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$\rm CO_2$ over the past 5 million years: Continuous simulation and new $\delta^{11}\rm B$ -based proxy data



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ABSTRACT

During the past five million yrs, benthic δ^{18} O records indicate a large range of climates, from warmer than today during the Pliocene Warm Period to considerably colder during glacials. Antarctic ice cores have revealed Pleistocene glacial-interglacial CO₂ variability of 60-100 ppm, while sea level fluctuations of typically 125 m are documented by proxy data. However, in the pre-ice core period, CO₂ and sea level proxy data are scarce and there is disagreement between different proxies and different records of the same proxy. This hampers comprehensive understanding of the long-term relations between CO₂, sea level and climate. Here, we drive a coupled climate-ice sheet model over the past five million years, inversely forced by a stacked benthic δ^{18} O record. We obtain continuous simulations of benthic δ^{18} O, sea level and CO₂ that are mutually consistent. Our model shows CO₂ concentrations of 300 to 470 ppm during the Early Pliocene. Furthermore, we simulate strong CO₂ variability during the Pliocene and Early Pleistocene. These features are broadly supported by existing and new δ^{11} B-based proxy CO₂ data, but less by alkenone-based records. The simulated concentrations and variations therein are larger than expected from global mean temperature changes. Our findings thus suggest a smaller Earth System Sensitivity than previously thought. This is explained by a more restricted role of land ice variability in the Pliocene. The largest uncertainty in our simulation arises from the mass balance formulation of East Antarctica, which governs the variability in sea level, but only modestly affects the modeled CO_2 concentrations.

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1. Introduction

The long-term interactions between CO_2 , temperature and sea level are a topical issue in climate science. Recently, there have been a number of attempts to quantify these interactions by studying data from paleo archives. For instance, CO_2 data of Antarctic ice cores and sea level reconstructions from Red Sea sedimentary archives show a close linear correlation over the past 516 ka (Foster and Rohling, 2013). However, an analysis of sea level and temperature records spanning the Cenozoic has indicated a nonlinear relation between these variables (Gasson et al., 2012). In the pre-ice core period, CO_2 and sea level data remain scarce. Moreover, uncertainties in CO_2 reconstructions are large and there is

http://dx.doi.org/10.1016/j.epsl.2016.01.022 0012-821X/© 2016 Elsevier B.V. All rights reserved. inter-proxy as well as intra-proxy disagreement (Masson-Delmotte et al., 2013; Beerling and Royer, 2011). This limits either the scope or the skill of such reconciling studies.

Benthic foraminiferal δ^{18} O records currently provide a more continuous and abundant data source on multi-million year timescales (Lisiecki and Raymo, 2005; Zachos et al., 2008). A complicating factor, however, is the interpretation of benthic δ^{18} O, because it comprises both an ice volume and a deep-sea temperature component. To untangle their relative contributions, two different approaches have been applied so far, namely (1) the use of independent deep-sea temperature proxies such as Mg/Ca of foraminiferal tests, and (2) ice-sheet modeling.

In this study, we expand on the model-based approach to deconvolute the δ^{18} O signal into temperature and sea level, including the simulation of CO₂. We introduce an inverse routine to iteratively calculate CO₂ concentrations over the past five million years from benthic δ^{18} O. The CO₂ is used to drive a recently developed

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Table 1

Model parameters for the ISM: centre height H_{cnt} , slope of the initial bed *s*, reference precipitation P_0 , critical radius R_c , ablation parameter C_{abl} , isotopic sensitivity β_T and isotopic lapse rate β_Z . Starred values indicate parabolic profiles, values given in m⁻¹.

Parameter	Unit	EuIS	NaIS	GrIS	EAIS	WAIS
H _{cnt}	m	1400	1400	800	1450	400
S	-	-0.0000165^{*}	-0.0000115^{*}	-0.0014	-0.0010	-0.0011
Po	m yr ⁻¹	0.88	1.15	1.34	0.71	1.37
R _c	km	1500	1800	750	2000	700
C _{abl}	-	-51	-41	-48	-30	-5
β_T	$\% K^{-1}$	0.35	0.35	0.35	0.6	0.8
β_Z	$\% {\rm km^{-1}}$	-6.2	-6.2	-6.2	-11.2	-11.2

coupled ice sheet-climate model (Bintanja, 1997; De Boer et al., 2010: Stap et al., 2014), which contains a scheme to calculate benthic δ^{18} O. In earlier work, this coupled model has been shown to be capable of reproducing glacial-interglacial cycles of ice volume and temperature over the past 800 kyr in forward mode, using CO₂ from ice cores as input (EPICA community members, 2004; Stap et al., 2014). We now force the model inversely by a stacked benthic δ^{18} O record (Lisiecki and Raymo, 2005), which enables us to study ice sheet-climate interactions in a broader range of climates. This new integrated approach improves upon earlier studies using an inverse benthic δ^{18} O routine (Bintanja and Van de Wal, 2008; De Boer et al., 2010, 2014) by including a climate model in the coupled set-up. It therefore facilitates a better representation of the deep-ocean temperature, as well as the simulation of seasonally-varving meridional-temperature profiles, rather than annual mean and globally uniform temperature perturbations with respect to pre-industrial climate. Moreover, in these earlier studies information on CO₂ was lacking. Taking a hybrid model-data approach, Van de Wal et al. (2011) obtained a continuous CO₂ reconstruction from a log-linear fit between modeled temperature and proxy CO₂ data. Here, however, CO₂ is incorporated in the model as a prognostic variable. Therefore, the simulated CO₂ is mutually consistent with eustatic sea level (ice volume equivalent), and with monthly mean atmospheric and oceanic temperatures, as deduced from benthic δ^{18} O. This improves our understanding of the role of CO₂ in climate variability.

