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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 years, 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

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 CO_2 that are mutually consistent. Our model shows CO_2 concentrations of 300 to 470 ppm during the Early Pliocene. Furthermore, we simulate strong CO_2 variability during the Pliocene and Early Pleistocene. These features are broadly supported by existing and new δ^{11} B-based proxy CO_2 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.

Keywords:

Carbon dioxide, global climate, global ice volume, sea level,

Plio-Pleistocene, proxy data

1 1. Introduction

The long-term interactions between CO_2 , temperature and sea level are 2 a topical issue in climate science. Recently, there have been a number of at-3 tempts to quantify these interactions by studying data from paleo archives. 4 For instance, CO_2 data of Antarctic ice cores and sea level reconstructions 5 from Red Sea sedimentary archives show a close linear correlation over the 6 past 516 ka (Foster and Rohling, 2013). However, an analysis of sea level 7 and temperature records spanning the Cenozoic has indicated a non-linear 8 relation between these variables (Gasson et al., 2012). In the pre-ice core 9 period, CO_2 and sea level data remain scarce. Moreover, uncertainties in 10

¹¹ CO₂ reconstructions are large and there is inter-proxy as well as intra-proxy
¹² disagreement (Masson-Delmotte et al., 2013; Beerling and Royer, 2011). This
¹³ limits either the scope or the skill of such reconciling studies.

Benthic for a more continuous δ^{18} O records currently provide a more continuous 14 and abundant data source on multi-million year timescales (Lisiecki and 15 Raymo, 2005; Zachos et al., 2008). A complicating factor, however, is the 16 interpretation of benthic δ^{18} O, because it comprises both an ice volume and 17 a deep-sea temperature component. To untangle their relative contributions, 18 two different approaches have been applied so far, namely (1) the use of inde-19 pendent deep-sea temperature proxies such as Mg/Ca of foraminiferal tests, 20 and (2) ice-sheet modeling. 21

In this study, we expand on the model-based approach to deconvolute the 22 δ^{18} O signal into temperature and sea level, including the simulation of CO₂. 23 We introduce an inverse routine to iteratively calculate CO_2 concentrations 24 over the past five million years from benthic δ^{18} O. The CO₂ is used to drive a 25 recently developed coupled ice sheet-climate model (Bintanja, 1997; De Boer 26 et al., 2010; Stap et al., 2014), which contains a scheme to calculate benthic 27 δ^{18} O. In earlier work, this coupled model has been shown to be capable of 28 reproducing glacial-interglacial cycles of ice volume and temperature over the 29 past 800 kyr in forward mode, using CO_2 from ice cores as input (EPICA 30 community members, 2004; Stap et al., 2014). We now force the model in-31 versely by a stacked benthic δ^{18} O record (Lisiecki and Raymo, 2005), which 32 enables us to study ice sheet-climate interactions in a broader range of cli-33 mates. This new integrated approach improves upon earlier studies using an 34 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 36 facilitates a better representation of the deep-ocean temperature, as well as 37 the simulation of seasonally-varying meridional-temperature profiles, rather 38 than annual mean and globally uniform temperature perturbations with re-39 spect to pre-industrial climate. Moreover, in these earlier studies informa-40 tion on CO_2 was lacking. Taking a hybrid model-data approach, Van de Wal 41 et al. (2011) obtained a continuous CO_2 reconstruction from a log-linear fit 42 between modeled temperature and proxy CO_2 data. Here, however, CO_2 is 43 incorporated in the model as a prognostic variable. Therefore, the simulated 44 CO_2 is mutually consistent with eustatic sea level (ice volume equivalent), 45 and with monthly mean atmospheric and oceanic temperatures, as deduced 46 from benthic δ^{18} O. This improves our understanding of the role of CO₂ in 47 climate variability. 48

We interpret our simulated CO_2 by studying the long-term relation be-40 tween CO_2 and global surface air temperature in our model, known as Earth 50 System Sensitivity (ESS), which is affected by the interaction between ice 51 sheets and climate. In addition, we test the sensitivity of our CO_2 simula-52 tion to the modeled strength of the meridional ocean overturning, as well as 53 to the formulation of the mass balance of East Antarctica and the relation 54 between deep-sea temperature and δ^{18} O. Finally, we compare our simulated 55 CO₂ to existing CO₂ proxy data (Hönisch et al., 2009; Seki et al., 2010; Bar-56 toli et al., 2011; Martínez-Botí et al., 2015a; Pagani et al., 2010; Zhang et al., 57 2013; Badger et al., 2013), complemented by a new δ^{11} B-based record. 58

⁵⁹ 2. Model and methods



 $_{60}$ 2.1. Coupled ice-sheet-climate model and benthic $\delta^{18}O$ calculation

Figure 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.

