Paleocene-Eocene), where the amount of ice is thought to be insignificant: it implies that on longer timescales (∼Myr) the polar bottom water amplification was weak or even absent. On shorter timescales, at around the PETM, we quantified the amplification factor as somewhere around two (section 4.1.1.2). That means, not only climate sensitivity [PALAESENS Project Members, 2012] but also polar amplification should be evaluated in dependence on the analyzed timescale of the changes.

5. Conclusions and Future Directions

We analyzed a large data compilation of marine benthic δ¹⁸O records [Cramer et al., 2009] by means of statistical tools [Mudelsee, 2010]. This allowed quantification of Cenozoic climate evolution in terms of ice volume and temperature. We determined the timing and amplitudes of climate transitions and events (Figure 1) with realistic error bars, taking into account the various sources of uncertainty (measurement and proxy noise, dating errors). Even so, the uneven spatiotemporal data distribution may introduce bias in our results, and also, our approach of comparing low with high latitudes may yield biased results for time intervals of strong latitudinal dependent evolutionary processes. Therefore, the taken uncertainty-analytical approach was conservative, that means, the reported error bars and constructed uncertainty bands are to be seen as upper bounds; it is unlikely that they underestimate the true magnitude of the uncertainty. We further constructed two stacks (for low and high latitudes) of benthic δ¹⁸O with error band. This form of quantitative reanalysis review is thought to contribute to advancing the quantitative and causal understanding of Cenozoic climate changes.

During the Paleocene–Eocene epochs, a gradual warming (called PE-Trend) occurred, which was punctuated by the PETM warming event ∼55.76 Ma. PE-Trend led to the EECO (∼54 to 49 Ma), the warmest longer phase during the entire Cenozoic.

The Eocene saw a long-term global cooling, which can be separated into two gradual transitions (LTEC-I and LTEC-II), in between of which was the MECO. Low-latitude deep water cooled stronger than high- (southern) latitude deep water, ice volume had only a minor signal proportion.

The transition from the Eocene to the Oligocene (EOT) was a cooling associated with ice buildup on Antarctica. This was followed by a partial recovery to warmer conditions, an overshoot behavior that led to the EOGM.

Subsequent to climate swings comprising warming/deglaciation and cooling/glaciation during the Oligocene (termed “O-Swings”), the OMB at the boundary to the Miocene (∼23 Ma) constituted a second major glaciation step during the Cenozoic.
Table 15. Rates of Change\(^a\) of Cenozoic Climate Transitions

<table>
<thead>
<tr>
<th>Transition</th>
<th>Glaciation/Cooling (‰ Myr(^{-1}))</th>
<th>Deglaciation/Warming (‰ Myr(^{-1}))</th>
</tr>
</thead>
<tbody>
<tr>
<td>PE-Trend(^b)</td>
<td>+1.57(^{+0.016}_{-0.018})</td>
<td>+1.68(^{+0.776}_{-0.262})</td>
</tr>
<tr>
<td>PETM(^c)</td>
<td>-1.46(^{+0.046}_{-0.223})</td>
<td>+1.68(^{+0.014}_{-0.013})</td>
</tr>
<tr>
<td>LTERC-I(^d)</td>
<td>-0.16(^{+0.013}_{-0.014})</td>
<td>+1.68(^{+0.097}_{-0.074})</td>
</tr>
<tr>
<td>LTERC-II(^e)</td>
<td>-0.41(^{+0.058}_{-0.074})</td>
<td>+1.68(^{+0.296}_{-0.265})</td>
</tr>
<tr>
<td>EOT(^f)</td>
<td>-1.87(^{+0.026}_{-0.025})</td>
<td>+0.520 \pm 0.223</td>
</tr>
<tr>
<td>EOT recovery(^g)</td>
<td>+0.114(^{+0.024}_{-0.052})</td>
<td></td>
</tr>
<tr>
<td>OMB(^h)</td>
<td>-1.05(^{+0.163}_{-0.097})</td>
<td></td>
</tr>
<tr>
<td>MMCO-Start(^i)</td>
<td>+0.114(^{+0.024}_{-0.052})</td>
<td></td>
</tr>
<tr>
<td>MMCT(^j)</td>
<td>-0.43(^{+0.031}_{-0.032})</td>
<td></td>
</tr>
</tbody>
</table>

\(^a\)Values are rounded; weighted averages of amplitude/duration with external error (superscript) and internal error (subscript). Negative and positive rates indicate glaciation/cooling and deglaciation/warming, respectively.