We interpret our simulated CO₂ by studying the long-term relation between CO₂ and global surface air temperature in our model, known as Earth System Sensitivity (ESS), which is affected by the interaction between ice sheets and climate. In addition, we test the sensitivity of our CO₂ simulation to the modeled strength of the meridional ocean overturning, as well as to the formulation of the mass balance of East Antarctica and the relation between deepsea temperature and δ^{18} O. Finally, we compare our simulated CO₂ to existing CO₂ proxy data (Hönisch et al., 2009; Seki et al., 2010; Bartoli et al., 2011; Martínez-Botí et al., 2015a; Pagani et al., 2010; Zhang et al., 2013; Badger et al., 2013), complemented by a new δ^{11} B-based record.

2. Model and methods

2.1. Coupled ice-sheet-climate model and benthic δ^{18} O calculation

We use a recently developed coupled climate-ice sheet model (Stap et al., 2014). In this coupled set-up, the climate component is represented by a zonally averaged energy balance climate model, developed by Bintanja (1997) based on the model of North (1975). This climate model is tested for sensitivity to some important parameters in Bintanja (1997). It calculates surface temperature in zonal belts of 5° latitudinal and one layer vertical resolution, forced by 1000-year resolution insolation (Laskar et al., 2004). It uses a radiative transfer scheme and parameterizes energy transfer from the equator towards the poles as a diffusive process. Surface albedo is determined by the subdivision of the land surface into (potentially snow covered) grass, forest and land ice. A zonally

averaged ocean component of 5° resolution with 6 vertical layers, including a 1.25° thermodynamical sea-ice routine, is used to simulate the large-scale meridional ocean overturning. The ocean overturning strength is variable depending on the temperature difference between the polar and equatorial waters (Stap et al., 2014). The sensitivity of the coupled model to this formulation will be tested in Section 3.2.2.

The climate model is forced by CO₂ yielded by the inverse routine (Section 2.2). As discussed in Stap et al. (2014), the radiative forcing of CO₂ is enhanced by a factor 1.3 to account for the influence of other greenhouse gases (CH₄ and NO₂). The climate model is first run for 500 model years. Thereafter, a one-dimensional ice sheet model is run for the same 500 years. This ice sheet model, described in detail in De Boer et al. (2010), obtains ice velocities from the commonly used Shallow Ice Approximation (SIA). It calculates ice volume and surface height change of the five hypothetical axisymmetrical continents where the major ice sheets grow (North America (NaIS), Eurasia (EuIS), Greenland (GrIS), East-Antarctica (EAIS) and West-Antarctica (WAIS)), including the height-massbalance feedback. The continents are located at different latitudes and initially they are cone-shaped. Their different centre heights and slopes determine the maximum size and sensitivity to temperature of the ice sheets. The mass balance routine is forced by monthly temperatures (T) from the latitude in the climate model where the ice sheets are located (Stap et al., 2014). Precipitation P is obtained based on the Clausius-Clapeyron equation:

$$P = P_0 e^{0.04T - R/R_c},\tag{1}$$

where R is the radius of the ice sheet. P_0 and R_c are the presentday precipitation and critical radius respectively. An insolationtemperature melt equation is used to calculate ablation on the different ice sheets:

$$M = [10T + 0.513(1 - \alpha)Q + C_{abl}]/100.$$
 (2)

Here, α is surface albedo, and *Q* local radiation obtained from Laskar et al. (2004), Stap et al. (2014). Ice-sheet dependent tuning factors C_{abl} determine the threshold for which ablation starts. In Section 3.2.1 we will test the sensitivity of the model to the C_{abl} value of East Antarctica. All free parameter values (centre height, slope, P_0 , R_c and C_{abl}) for the ice sheets are listed in Table 1. In Stap et al. (2014), the tuning targets are discussed.

After the ice-sheet model has run 500 model years, the climate model receives the ice volume and surface height change (Δh_s) information. This is translated into ice extent, affecting the surface albedo, and surface height at the latitudes where the ice sheets are assumed to be located (Stap et al., 2014). With these new boundary conditions implemented, the climate model runs the next 500 years (Fig. 1). Applying a shorter coupling time interval does not lead to significantly different temperature and sea level output (Stap et al., 2014).

The ice sheet model includes a parameterization of benthic δ^{18} O values (De Boer et al., 2010) using the following equation:

$$\delta^{18} \mathbf{O} = [\delta^{18} \mathbf{O}_b]_{PD} - \frac{\overline{\delta^{18} \mathbf{O}_i V_i}}{V_o} + \left[\frac{\overline{\delta^{18} \mathbf{O}_i V_i}}{V_o} \right]_{PD} + \gamma \,\Delta T_o. \tag{3}$$



Fig. 1. Schematic overview of the coupled model. Novelties in the set-up with respect to Stap et al. (2014) are marked by a red dashed line, and to De Boer et al. (2010) by a blue dashed line.