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We use a recently developed coupled climate-ice sheet model (Stap et al., 66 2014). In this coupled set-up, the climate component is represented by a zon-67 ally averaged energy balance climate model, developed by Bintanja (1997) 68 based on the model of North (1975). This climate model is tested for sensi-69 tivity to some important parameters in Bintanja (1997). It calculates surface 70 temperature in zonal belts of 5° latitudinal and one layer vertical resolution, 71 forced by 1000-yr resolution insolation (Laskar et al., 2004). It uses a ra-72 diative transfer scheme and parameterises energy transfer from the equator 73 towards the poles as a diffusive process. Surface albedo is determined by the 74 subdivision of the land surface into (potentially snow covered) grass, forest 75

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 CO_2 yielded by the inverse routine (Sec-82 tion 2.2). As discussed in Stap et al. (2014), the radiative forcing of CO_2 is 83 enhanced by a factor 1.3 to account for the influence of other greenhouse gases 84 $(CH_4 \text{ and } NO_2)$. The climate model is first run for 500 model years. There-85 after, a one-dimensional ice sheet model is run for the same 500 years. This 86 ice sheet model, described in detail in De Boer et al. (2010), obtains ice veloc-87 ities from the commonly used Shallow Ice Approximation (SIA). It calculates 88 ice volume and surface height change of the five hypothetical axisymmetrical 80 continents where the major ice sheets grow (North America (NaIS), Eura-90 sia (EuIS), Greenland (GrIS), East-Antarctica (EAIS) and West-Antarctica 91 (WAIS)), including the height-mass-balance feedback. The continents are 92 located at different latitudes and initially they are cone-shaped. Their differ-93 ent centre heights and slopes determine the maximum size and sensitivity to 94 temperature of the ice sheets. The mass balance routine is forced by monthly 95 temperatures (T) from the latitude in the climate model where the ice sheets 96 are located (Stap et al., 2014). Precipitation P is obtained based on the 97 Clausius-Clapeyron equation:

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

⁹⁹ where R is the radius of the ice sheet. P_0 and R_c are the present-day pre-

cipitation and critical radius respectively. An insolation-temperature 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 3.3.1 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.

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 .

						010 110	
	Parameter	Unit	EuIS	NaIS	GrIS	EAIS	WAIS
112	H_{cnt}	m	1,400	1,400	800	1,450	400
	s	-	-0.0000165*	-0.0000115*	-0.0014	-0.0010	-0.0011
	P_0	${\rm m~yr^{-1}}$	0.88	1.15	1.34	0.71	1.37
	R_c	km	1,500	1,800	750	2,000	700
	C_{abl}	-	-51	-41	-48	-30	-5
	β_T	$\% \mathrm{K}^{-1}$	0.35	0.35	0.35	0.6	0.8
	β_Z	$\% \mathrm{~km^{-1}}$	-6.2	-6.2	-6.2	-11.2	-11.2

Starred values indicate parabolic profiles, values given in m^{-1}

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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 parameterisation of benthic δ^{18} O values (De Boer et al., 2010) using the following equation:

$$\delta^{18}O = [\delta^{18}O_b]_{PD} - \frac{\overline{\delta^{18}O_i}V_i}{V_o} + \left[\frac{\overline{\delta^{18}O_i}V_i}{V_o}\right]_{PD} + \gamma\Delta T_o.$$
 (3)

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):