\(^b\)Table 4.

\(^c\)Table 5.

\(^d\)Table 6.

\(^e\)Table 7.

\(^f\)Table 8.

\(^g\)Section 4.1.3.1; only one record (ODP 1218).

\(^h\)Table 11.

\(^i\)Table 12.

\(^j\)Table 13.

The warm Mid-Miocene Climatic Optimum (MMCO), estimated as from \(~17\) to \(15\) Ma, ended with the third step of increased ice volume (estimated relative signal proportion \(60\ \pm 14\%)\). The late Miocene showed a minor longer-term cooling trend.

Regarding causal explanations for the various Cenozoic climate events and transitions, one cannot neglect the changes in atmospheric greenhouse gas concentrations, which have the potential to act also on short timescales (e.g., the PETM), and one cannot neglect long-term tectonic changes, such as those affecting the Southern Ocean circulation (e.g., the EOT). With significant ice volume arriving on Antarctica, the EOT brought into play another forcing and responding climate variable. This made the interaction of the relevant climate variables more complex, and it should also have enhanced the geographical differentiation (low and high latitudes). This differentiation may also be responsible for the long-term systematic deviations in the new benthic \(\delta^{18}O\) stacks between low latitudes (warmer) and high latitudes (colder) during the interval from \(17\) to \(31\) Ma. In other time periods, the new stacks do overlap considerably, and they do also agree with a previous, global benthic \(\delta^{18}O\) stack \([Zachos et al., 2001a]\).

Cenozoic climate transitions may be evaluated also in terms of rate of change (‰ per Myr). In analogy to considering the climate change rate in the recent past (few decades) and its causal agents \([Solomon et al., 2007; Stocker et al., 2013]\), also, the rate of change in the geologic past can be used to shed light on the Earth’s climatic situation. The results (Table 15) indicate that the PETM was a strong event, both in its earlier warming and in its later cooling phase. However, the PETM is somewhat different from the other transitions in that it was a rather short-term event. From the other, longer-term transitions, the EOT (\(-1.874\%\) Myr\(^{-1}\)), followed by the OMB (\(-1.058\%\) Myr\(^{-1}\)), are the two strongest Cenozoic climate transitions; both of which brought glaciation/cooling. The EOT was also strong in its recovery from the overshoot (\(+0.520\%\) Myr\(^{-1}\)). Notably, this recovery was stronger than the other warming transitions (PE-Trend or MMCO-Start). The third glaciation step (MMCT) was significantly weaker than EOT or OMB. These assessments are robust as regards the selection of the geologic timescale: A reanalysis of the PE-Trend rates following \(Cande and Kent\ [1995]\) instead of \(Gradstein et al. [2004]\) resulted in values that are within error bars indistinguishable (results not shown). The agreement for transitions or events that occurred later is likely not worse than for PE-Trend.
Future research may be enhanced along three directions (data, statistics, and models), leading to an increase in the current knowledge about the Cenozoic climate evolution. Crucial from an epistemological viewpoint is that researchers deepen or acquire the ability to integrate all three directions into their arsenal of methods.

5.1. Data
Since it is hard to disagree with the simple statement “more data are better,” the task here is rather to identify those dimensions in the data space where invested resources may yield a maximum of new information. Regarding the dimension “time,” a look on the result tables quickly informs that the Paleocene and Eocene epochs are clearly less well documented than later epochs. Regarding “geographical space,” it appears that currently we have fewer records from low latitudes. Since preservation of the information carriers (shells of foraminifera) depends on the CCD and, thus, the geographical location, it is likely rather difficult to achieve a state with more refined spatial information (e.g., several latitudinal belts). Cramer et al. [2009] notably study the role of different ocean basins.