The first term on the right hand side is the observed present-day value of benthic δ^{18} O. The influence of the ice sheets on the signal is represented by the second and third term. Here, V_o and V_i are volume of the ocean and land ice respectively. The following formulation of the isotopic content of the ice sheets is adopted (Cuffey, 2000):

$$\delta^{18}O_i = \delta^{18}O_{PD} + \beta_T \Delta T + \beta_Z \Delta Z.$$
(4)

Here, β_T and β_Z are ice-sheet dependent parameters, that determine the influence of annual mean temperature change (ΔT) and surface height change (ΔZ); their values are the same as used by De Boer et al. (2010) (Table 1). Present-day isotopic contents match the modeled values of an earlier study by Lhomme et al. (2005). The final term on the right hand side of Eq. (2) quantifies the influence of deep-sea temperature change with respect to present day (ΔT_o). Gain factor γ is set to 0.28 % K⁻¹, taken from a paleotemperature equation (Duplessy et al., 2002). We assess the sensitivity of the model to this value in Section 3.2.3. The deep-sea temperature perturbation is determined from the climate model as the 40–80°N mean of the second vertical ocean layer, representative of the mid-latitude North Atlantic deep ocean.

The energy balance and 1-D ice-sheet models used in our study are less comprehensive than current intermediate complexity (EMICs) and general circulation models (GCMs), and 3-D ice sheet models. However, they have the advantage of allowing for several five-million-year integrations of the coupled ice sheetclimate system, while capturing the relevant large-scale physical processes, notably the interaction between ice sheets and climate (Stap et al., 2014).

2.2. Inverse benthic δ^{18} O routine to calculate CO₂ concentration

We use an inverse forward modeling approach to calculate CO₂ from benthic δ^{18} O data. This is achieved by a two-step iterative routine. Each 1000-year cycle starts with an update of the insolation input. At this first iteration step, a new CO₂ concentration is obtained from the difference between the modeled benthic δ^{18} O value and the observed value 500 years later:

$$CO_2 = \overline{CO_2} * \exp[c * \{\delta^{18}O(t) - \delta^{18}O_{obs}(t + 0.5 \text{ kyr})\}].$$
 (5)

The coupled model is run for 500 model years. Thereafter, as a second iteration step Eq. (5) is applied again. The model is then run for another 500 years, using the updated CO_2 , but still forced by the same insolation. While in principle this yields 500 year-resolution CO_2 , only the results after the second iteration step are recorded and displayed in this paper. The temporal resolution of



Fig. 2. Data-model comparison of CO₂ over the past 800 kyr. Modeled CO₂ over the past 800 kyr (red), compared to the EPICA ice-core record (EPICA community members, 2004) interpolated to 1000-yr temporal resolution (blue).

the simulated CO_2 is therefore 1000 years. This is the desired resolution, as the physics in our model are not detailed enough to capture sub-millennial climate variations (Stap et al., 2014). We justify excluding the intermediate CO_2 values, by running the model again in forward mode, forced by the 1000-year resolution simulated CO_2 record; this does not significantly alter the resulting climate and ice volume records.

In Eq. (5), $\overline{CO_2}$ is the mean CO_2 concentration of the preceding 15 kyr, which reflects the long-term timescale of the carbon cycle. Together with parameter c, which is set to $0.45\%^{-1}$, it determines the strength of the response of CO_2 to changes in $\delta^{18}O$. While *c* is kept constant, it is important to stress that a variable relation between δ^{18} O and CO₂ is ensured by the carbon-cycle timescale, and most importantly by the second iteration step in the inverse routine. Both the carbon-cycle timescale and c are used to tune the modeled CO_2 over the past 800 kyr to match the EPICA ice-core record (EPICA community members, 2004) (Fig. 2). When 20-kyr running averages of both the simulation and this data are considered, the agreement is very good (root mean square error (RMSE) = 18 ppm, coefficient of determination $r^2 = 0.73$). However, also on the original 1000-year resolution, model and data show reasonable agreement (RMSE = 26 ppm, $r^2 = 0.59$); the model bias is then -3.9 ppm. For the observed δ^{18} O, we use the stacked record of Lisiecki and Raymo (2005), linearly interpolated with a 5-kyr running average to 100-year resolution and smoothed over six data points. The value chosen for c results in the best fit of our modeled δ^{18} O to Lisiecki and Raymo (2005), with a RMSE of 0.16‰ ($r^2 = 0.95$). Also when only considering the Pliocene (5 to 2.6 Myr ago), a different value for c does not lead to a better agreement with Lisiecki and Raymo (2005).

2.3. Boron isotope data

Most atmospheric CO₂ proxies suffer from large uncertainties, but the foraminiferal boron isotope based estimates are promising, since they show a good agreement with ice-core data during the Pleistocene (Hönisch et al., 2009). The δ^{11} B of surface dwelling planktic foraminifera is a function of seawater pH, which is in turn related to the CO₂ concentration in the mixed layer. We provide some new Plio-Pleistocene foraminiferal (*G. sacculifer*) δ^{11} B data (Extended Data Table). Our new dataset has a relatively low temporal resolution (on average 250 kyr), but covers a long period from 6.35 until 0.54 Myrs ago and thereby a wide range of CO₂ from 152^{+10}_{-9} to 507^{+46}_{-41} ppm.