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

Here, β_T and β_Z are ice-sheet dependent parameters, that determine the in-128 fluence of annual mean temperature change (ΔT) and surface height change 129 (ΔZ) ; their values are the same as used by De Boer et al. (2010) (Table 1). 130 Present-day isotopic contents match the modeled values of an earlier study by 131 Lhomme et al. (2005). The final term on the right hand side of Eq. 2 quanti-132 fies the influence of deep-sea temperature change with respect to present day 133 (ΔT_o) . Gain factor γ is set to 0.28 $\% K^{-1}$, taken from a paleotemperature 134 equation (Duplessy et al., 2002). We assess the sensitivity of the model to 135 this value in Section 3.3.3. The deep-sea temperature perturbation is deter-136 mined from the climate model as the $40-80^{\circ}$ N mean of the second vertical 137

ocean layer, representative of the mid-latitude North Atlantic deep ocean. 138

The energy balance and 1-D ice-sheet models used in our study are less 139 comprehensive than current intermediate complexity (EMICs) and general 140 circulation models (GCMs), and 3-D ice sheet models. However, they have 141 the advantage of allowing for several five-million-year integrations of the cou-142 pled ice sheet-climate system, while capturing the relevant large-scale phys-143 ical processes, notably the interaction between ice sheets and climate (Stap 144 et al., 2014). 145



2.2. Inverse benthic δ^{18} O routine to calculate CO₂ concentration 146

Figure 2: Data-model comparison of CO_2 over the past 800 kyr. 148 Modeled CO_2 over the past 800 kyr (red), compared to the EPICA ice-core 149 record (EPICA community members, 2004) interpolated to 1000-yr temporal 150 resolution (blue). 151

¹⁵³ We use an inverse forward modeling approach to calculate CO₂ from ben-¹⁵⁴ thic δ^{18} O data. This is achieved by a two-step iterative routine. Each 1000-¹⁵⁵ year cycle starts with an update of the insolation input. At this first iteration ¹⁵⁶ step, a new CO₂ concentration is obtained from the difference between the ¹⁵⁷ modeled benthic δ^{18} O value and the observed value 500 years later:

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

The coupled model is run for 500 model years. Thereafter, as a second it-158 eration step Eq. 5 is applied again. The model is then run for another 500 159 years, using the updated CO_2 , but still forced by the same insolation. While 160 in principle this yields 500 year-resolution CO_2 , only the results after the 161 second iteration step are recorded and displayed in this paper. The temporal 162 resolution of the simulated CO_2 is therefore 1000 years. This is the desired 163 resolution, as the physics in our model are not detailed enough to capture 164 sub-millennial climate variations (Stap et al., 2014). We justify excluding 165 the intermediate CO_2 values, by running the model again in forward mode, 166 forced by the 1000-yr resolution simulated CO_2 record; this does not signifi-167 cantly alter the resulting climate and ice volume records. 168

In Eq. 5, $\overline{CO_2}$ is the mean CO_2 concentration of the preceding 15 kyr, 169 which reflects the long-term timescale of the carbon cycle. Together with 170 parameter c, which is set to $0.45\%^{-1}$, it determines the strength of the re-171 sponse of CO₂ to changes in δ^{18} O. While c is kept constant, it is important 172 to stress that a variable relation between $\delta^{18}O$ and CO_2 is ensured by the 173 carbon-cycle timescale, and most importantly by the second iteration step 174 in the inverse routine. Both the carbon-cycle timescale and c are used to 175 tune the modeled CO_2 over the past 800 kyr to match the EPICA ice-core 176

record (EPICA community members, 2004) (Fig. 2). When 20-kyr running 177 averages of both the simulation and this data are considered, the agreement 178 is very good (Root mean square error (RMSE)=18 ppm, coefficient of deter-179 mination $r^2=0.73$). However, also on the original 1000-yr resolution, model 180 and data show reasonable agreement (RMSE=26 ppm, r^2 =0.59); the model 181 bias is then -3.9 ppm. For the observed δ^{18} O, we use the stacked record of 182 Lisiecki and Raymo (2005)[LR04], linearly interpolated with a 5-kyr running 183 average to 100-year resolution and smoothed over six data points. The value 184 chosen for c results in the best fit of our modeled δ^{18} O to LR04, with a RMSE 185 of 0.16 $(r^2=0.95)$. Also when only considering the Pliocene (5 to 2.6 Myr 186 ago), a different value for c does not lead to a better agreement with LR04. 187 188