Increasing the accuracy of the timescales of the Cenozoic isotope records (Cramer et al. [2011] give for the relative precision, \( \sigma_{t} \), a value of less than 0.1 Myr) is thought to be a major challenge. Absolute age determinations, less dependent on the validity of correctly assumed Milankovitch variations so far back in time, could be based from a chronostratigraphic framework that is based also on documented global stratigraphic events (e.g., the deposition of volcanic ashes): a huge and difficult compilation work to be carried out.

Another dimension in the data space is “proxy variable,” and \( \delta^{13}C \) is here a prime candidate for further analysis of changes in ventilation, productivity, and flow of water masses [Kroopnick, 1985; Zahn et al., 1986; Wright et al., 1992; Sarthein et al., 1994; Diester-Haass et al., 2009, 2013]. The \( \delta^{13}C \) compilation [Cramer et al., 2009] could be analyzed in a similar manner as in the present paper (for which it would be beyond the scope). Proxy variables for water temperature changes are clearly required to solve the \( \delta^{18}O \) partition problem (ice volume and temperature), and we expect future advances from Mg/Ca paleothermometry and also the more recently developed carbonate clumped isotope thermometry [Eiler, 2007; Tripati et al., 2010; Eiler, 2011].

5.2. Statistical Analysis
As regards parametric statistical time series analysis, the employed simple regression tools (linear, ramp, and break) were found useful for quantifying the observed Cenozoic climate events and transitions. However, for achieving a better quantitative understanding of the overshoot behavior (e.g., at around the OMB), it should be useful to develop slightly more complex parametric models, comprising more parameters—and constituting more difficult numerical challenges. It should be kept in mind that it makes sense to fit such advanced models only to time series data of high quality, coverage, and resolution.

As regards nonparametric statistical time series analysis, alternative procedures of stack construction via smoothing records individually before averaging should be theoretically examined. This may be achieved by means of Monte Carlo simulations, where a prescribed (i.e., known) stack target is superimposed with noise added to the individual records, and the performance of stack reconstruction techniques is assessed via error measures.

Timescale construction for marine sedimentary records currently uses interpolation and astronomical tuning on the basis of a set of dated fixed points (biostratigraphic, magnetostratigraphic, and other chrono-stratigraphic events). The Bayesian or frequentist methods adopted by the coral [Hendy et al., 2012], ice core [Parrenin et al., 2007; Klaubenbers et al., 2011], and speleothem [Scholz and Hoffmann, 2011; Hercman and Pawlak, 2012] communities could also significantly benefit the marine research community. It is important that a timescale construction algorithm delivers not only the best fit timescale but also simulated timescales (taking dating errors into account), which can be fed into computer simulation methods of climate time series analysis [Mudelsee, 2010]. Such methods are usually a hybrid of a parametric (timescale simulation) and a nonparametric part (bootstrap). They are computing intensive because both parts require many numerical random operations.

With the advent of new proxy records of Cenozoic climate, more spatial raw information enters the database. This principally allows construction of spatiotemporal stochastic models and fit them to data [Rue and Held,
2005; Diggle and Ribeiro, 2007; Cressie and Wikle, 2011; Tingley et al., 2012]. This could help to relate climate changes in time quantitatively to changes or gradients in geographical space. Evidently, observational data can also be compared with climate model output by means of spatiotemporal stochastic models. To slightly dampen the expectations, it should take a considerable amount of time until measurement capacities and computing power are able to supply enough meaningful data.