2.3.1. Sample locations

Ocean Drilling Program (ODP) Site 1264 (28.53° S; 2.85° E, 2505 m water depth) is located on the central Walvis Ridge in the eastern sector of the South Atlantic subtropical Gyre. ODP Site 1264 is part of a depth transect along the shallow sloping northern flank of Walvis Ridge (Shipboard Scientific Party, 2004), which forms a prominent topographic feature within the Southeast Atlantic Ocean, separating the Angola Basin to the north and the Cape Basin to the south. Preservation of planktic foraminifera in the Plio-Pleistocene sections of Site 1264 is generally good. The age model (Bell et al., 2014) is based on tuning the benthic oxygen isotope record to the LR04 stack (Lisiecki and Raymo, 2005) (Suppl. Data Fig. 1). Today, the surface ocean CO₂ is in equilibrium with the atmosphere.

2.3.2. Analytical methodology

Roughly 50–90 G. sacculifer tests (\sim 20 mg per individual sample) were hand-picked from the 250-355 mm size fraction. In contrast to previous studies (Hönisch et al., 2009; Bartoli et al., 2011), a smaller size fraction had to be used since the sediments generally lacked sufficient numbers of large G. sacculifers. It has been suggested that smaller specimens of *G. sacculifer* are susceptible to carbonate dissolution. However, the picked average size, normalized with the weight of the shells of the samples shows no large changes over the record, indicating that dissolution is not biasing the record (Suppl. Data Fig. 2). The tests were crushed between glass plates and cleaned following the protocol of Barker et al. (2003). Cleaned samples were subsequently dissolved in 2 N HCl to yield sample solutions with approximately 1 ng of B/ml. Five to eight aliquots of 1 ml solution with 1 ml of boron-free seawater were loaded onto rhenium filaments. Analysis was performed on a Thermal Ionization Mass Spectrometer (Thermo Scientific TRI-TON) at Lamont Doherty Earth Observatory. Ionization temperature was between 980 and 1020 °C. Samples that showed isotopic fraction exceeding 1% over the acquisition time (~30 min) were excluded. The data are standardized against the SRM NIST 951 boric acid standard. All reported δ^{11} B values are based on at least three measurements. Standard errors reported are two internal errors of an in-house consistency standard or two internal errors of repeat analyses of individual sample solutions, if that was larger than the external reproducibility. Two standard errors (2 s.e.) range between 0.28 and 0.7% and average 0.33% (Extended Data Table).

2.3.3. Determination of pH from δ^{11} B of G. sacculifer

Boron isotope ratios in planktic foraminifera tests are a function of seawater pH. The relative abundance and isotopic composition of the two main dissolved boron species in seawater (borate and boric acid) changes with pH. Since marine carbonates preferentially incorporate the species borate, the boron isotope composition of the carbonate also changes with seawater pH. With a second parameter of the carbonate system (e.g. total alkalinity or carbonate ion concentration), atmospheric pCO_2^{atm} can be inferred from the pH values.

Ocean pH can be calculated from the δ^{11} B of the borate as follows:

$$pH = pK_B - \log \left[-(\delta^{11}B_{sw} - \delta^{11}B_{Borate}) / (\delta^{11}B_{sw} - \delta^{11}B_{Borate}(^{11,10}K_B - 1)) \right],$$
(6)

where pK_B is the equilibrium constant for the boric acid/borate system for a given temperature and salinity, $\delta^{11}B_{sw}$ is the isotopic composition of seawater (39.61%; Foster et al., 2010), $\delta^{11}B_{Borate}$ is the isotopic composition of the borate ion and K_B is the isotopic fractionation between the two aqueous species of boron in seawater (1.0272 ± 0.0006) (Klochko et al., 2006).

G. ruber Mg/Ca based SSTs for Site 1264 show no apparent trend over the past 5 Myr (Dekens et al., 2012). The reconstructed SSTs for the area in our climate model show a slight cooling trend over the Plio-Pleistocene (around 0.3 °C cooling per 1 Myr). We apply these temperatures estimates and a constant salinity of 36 psu in our calculations. We note that these variables have a minor affect on the calculated pH and pCO₂ (~30 ppm for a ± 3 °C change; $\pm \sim 10$ ppm for a $\pm 3\%$ salinity change).

We account for small long-term changes in the boron isotopic composition of seawater ($\delta^{11}B_{sw}$) by using a linear extrapolation between modern $\delta^{11}B$ (39.61‰, Foster et al., 2010) and the $\delta^{11}B_{sw}$ determined by Foster et al. (2010) for the middle Miocene (12.72 Myr ago, $\delta^{11}B_{sw} = 37.8\%$). This approach is consistent with Martínez-Botí et al. (2015a).