189 2.3. Boron isotope data

Most atmospheric CO_2 proxies suffer from large uncertainties, but the 190 for a miniferal boron isotope based estimates are promising, since they show 191 a good agreement with ice-core data during the Pleistocene (Hönisch et al., 192 2009). The δ^{11} B of surface dwelling planktic for a function of 193 seawater pH, which is in turn related to the CO_2 concentration in the mixed 194 layer. We provide some new Plio-Pleistocene foraminiferal (G.sacculifer) 195 δ^{11} B data (Extended Data Table). Our new dataset has a relatively low 196 temporal resolution (on average 250 kyr), but covers a long period from 6.35 197 until 0.54 Myrs ago and thereby a wide range of CO_2 from 152^{+10}_{-9} to 507^{+46}_{-41} 198 ppm. 199

200 2.3.1. Sample locations

Ocean Drilling Program (ODP) Site 1264 (28.53°S; 2.85°E, 2505 m water 201 depth) is located on the central Walvis Ridge in the eastern sector of the 202 South Atlantic subtropical Gyre. ODP Site 1264 is part of a depth tran-203 sect along the shallow sloping northern flank of Walvis Ridge (Shipboard 204 Scientific Party, 2004), which forms a prominent topographic feature within 205 the Southeast Atlantic Ocean, separating the Angola Basin to the north and 206 the Cape Basin to the south. Preservation of planktic foraminifera in the 207 Plio-Pleistocene sections of Site 1264 is generally good. The age model (Bell 208 et al., 2014) is based on tuning the benthic oxygen isotope record to the LR04 209 stack (Lisiecki and Raymo, 2005) (Suppl. Data Fig. 1). Today, the surface 210 ocean CO_2 is in equilibrium with the atmosphere. 211

212 2.3.2. Analytical methodology

Roughly 50-90 G.sacculifer tests ($\sim 20 \text{ mg per individual sample}$) were 213 hand-picked from the 250-355 mm size fraction. In contrast to previous 214 studies (Hönisch et al., 2009; Bartoli et al., 2011), a smaller size fraction had 215 to be used since the sediments generally lacked sufficient numbers of large G. 216 sacculifers. It has been suggested that smaller specimen of G. sacculifer are 217 susceptible to carbonate dissolution. However, the picked average size, nor-218 malized with the weight of the shells of the samples shows no large changes 219 over the record, indicating that dissolution is not biasing the record (Suppl. 220 Data Fig. 2). The tests were crushed between glass plates and cleaned 221 following the protocol of Barker et al. (2003). Cleaned samples were subse-222 quently dissolved in 2N HCL to yield sample solutions with approximately 1 223 ng of B/ml. Five to eight aliquots of 1 ml solution with 1 ml of boron-free 224

seawater were loaded onto rhenium filaments. Analysis was performed on a 225 Thermal Ionization Mass Spectrometer (Thermo Scientific TRITON) at La-226 mont Doherty Earth Observatory. Ionization temperature was between 980 227 and 1020 °C. Samples that showed isotopic fraction exceeding 1% over the 228 acquisition time (~ 30 minutes) were excluded. The data are standardized 229 against the SRM NIST 951 boric acid standard. All reported δ^{11} B values 230 are based on at least three measurements. Standard errors reported are two 231 internal errors of an in-house consistency standard or two internal errors of 232 repeat analyses of individual sample solutions, if that was larger than the 233 external reproducibility. Two standard errors (2 s.e.) range between 0.28 234 and 0.7% and average 0.33% (Extended Data Table). 235

236 2.3.3. Determination of pH from $\delta^{11}B$ of G.sacculifer

Boron isotope ratios in planktic foraminifera tests are a function of sea-237 water pH. The relative abundance and isotopic composition of the two main 238 dissolved boron species in seawater (borate and boric acid) changes with pH. 239 Since marine carbonates preferentially incorporate the species borate, the 240 boron isotope composition of the carbonate also changes with seawater pH. 241 With a second parameter of the carbonate system (e.g. total alkalinity or 242 carbonate ion concentration), atmospheric pCO_2^{atm} can be inferred from the 243 pH values. 244

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

$$pH = pK_B - log \left[-(\delta^{11}B_{sw} - \delta^{11}B_{Borate}) / (\delta^{11}B_{sw} - \delta^{11}B_{Borate}) \right],$$
(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 (~30ppm for $a \pm 3^{\circ}$ 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 $\delta^{11}B_{con}$ (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):