5.3. Climate Models

Unraveling the climate variations during the Cenozoic era by means of climate models requires technical advances in data acquisition, conceptual understanding, and numerical modeling. Coupled general circulation models (GCMs) have been utilized to evaluate the magnitude of future climate change [Solomon et al., 2007; Stocker et al., 2013]. Validation of these models by simulating warm Cenozoic climate states is essential for understanding the sensitivity of the climate system to external forcing. The models are clearly unrivaled in their ability to simulate a broad range of large-scale phenomena on seasonal to decadal timescales [Meehl et al., 2007], but their reliability on longer timescales, when the Earth will most likely enter into a warmer climate, requires additional evaluation. Cenozoic climate records derived from paleoenvironmental proxies allow testing of these models because they provide records of past warm climate conditions. With this knowledge, we can study the critical issue whether in Earth system models the key processes associated, for example, with ice sheets, clouds, permafrost, and global biogeochemical cycles—that are all relevant to the climate system—are well captured and if the range of possible solutions is covered by the models [Schmidt, 2010; Schmidt et al., 2014]. On the other hand, the models may provide tools to interpret the data in a meaningful way (e.g., regarding temperature) and to understand mechanisms of heterogeneous climate evolution [Knorr and Lohmann, 2014]. Past episodes of greenhouse warming furthermore provide insight into the coupling of the climate and the carbon cycle and thus may help to predict the consequences of unabated carbon emissions in the future [Zachos et al., 2008].

For performing the model experiments, one has to know the input parameters, such as greenhouse gas concentrations or the tectonic configuration [Mikalajewicz et al., 1993; von der Heydt and Dijkstra, 2006; Sijp et al., 2009; Henrot et al., 2010; Butzin et al., 2011; Sijp et al., 2011; Yang et al., 2013], and recently, such data sets of the gateway configurations have become available [Herold et al., 2008]. Substantial efforts have been made in this direction using Earth system models [Steph et al., 2006; You et al., 2009; Knorr et al., 2011], but a fully coupled approach including interactive ice sheets and atmosphere-ocean dynamics is still missing.

Until now, most climate models have difficulties in simulating the warm Cenozoic, especially at high latitudes [Michels et al., 2011; Huber, 2012; Dowsett et al., 2013; Salzmann et al., 2013]. It seems that the models systematically underestimate the climate sensitivity on long timescales [Lohmann et al., 2013a; Salzmann et al., 2013] and that they lack the ability to adequately simulate abrupt events such as the PETM [Valdes, 2011]. It may be that the models miss feedback in the system related, for example, to ocean mixing [Rose and Ferreira, 2013; Green and Huber, 2013] and/or the quantitative translation of the proxy records into climate variables is problematic. Furthermore, the modeling of the Cenozoic climate is limited to either time-slice experiments performed with GCMs [Bradshaw et al., 2012] or transient experiments performed with simplified climate models [Merico et al., 2008; Langebroek et al., 2009; Willett et al., 2013]. In the future, paleoclimate modeling shall reach a new stage, as increased computer power makes transient simulations with comprehensive global climate models (still with a low spatial horizontal resolution of several hundred kilometers) feasible. In contrast to conventional time-slice experiments, this approach is not restricted to equilibrium transitions and is, furthermore, capable of utilizing all available data for validation. Earth system models can be used for studying Cenozoic climate dynamics, as a “surrogate laboratory” for numerical experimentation with the climate system to explore mechanisms of the transitions during the Cenozoic (as documented here). In the Earth system model approach, synergetic effects of the climate components can be evaluated. For example, the vegetation effect on the ocean circulation may be one important mechanism explaining the relatively warm late Miocene climate [Knorr et al., 2011, and references therein]. The special challenges of paleoclimate modeling at these long timescales are balanced by unique opportunities to study the evolution of climate and the Earth system over its full dynamical range including multiple equilibria and thresholds [Pollard and DeConto, 2005; DeConto et al., 2008; Huber, 2012]. A particular aspect is related to proxy models (e.g., for δ18O), which are necessary to interpret the recorder systems and their large-scale climate information. This approach allows a more direct comparison of data and models. It offers—as in the case of δ18O—the possibility to distinguish between effects of temperature, hydrological cycle, and sea
level variations [Langebroek et al., 2010; de Boer et al., 2012]. Finally, it allows to test the assumptions about the climate recorder systems regarding leads, lags, and other filter properties [Laepple et al., 2011; Lohmann et al., 2013b].

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