In order to calculate pH using the equation above, the δ^{11} B value of the foraminifera has to be corrected for size fraction effect (-2.25‰, Hönisch and Hemming, 2004), and further corrected for a species-specific difference between the δ^{11} B_{Borate} in ambient seawater and the δ^{11} B_{Calcite} of the foraminiferal tests. We use an empirical equation for *G. sacculifer* of Martínez-Botí et al. (2015b):

$$\delta^{11}B_{\text{Borate}} = (\delta^{11}B_{\text{Calcite}} - 3.6)/0.834.$$
⁽⁷⁾

The empirical calibration of Martínez-Botí et al. (2015b) is based on $\delta^{11}B$ datasets, combining results from MC-ICP-MS with N-TIMS data that were corrected for an analytical offset of 3.32‰. This offset between the two techniques can however not generally be applied. It has been demonstrated that the instrumental offset is matrix dependent (Foster et al., 2013) and can even vary for different foraminifera species (Hönisch et al., 2009). Here, we apply a correction offset of 0.9%, which is the average offset for foraminifera samples between measurements on the LDEO N-TIMS and the BIG MC-ICP-MS (Foster et al., 2013). The uncertainty in instrument specific offsets and the impact of matrix effects are certainly a major issue in the boron isotope analysis of marine carbonates that needs further investigation. However, it has also been demonstrated that relative differences in δ^{11} B in a sample set of a given matrix can be reconstructed regardless of the applied measurement technique (Foster et al., 2013). Using the corrections above we derive reasonable pH estimates from the Site 1264 samples for the well-constrained Pleistocene part. The uncertainty in pH is dominated by the uncertainty in the δ^{11} B measurement and is on the order of ± 0.04 pH units.

2.3.4. Determination of pCO_2^{atm} from $\delta^{11}B$ -derived pH

To estimate atmospheric pCO₂, a second parameter of the carbonate system is needed. Seki et al. (2010) have compared two different approaches. They reconstructed pCO₂ from modeled $[CO_3^{2-}]$ (Tyrrell and Zeebe, 2004) as well as assuming constant total alkalinity varying with only up to ±5%. The comparison between these



Fig. 3. Five-million-year time series of benthic δ^{18} O, CO₂, sea level and global temperature. (A) Simulated benthic δ^{18} O (green), (B) simulated CO₂ (red), with error margins based on simulations with increased Antarctic ablation (ABL) and fixed pre-industrial ocean overturning strength (OT) described in Section 3.2, (C) simulated sea level in meters above present day (blue), (D) simulated global mean temperature anomaly with respect to PI (T_{glob} ; grey). Black lines represent 400-kyr running averages. Highlighted in yellow are the Late Pliocene period 3.5 to 2.5 Myr ago and the Mid-Pleistocene Transition (MPT; 1.5 to 0.7 Myr ago) discussed in the main text.

approaches demonstrates that estimated pCO₂ is relatively insensitive to the second carbonate system parameter and is largely dependent on the recorded pH change as determined by δ^{11} B values. For our calculation we assume a constant total alkalinity of 2300 mmol/kg sea water. The uncertainty in the pCO₂ estimates in largely dominated by the analytical uncertainty in δ^{11} B. Taking into account additional uncertainties in estimated salinity, sea surface temperature and carbonate ion concentration we estimate the uncertainty in the reconstructed pCO₂ on the order of ±70 ppm in line with earlier studies (Bartoli et al., 2011).

3. Results and discussion

3.1. Five-million-year simulation

Our simulated global mean temperatures during glacials are typically 4 to 5 K below the pre-industrial average (PI), which is consistent with a data reconstruction of the Last Glacial Maximum (Annan and Hargreaves, 2013). In addition, modeled sealevel variability over the past five glacial cycles of 80 to 125 m



Fig. 4. Simulated CO_2 concentrations. The red line shows our simulated CO_2 record over the past five million years. To compare, the hybrid model-data reconstruction of Van de Wal et al. (2011) is shown (cyan line). Dashed lines represent 400-kyr running averages.

is in broad agreement with data records (e.g. Grant et al., 2014; Austermann et al., 2013). The modeled CO₂, sea level and global mean temperature records all show decreasing long-term trends over the past 5 Myr, while benthic δ^{18} O values gradually increase (Fig. 3). During the early Pliocene (5 to 3.3 Myr ago), global mean temperature is up to 1.7 K higher than PI, slightly lower than the 1.8 to 3.6 K range calculated by the PlioMIP GCM ensemble (Haywood et al., 2013).

Our simulation shows CO₂ concentrations of 300 up to 470 ppm during this period (Fig. 4, red line). These levels are considerably higher than found in an earlier reconstruction by Van de Wal et al. (2011) (Fig. 4, cyan line). In addition, our Pliocene CO₂ exhibits much larger shorter-term variability than this hybrid model-data reconstruction. From the beginning of the Pleistocene (2.5 Myr ago) onwards, the long-term averages of both records nearly coincide. They show a similarly weakly declining trend over the Mid-Pleistocene Transition (1.5 to 0.7 Myr ago), when power in the δ^{18} O spectrum shifts from 41 kyr to 100 kyr (Lisiecki and Raymo, 2005; Zachos et al., 2008; Bintanja and Van de Wal, 2008). Conversely, the higher variability in our simulation continues longer, lasting until the end of the Mid-Pleistocene Transition (0.8 Myr ago). Most prominently, our simulation shows more fiercely falling CO_2 levels during the M2 $\delta^{18}O$ excursion 3.3 Myr ago (415 to 200 ppm) and during the onset of periodic northern hemispheric glaciation 2.7 Myr ago (400 to 180 ppm).