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

The empirical calibration of Martínez-Botí et al. (2015b) is based on $\delta^{11}B$ 269 datasets, combining results from MC-ICP-MS with N-TIMS data that were 270 corrected for an analytical offset of 3.32%. This offset between the two tech-271 niques can however not generally be applied. It has been demonstrated that 272 the instrumental offset is matrix dependent (Foster et al., 2013) and can even 273 vary for different foraminifera species (Hönisch et al., 2009). Here, we apply 274 a correction offset of 0.9‰, which is the average offset for foraminifera sam-275 ples between measurements on the LDEO N-TIMS and the BIG MC-ICP-MS 276 (Foster et al., 2013). The uncertainty in instrument specific offsets and the 277 impact of matrix effects are certainly a major issue in the boron isotope 278 analysis of marine carbonates that needs further investigation. However, it 279 has also been demonstrated that relative differences in δ^{11} B in a sample set 280 of a given matrix can be reconstructed regardless of the applied measure-281 ment technique (Foster et al., 2013). Using the corrections above we derive 282 reasonable pH estimates from the Site 1264 samples for the well-constrained 283 Pleistocene part. The uncertainty in pH is dominated by the uncertainty in 284 the δ^{11} B measurement and is on the order of ± 0.04 pH units. 285

286 2.3.4. Determination of pCO₂^{atm} from δ^{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 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 \pm 70 ppm in line with earlier studies (Bartoli et al., 2011).

300 3. Results and Discussion

301 3.1. Five-million-year simulation



³⁰³ Figure 3: Five-million-year time series of benthic δ^{18} O, CO₂, sea ³⁰⁴ 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 Sect. 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.

Our simulated global mean temperatures during glacials are typically 4 313 to 5 K below the pre-industrial average (PI), which is consistent with a data 314 reconstruction of the Last Glacial Maximum (Annan and Hargreaves, 2013). 315 In addition, modeled sea-level variability over the past five glacial cycles of 316 80 to 125 m is in broad agreement with data records (e.g. Grant et al., 2014; 317 Austermann et al., 2013). The modeled CO_2 , sea level and global mean 318 temperature records all show decreasing long-term trends over the past 5 319 Myr, while benthic δ^{18} O values gradually increase (Fig. 3). During the early 320 Pliocene (5 to 3.3 Myr ago), global mean temperature is up to 1.7 K higher 321 than PI, slightly lower than the 1.8 to 3.6 K range calculated by the PlioMIP 322 GCM ensemble (Haywood et al., 2013). 323



Figure 4: Simulated CO₂ concentrations. The red line shows our simulated CO₂ 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.

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Our simulation shows CO_2 concentrations of 300 up to 470 ppm during this 330 period (Fig. 4, red line). These levels are considerably higher than found 331 in an earlier reconstruction by Van de Wal et al. (2011) (Fig. 4, [W11] 332 cyan line). In addition, our Pliocene CO_2 exhibits much larger shorter-term 333 variability than this hybrid model-data reconstruction. From the beginning 334 of the Pleistocene (2.5 Myr ago) onwards, the long-term averages of both 335 records nearly coincide. They show a similarly weakly declining trend over 336 the Mid-Pleistocene Transition (1.5 to 0.7 Myr ago), when power in the 337 $\delta^{18}{\rm O}$ spectrum shifts from 41 kyr to 100 kyr (Lisiecki and Raymo, 2005; Za-338 chos et al., 2008; Bintanja and Van de Wal, 2008). Conversely, the higher 339

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 CO₂ levels during the M2 δ^{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).





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Figure 5: Relation between global temperature anomalies and CO₂.
The relation between logarithmic CO₂ 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.

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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 354 climate model. They inferred a constant log-linear relation between this 355 record and several CO_2 proxy data records. We now include a climate model 356 in the set-up and derive CO_2 as a prognostic variable (Fig. 1, dashed blue 357 line). Therefore, Earth System Sensitivity (ESS) is not a priori fixed in our 358 model (Fig. 5). Instead, it is primarily influenced by ice sheet-climate in-359 teractions, which we capture in our coupled set-up. During the Pliocene, 360 our simulated CO_2 levels are very variable and show a clear decreasing trend 361 over time (Fig. 3). Meanwhile, ice-volume equivalent sea level is far less 362 variable; its long-term average remains virtually constant, slightly above its 363 present-day value. During this time, in our model the climate is not cold 364 enough for large scale glaciation of the Northern Hemisphere. At the same 365 time, the Antarctic ice sheet has already reached its carrying capacity, the 366 ice sheet size that the continent can maximally sustain (De Boer et al., 2010; 367 Foster and Rohling, 2013). The CO_2 concentrations thus vary between the 368 thresholds for initiation of northern and southern hemispheric glaciation. 360 Through the albedo-temperature feedback, ice volume variability amplifies 370 temperature perturbations, particularly in polar regions (Stap et al., 2014; 371 Masson-Delmotte et al., 2013). Therefore, a reduction of ice volume vari-372 ability during the Pliocene requires larger changes in CO_2 levels to obtain 373 the same temperature fluctuations (Fig. 5). This implies that ESS is lower 374 during the Pliocene than during the Pleistocene and Holocene in our model, 375 whereas the constant relation between CO_2 and temperature in the record of 376 Van de Wal et al. (2011) connotes the same ESS during these periods. The 377 reduced ESS leads to higher simulated Pliocene CO_2 levels and larger CO_2 378

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 warmerthan-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 .





Figure 6: Simulated sea level and CO_2 concentrations. (a) Mod-393 eled sea level over the Late Pliocene period 3.5 to 2.5 Myr ago and (b-d) 394 modeled CO_2 over the past five million years. In red, our reference simu-395 lation; in black, simulations with increased Antarctic ablation (ABL), fixed 396 pre-industrial ocean overturning strength (OT), and a smaller influence of 397 deep-sea temperature on benthic $\delta^{18}O$ (GAM). The green line in panel (a) 398 shows the sea level reconstruction of Rohling et al. (2014)[R14], interpolated 399 to 100-year resolution. The blue triangles (Miller) in panel (a) represent a 400 multi-method proxy data reconstruction of peak sea level (Miller et al., 2012), 401 with error bars as indicated by that study. The thick dashed lines in panel 402 (**b-d**) represent 400-kyr running averages. 403

405 3.2.1. Influence of stability East Antarctica

In our reference experiment, we simulate a very stable Pliocene Antarc-406 tic ice sheet, leading to small variability in sea level. This is in agreement 407 with earlier modelling studies (Huybrechts, 1993; Pollard and DeConto, 2009; 408 De Boer et al., 2014) as well as some data studies (Denton et al., 1993). How-409 ever, there are also other data suggesting that sea level was more variable 410 during this time (Masson-Delmotte et al., 2013; Miller et al., 2012; Rohling 411 et al., 2014; Cook et al., 2013). In a sensitivity experiment (ABL), we lower 412 (in absolute sense) the ablation threshold parameter C_{abl} (Eq. 2) for the East 413 Antarctic ice sheet from -30 to -5 during the entire run. This altered value 414 leads to ablation starting at lower temperatures and hence to a decreased 415 glaciation threshold in our model. The initial tuning target of Antarctic 416 glaciation starting at around 750 ppm CO_2 (Stap et al., 2014) is thus com-417 promised. However, this glaciation threshold is debated and it is suggested 418 to be model-dependent (Hansen et al., 2013; Gasson et al., 2014). 419

In the ABL run, there is still very little surface melt on East Antarctica 420 during the past 2.7 Myr. Therefore, modeled sea level remains approximately 421 the same as in our reference simulation. Conversely, during the Pliocene sea 422 level now varies between 5 m below and 30 m above present; it reaches up 423 to +20 m during the Late Pliocene (Fig. 6a, black line). This corresponds 424 better to a recent multi-method proxy data reconstruction of peak sea-level 425 height (Miller et al., 2012) than our reference-run sea level (Fig. 6a, red line). 426 However, continuous high sea level, such as reconstructed by Rohling et al. 427 (2014) (Fig. 6a, [R14], green line), cannot be reconciled with the δ^{18} O input 428

429 by our model.