The reconstruction by Van de Wal et al. (2011) used the northern hemispheric temperature record of De Boer et al. (2010), which was obtained using an inverse routine forcing the ice sheet model in stand-alone form without climate model. They inferred a constant log-linear relation between this record and several CO2 proxy data records. We now include a climate model in the set-up and derive CO₂ as a prognostic variable (Fig. 1, dashed blue line). Therefore, Earth System Sensitivity (ESS) is not a priori fixed in our model (Fig. 5). Instead, it is primarily influenced by ice sheetclimate interactions, which we capture in our coupled set-up. During the Pliocene, our simulated CO₂ levels are very variable and show a clear decreasing trend over time (Fig. 3). Meanwhile, icevolume equivalent sea level is far less variable; its long-term average remains virtually constant, slightly above its present-day value. During this time, in our model the climate is not cold enough for large scale glaciation of the Northern Hemisphere. At the same time, the Antarctic ice sheet has already reached its carrying capacity, the ice sheet size that the continent can maximally sustain (De Boer et al., 2010; Foster and Rohling, 2013). The CO₂ concen-



Fig. 5. Relation between global temperature anomalies and CO_2 . The relation between logarithmic CO_2 and global temperature perturbations with respect to their pre-industrial (PI) values (280 ppm and 287.7 K respectively) is clearly non-linear in our model.

trations thus vary between the thresholds for initiation of northern and southern hemispheric glaciation. Through the albedotemperature feedback, ice volume variability amplifies temperature perturbations, particularly in polar regions (Stap et al., 2014; Masson-Delmotte et al., 2013). Therefore, a reduction of ice volume variability during the Pliocene requires larger changes in CO₂ levels to obtain the same temperature fluctuations (Fig. 5). This implies that ESS is lower during the Pliocene than during the Pleistocene and Holocene in our model, whereas the constant relation between CO₂ and temperature in the record of Van de Wal et al. (2011) connotes the same ESS during these periods. The reduced ESS leads to higher simulated Pliocene CO₂ levels and larger CO₂ variability compared to Van de Wal et al. (2011), as well as compared to Hansen et al. (2013) who used a conceptual δ^{18} O-based climate model and did not take ice-sheet physics explicitly into account. A similar result as ours was obtained by Lunt et al. (2010), who found a reduction of ESS in warmer-than-PI climates (400 ppm CO₂) compared to colder-than-PI climates.

We note that in our model the radiative forcing of CO_2 is enhanced by a factor 1.3 to account for non- CO_2 greenhouse gasses (GHGs). This factor gives accurate results for the ice-core period (Stap et al., 2014), but an increase or decrease in the relative contribution of non- CO_2 is indeterminable in our model. Such a shift would need a compensating opposite change in CO_2 .

3.2. Sensitivity analysis

3.2.1. Influence of stability East Antarctica

In our reference experiment, we simulate a very stable Pliocene Antarctic ice sheet, leading to small variability in sea level. This is in agreement with earlier modeling studies (Huybrechts, 1993; Pollard and DeConto, 2009; De Boer et al., 2014) as well as some data studies (Denton et al., 1993). However, there are also other data suggesting that sea level was more variable during this time (Masson-Delmotte et al., 2013; Miller et al., 2012; Rohling et al., 2014; Cook et al., 2013). In a sensitivity experiment (ABL), we lower (in absolute sense) the ablation threshold parameter C_{abl} (Eq. (2)) for the East Antarctic ice sheet from -30 to -5

during the entire run. This altered value leads to ablation starting at lower temperatures and hence to a decreased glaciation threshold in our model. The initial tuning target of Antarctic glaciation starting at around 750 ppm CO_2 (Stap et al., 2014) is thus compromised. However, this glaciation threshold is debated and it is suggested to be model-dependent (Hansen et al., 2013; Gasson et al., 2014).

In the ABL run, there is still very little surface melt on East Antarctica during the past 2.7 Myr. Therefore, modeled sea level remains approximately the same as in our reference simulation. Conversely, during the Pliocene sea level now varies between 5 m below and 30 m above present; it reaches up to +20 m during the Late Pliocene (Fig. 6a, black line). This corresponds better to a recent multi-method proxy data reconstruction of peak sealevel height (Miller et al., 2012) than our reference-run sea level (Fig. 6a), red line). However, continuous high sea level, such as reconstructed by Rohling et al. (2014) (Fig. 6a, green line), cannot be reconciled with the δ^{18} O input by our model.

As a consequence of the increased amount of ice volume variability during this time leading to a strengthening of the albedotemperature feedback, we expect to find lower Pliocene CO₂ levels. Indeed, Pliocene CO₂ levels are reduced with respect to our reference (Fig. 6b). The difference is at most 70 ppm, but the average decrease over the period 5 to 2.7 Myr ago is only 28.5 ppm. The effect is relatively limited because, in our model, the grassland vegetation that replaces the retreated ice remains snow-covered throughout most of the year. The surface albedo reduction, which is the dominant effect of land ice on climate (Stap et al., 2014), is therefore small on the Antarctic continent. Hence, even if sea level variability is increased during the Pliocene, CO₂ concentrations remain significantly higher than reconstructed by Van de Wal et al. (2011). Alternatively, if EAIS variations are driven by marine-based instabilities as suggested by Pollard et al. (2015), the effect may be different, as this would not leave ice-free land when the EAIS retreats but rather open or sea-ice-covered ocean. Our one-dimensional SIA-based ice sheet model cannot reproduce such effects.