As a consequence of the increased amount of ice volume variability during 430 this time leading to a strengthening of the albedo-temperature feedback, we 431 expect to find lower Pliocene CO_2 levels. Indeed, Pliocene CO_2 levels are 432 reduced with respect to our reference (Fig. 6b). The difference is at most 70 433 ppm, but the average decrease over the period 5 to 2.7 Myr ago is only 28.5 434 ppm. The effect is relatively limited because, in our model, the grassland 435 vegetation that replaces the retreated ice remains snow-covered throughout 436 most of the year. The surface albedo reduction, which is the dominant effect 437 of land ice on climate (Stap et al., 2014), is therefore small on the Antarc-438 tic continent. Hence, even if sea level variability is increased during the 439 Pliocene, CO_2 concentrations remain significantly higher than reconstructed 440 by Van de Wal et al. (2011). Alternatively, if EAIS variations are driven 441 by marine-based instabilities as suggested by Pollard et al. (2015), the effect 442 may be different, as this would not leave ice-free land when the EAIS retreats 443 but rather open or sea-ice-covered ocean. Our one-dimensional SIA-based ice 444 sheet model cannot reproduce such effects. 445

446 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 454 CO_2 is indeed also limited (Fig. 6c). The simulated CO_2 in OT is on average 455 4.3 ppm higher than in the reference run. During the Pliocene (5 to 2.5 Myrs 456 ago), this difference increases to 19 ppm. In run OT, overturning strength 457 no longer increases when the climate warms, as it does in the reference ex-458 periment. The consequent weaker downwelling leads to cooler deep-ocean 459 temperatures. As compensation, higher CO_2 is simulated. The maximum 460 difference between the long-term (400 kyr) running averages of both simula-461 tions is 33 ppm, and occurs during the early Pliocene. We conclude that the 462 effect of increased ocean overturning strength on simulated CO_2 becomes 463 important during climates significantly warmer than pre-industrial in our 464 model. Moreover, the M2 δ^{18} O excursion 3.3 Myr ago is not fully captured 465 in run OT (not shown), marking the importance of variable meridional ocean 466 overturning during this event. 467

Although ocean circulation is allowed to change, our model is forced only by insolation and CO₂. Independent changes in ocean circulation, for instance resulting from tectonic movement, are not 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.

474 3.2.3. Influence of relation between deep-sea temperature and $\delta^{18}O$

The parameter γ (Eq. 3), relating deep-sea-temperature to benthic $\delta^{18}O$ 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 PI 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.



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

⁴⁹² New foraminiferal boron isotope based CO₂ data is derived from Inte-⁴⁹³ grated Ocean Drilling Program (IODP) Site 1264 on the Walvis Ridge in ⁴⁹⁴ the South Atlantic subtropical gyre (Sect. 2.3). This data is shown is Fig. ⁴⁹⁵ 7. We compare our model results to a compilation of CO₂ records obtained ⁴⁹⁶ from alkenones (Fig. 8a), and from foraminiferal δ^{11} B including this new ⁴⁹⁷ data (Fig. 8b).



Figure 8: CO_2 model-data comparison. The red line shows our simulated CO_2 record. The cyan line shows the hybrid model-data reconstruc-

tion of Van de Wal et al. (2011) [W11]. The black dots indicate our new 502 δ^{11} B-based data, with error bars based on the standard deviation of repeated 503 measurements. (a) Comparison with alkenone-based CO_2 data. Symbols in-504 dicate alkenone-based proxy CO_2 data (Zhang et al. (2013), orange triangles; 505 Seki et al. (2010), lightgreen asterisks; Badger et al. (2013), blue plusses; 506 Pagani et al. (2010), darkgreen crosses), with different error bars as indi-507 cated by these studies. (b) Comparison with boron-isotope-based CO_2 data. 508 Symbols indicate previously published $\delta^{11}B$ -based proxy CO_2 data (Martínez-509 Botí et al. (2015a), orange triangles; Seki et al. (2010), lightgreen asterisks; 510 Bartoli et al. (2011), blue plusses; Hönisch et al. (2009), darkgreen crosses), 511 with different error bars as indicated by these studies. 512

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Over the past 2 million years, the model results, as well as the new data, 514 are largely consistent with Hönisch et al. (2009) and Seki et al. (2010). Al-515 though the RMSE of our simulation with respect to Hönisch et al. (2009) 516 (43.6 ppm) is larger than the RMSE of Van de Wal et al. (2011) (26.4 ppm), 517 the increased variability in our simulation during this time, demonstrated 518 by standard deviation (SD) of 33.8 ppm to 20.2 ppm in Van de Wal et al. 519 (2011), agrees better with Hönisch et al. (2009) (SD=38.9 ppm). However, 520 the simulation varies with a larger frequency than is reconstructed by the 521 data. Therefore, model-data comparison would benefit from a more exten-522 sive data record. The high CO_2 levels in the alkenone-based records of Zhang 523 et al. (2013) and Pagani et al. (2010) are not supported by our model. 524



Figure 9: CO₂ 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 plusses. The red line shows our simulated CO₂ record. The cyan line shows the hybrid model-data reconstruction of Van de Wal et al. (2011) [W11].

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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 ⁵⁴¹ Martínez-Botí et al. (2015a) and Bartoli et al. (2011). Conversely, we do not ⁵⁴² model high CO₂ values around 2.9 Myr ago, where the boron-isotope based ⁵⁴³ records, as well as Pagani et al. (2010), agree upon. This most likely signifies ⁵⁴⁴ 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 alkenone-based record of Pagani et al. (2010).

551 4. Summary and conclusion

⁵⁵² We have presented a continuous simulation of CO₂ over the past five mil-⁵⁵³ lion years. It is obtained using a coupled ice-sheet climate model, forced in-⁵⁵⁴ versely by a stacked benthic δ^{18} O record (Lisiecki and Raymo, 2005). There-⁵⁵⁵ fore, the simulated CO₂ is in mutual agreement with modeled benthic δ^{18} O, ⁵⁵⁶ global sea level and temperature. As such, the records capture our under-⁵⁵⁷ standing of the interaction between CO₂, sea level and the climate.

⁵⁵⁸ Our results clearly show that the relation between CO_2 and global tem-⁵⁵⁹ perature that holds over the ice-core period cannot be extended into the ⁵⁶⁰ Pliocene. During this time, a weakening of the albedo-temperature feedback ⁵⁶¹ with the absence of large Northern Hemisphere ice sheets reduces Earth Sys-⁵⁶² tem Sensitivity (ESS). Our resuls show more variable and generally higher ⁵⁶³ CO_2 values during the Pliocene than an earlier study that hypothesised con-⁵⁶⁴ stant 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 571 stable during the Pliocene. When the ablation on the EAIS is increased, it 572 is more dynamic, and consequently the Pliocene sea level is more variable. 573 Peak sea level is then in better agreement with the multi-proxy synthesis of 574 Miller et al. (2012). The increased sea level variability affects the simulated 575 CO_2 , but only to a relatively minor extent. This is explained by the ice-free 576 land remaining snow-covered throughout most of the year, resulting in rela-577 tively small changes of the surface albedo. 578

⁵⁷⁹ Our simulated CO₂ is in broad agreement with existing and new δ^{11} B-⁵⁸⁰ based proxy-CO₂ data. Although RMSE and model bias remain large, these ⁵⁸¹ records are generally more in line with the modeled variability during the Late ⁵⁸² Pliocene and Early Pleistocene than alkenone-based CO₂ records. They also ⁵⁸³ agree more with the higher CO₂ simulated during the Early Pliocene. This ⁵⁸⁴ means that the CO₂ concentrations obtained from the δ^{11} B proxy are more ⁵⁸⁵ easily reconcilable with the benthic δ^{18} O record.

⁵⁸⁶ For higher-than-PI levels of CO₂, the reconstruction of Van de Wal et al. ⁵⁸⁷ (2011) is predominantly determined by seemingly low CO₂ values (400-500 ⁵⁸⁸ ppm) documented by proxy data during the Middle Miocene. We attain ⁵⁸⁹ 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.

⁵⁹⁴ Supplementary Data Figures



Supplementary Data Figure 1. Planktic foraminiferal boron isotope
based pCO₂ reconstructions from IODP Site 1264 (red dots) and the benthic
foraminiferal oxygen isotope record of Site 1264 (grey line, Bell et al. (2014)).



Supplementary Data Figure 2. Planktic foraminiferal δ^{11} B with analytical error bars versus foraminiferal shell weight measured after cleaning. No correlation between shell weight and δ^{11} B values suggests that the picked foraminifer shells used for δ^{11} B analysis are unlikely to be biased by dissolution.

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