3.2.2. Influence of ocean overturning strength

In our reference run, the strength of the meridional ocean overturning is determined by the difference in temperature between polar and equatorial waters (Stap et al., 2014). To test the influence of this formulation, we conduct a separate run of the model where we keep overturning fixed at pre-industrial strength (run OT). In a similar way, Stap et al. (2014) inferred only little influence of the overturning strength on simulated temperature and ice volume during the past 800 thousand years. During the Pleistocene and Holocene (2.5 Myrs ago to PD), the effect of fixing the strength on modeled CO_2 is indeed also limited (Fig. 6c). The simulated CO_2 in OT is on average 4.3 ppm higher than in the reference run. During the Pliocene (5 to 2.5 Myrs ago), this difference increases to 19 ppm. In run OT, overturning strength no longer increases when the climate warms, as it does in the reference experiment. The consequent weaker downwelling leads to cooler deep-ocean temperatures. As compensation, higher CO₂ is simulated. The maximum difference between the long-term (400 kyr) running averages of both simulations is 33 ppm, and occurs during the early Pliocene. We conclude that the effect of increased ocean overturning strength on simulated CO₂ becomes important during climates significantly warmer than pre-industrial in our model. Moreover, the M2 δ^{18} O excursion 3.3 Myr ago is not fully captured in run OT (not shown), marking the importance of variable meridional ocean overturning during this event.

Although ocean circulation is allowed to change, our model is forced only by insolation and CO_2 . Independent changes in ocean circulation, for instance resulting from tectonic movement, are not



Fig. 6. Simulated sea level and CO₂ concentrations. (a) Modeled sea level over the Late Pliocene period 3.5 to 2.5 Myr ago and (b)–(d) modeled CO₂ over the past five million years. In red, our reference simulation; in black, simulations with increased Antarctic ablation (ABL), fixed pre-industrial ocean overturning strength (OT), and a smaller influence of deep-sea temperature on benthic δ^{18} O (GAM). The green line in panel (a) shows the sea level reconstruction of Rohling et al. (2014), interpolated to 100-year resolution. The blue triangles (Miller) in panel (a) represent a multi-method proxy data reconstruction of peak sea level (Miller et al., 2012), with error bars as indicated by that study. The thick dashed lines in panel (b)–(d) represent 400-kyr running averages.

incorporated. Furthermore, we do not take into account any vegetation changes. However, Foster and Rohling (2013) found that these processes only play a secondary role in long-term climate change over our simulated period.

3.2.3. Influence of relation between deep-sea temperature and δ^{18} O

The parameter γ (Eq. (3)), relating deep-sea-temperature to benthic δ^{18} O may be debated. Therefore, it is a factor of model uncertainty. In our reference run it is taken from a paleotemperature equation (Duplessy et al., 2002): 0.28‰ K⁻¹. However, Marchitto et al. (2014) suggested a lower value of 0.22‰ K⁻¹. We implement this lower value for γ in run GAM. In this run, larger changes in CO₂ with respect to Pl have to compensate the decreased effect of deep-sea temperature on δ^{18} O (Fig. 6d). Indeed, the CO₂ we simulate during the Pliocene is generally higher than our reference experiment (on average 8.4 ppm), and lower during the Pleistocene and Holocene (7.4 ppm). The long-term averages differ maximally 23.4 ppm. The model uncertainty imposed by the precise calculation of benthic δ^{18} O is therefore modest.

3.3. Comparison with existing and new proxy data

New foraminiferal boron isotope based CO₂ data is derived from Integrated Ocean Drilling Program (IODP) Site 1264 on the Walvis Ridge in the South Atlantic subtropical gyre (Section 2.3). This data is shown is Fig. 7. We compare our model results to a compilation of CO₂ records obtained from alkenones (Fig. 8A), and from foraminiferal δ^{11} B including this new data (Fig. 8B).



Fig. 7. New CO₂ data. New proxy-CO₂ data based on foraminiferal δ^{11} B, derived from Integrated Ocean Drilling Program (IODP) Site 1264 on the Walvis Ridge in the South Atlantic subtropical gyre.

Over the past 2 million years, the model results, as well as the new data, are largely consistent with Hönisch et al. (2009) and Seki et al. (2010). Although the RMSE of our simulation with respect to Hönisch et al. (2009) (43.6 ppm) is larger than the RMSE of Van de Wal et al. (2011) (26.4 ppm), the increased variability in our simulation during this time, demonstrated by standard deviation (SD) of 33.8 ppm to 20.2 ppm in Van de Wal et al. (2011), agrees better



B) alkenone-based data



Fig. 8. CO₂ model-data comparison. The red line shows our simulated CO₂ record. The cyan line shows the hybrid model-data reconstruction of Van de Wal et al. (2011). The black dots indicate our new δ^{11} B-based data, with error bars based on the standard deviation of repeated measurements. (A) Comparison with alkenone-based CO₂ data. Symbols indicate alkenone-based proxy CO₂ data (Zhang et al., 2013, orange triangles; Seki et al., 2010, lightgreen asterisks; Badger et al., 2013, blue pluses; Pagani et al., 2010, darkgreen crosses), with different error bars as indicated by these studies. (B) Comparison with boron-isotope-based CO₂ data. Symbols indicate previously published δ^{11} B-based proxy CO₂ data (Martínez-Botí et al., 2015a, orange triangles; Seki et al., 2010, lightgreen asterisks; Bartoli et al., 2011, blue pluses; Hönisch et al., 2009, darkgreen crosses), with different error bars as indicated by these studies.

with Hönisch et al. (2009) (SD = 38.9 ppm). However, the simulation varies with a larger frequency than is reconstructed by the data. Therefore, model-data comparison would benefit from a more extensive data record. The high CO_2 levels in the alkenone-based records of Zhang et al. (2013) and Pagani et al. (2010) are not supported by our model.

In the Late Pliocene period (3.5 to 2.5 Myr ago), our modeled CO₂ variability agrees more with the record of Martínez-Botí et al. (2015a), than with the stable CO₂ concentrations shown by Badger et al. (2013) (Fig. 9). However, the RMSE (76.2 ppm) and model bias (-46.1 ppm) with respect to Martínez-Botí et al. (2015a) are quite high, albeit smaller than those of Van de Wal et al. (2011) (RMSE = 87.1 ppm, bias = -73.1 ppm). The large simulated drop in CO₂ around 2.75 Myr ago is supported by the records of Martínez-Botí et al. (2015a) and Bartoli et al. (2011). Conversely, we do not model high CO₂ values around 2.9 Myr ago, where the



Fig. 9. CO_2 model-data comparison. Zoom-in on the Late Pliocene period 3.5 to 2.5 Myr ago, showing the data records with the highest resolution: Martínez-Botí et al. (2015a) (orange triangles), and Badger et al. (2013), blue pluses. The red line shows our simulated CO_2 record. The cyan line shows the hybrid model-data reconstruction of Van de Wal et al. (2011).

boron-isotope based records, as well as Pagani et al. (2010), agree upon. This most likely signifies a discrepancy between the benthic δ^{18} O record and the proxy CO₂ data.

Proxy CO_2 data are particularly scarce before 3.5 million years ago. Therefore, it is difficult to evaluate the reduced CO_2 variability in our simulation with respect to the Late Pliocene and Pleistocene. However, our higher CO_2 values than Van de Wal et al. (2011) seem to agree more favorably with the boron-isotope based records, including the new data, then with the alkenonebased record of Pagani et al. (2010).

4. Summary and conclusion

We have presented a continuous simulation of CO₂ over the past five million years. It is obtained using a coupled ice-sheet climate model, forced inversely by a stacked benthic δ^{18} O record (Lisiecki and Raymo, 2005). Therefore, the simulated CO₂ is in mutual agreement with modeled benthic δ^{18} O, global sea level and temperature. As such, the records capture our understanding of the interaction between CO₂, sea level and the climate.

Our results clearly show that the relation between CO_2 and global temperature that holds over the ice-core period cannot be extended into the Pliocene. During this time, a weakening of the albedo-temperature feedback with the absence of large Northern Hemisphere ice sheets reduces Earth System Sensitivity (ESS). Our results show more variable and generally higher CO_2 values during the Pliocene than an earlier study that hypothesized constant ESS (Van de Wal et al., 2011).

The model results are modestly affected by the ocean overturning strength, as well as by the amplitude of the deep-sea temperature effect on benthic δ^{18} O. Compared to the reference run, decreased strength of the overturning, as well as a weaker influence of deep-sea temperature, lead to smaller changes in benthic δ^{18} O at the same CO₂ concentrations. Hence, the simulated changes in CO₂ are larger.

In our reference simulation, the East Antarctic ice sheet (EAIS) is very stable during the Pliocene. When the ablation on the EAIS is increased, it is more dynamic, and consequently the Pliocene sea level is more variable. Peak sea level is then in better agreement with the multi-proxy synthesis of Miller et al. (2012). The increased sea level variability affects the simulated CO₂, but only to a relatively minor extent. This is explained by the ice-free land

remaining snow-covered throughout most of the year, resulting in relatively small changes of the surface albedo.

Our simulated CO₂ is in broad agreement with existing and new δ^{11} B-based proxy CO₂ data. Although RMSE and model bias remain large, these records are generally more in line with the modeled variability during the Late Pliocene and Early Pleistocene than alkenone-based CO₂ records. They also agree more with the higher CO₂ simulated during the Early Pliocene. This means that the CO₂ concentrations obtained from the δ^{11} B proxy are more easily reconcilable with the benthic δ^{18} O record.

For higher-than-PI levels of CO₂, the reconstruction of Van de Wal et al. (2011) is predominantly determined by seemingly low CO₂ values (400–500 ppm) documented by proxy data during the Middle Miocene. We attain these values already during the Pliocene, when benthic δ^{18} O is higher. Benthic δ^{18} O is approximately equally low during the Middle Miocene as during the Late Eocene, when much larger CO₂ concentrations are reconstructed by the same proxies. In future research, we will extent our simulation further back in time and investigate this apparent conundrum.

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Appendix A. Supplementary material

Supplementary material related to this article can be found online at http://dx.doi.org/10.1016/j.epsl.2016.01.022.

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