Alkenone isotopes show evidence of active carbon concentrating mechanisms in coccolithophores as aqueous carbon dioxide concentrations fall below 7 $\mu mol L^{-1}$

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Alkenone isotopes show evidence of active carbon concentrating mechanisms in coccolithophores as aqueous carbon dioxide concentrations fall below 7 μ molL⁻¹

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5 Abstract.

Coccolithophores and other haptophyte algae acquire the carbon required for metabolic processes from the water in which they live. Whether carbon is actively moved across the cell membrane via a carbon concentrating mechanism, or passively through diffusion, is important for haptophyte biochemistry. The possible utilisation of carbon concentrating mechanisms also has the potential to over-print one proxy method by which ancient atmospheric CO_2 is reconstructed using alkenone

- 10 isotopes. Here I show that carbon concentrating mechanisms are likely used when aqueous carbon dioxide concentrations are below 7 μ molL⁻¹. I use published alkenone based CO₂ reconstructions from multiple sites over the Pleistocene, which allows comparison to be made with ice core CO₂ records. Interrogating these records reveal that the relationship between proxy- and ice core- CO₂ breaks down when local aqueous CO₂ concentration falls below 7 μ molL⁻¹. The recognition of this threshold explains why many alkenone based CO₂ records fail to accurately replicate ice core CO₂ records, and suggests the alkenone
- 15 proxy is likely robust for much of the Cenozoic when this threshold was unlikely to be reached in much of the global ocean.

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1 Introduction

Alkenones are long-chain (C_{37-39}) ethyl- and methy- ketones (Figure 1; Brassell et al. (1986); Rechka and Maxwell (1987)) produced by a restricted group of photosynthetic haptophyte algae (Conte et al., 1994). Produced by a narrow group of organisms which live exclusively in the photic zone, alkenones allow probing of algal biogeochemistry, and as alkenones are often preserved in the sedimentary record, alkenones can also provide information about past environmental conditions. Two main proxy systems based on alkenone geochemistry exist, (1) for sea surface temperature (SST) relies on the changing degree of unsaturation of the C_{37} alkenone ($U_{37}^{K'}$) (Brassell et al., 1986) and (2) for atmospheric CO₂, based on reconstructing the isotopic fractionation which takes place during photosynthesis (ε_p) using the carbon istopic composition of the preserved

25 alkenones (referred to throughout as $CO_{2(\varepsilon_p-alk)}$) (Laws et al., 1995; Bidigare et al., 1997).



Figure 1. Alkenones are C₃₇ unsaturated methyl ketones (Brassell et al., 1986; Rechka and Maxwell, 1987).

In the modern ocean alkenones are produced primarily by two dominant coccolithophore species; *Emiliania huxleyi* and *Gephyrocapsa oceanica*. *E. huxleyi* first appeared 290 kyr ago, and began to dominate over *G. oceanica* around 82 kyrs ago (Gradstein et al., 2012; Raffi et al., 2006). However alkenones are commonly found in sediments throughout the Cenozoic, with the oldest reported detections from mid-Albian aged black shales (Farrimond et al., 1986). Prior to the evolution of *G. oceanica* alkenones were most likely produced by other closely related species from the Noelarhabdacaeae family (Marlowe

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et al., 1990; Volkman, 2000).

Proxies for atmospheric CO_2 including $CO_{2(\varepsilon_p-alk)}$, those based on the $\delta^{11}B$ of planktic foraminifera, geochemical modelling and stomatal density, broadly agree that over the Cenozoic atmospheric pCO_2 declined from high levels (>1000 µatm) in the "greenhouse" worlds of the Paleocene and Eocene to close to modern day values (around 400 µatm) in the Pliocene

- 35 (Pagani et al., 2005, 2011; Pearson et al., 2009; Anagnostou et al., 2016; Foster et al., 2017; Sosdian et al., 2018; Super et al., 2018; Zhang et al., 2013; Beerling and Royer, 2011). However recently discrepencies have emerged between the alkenone and other CO_2 proxies at the moderate to low levels of atmospheric CO_2 of the Pleistocene (i.e. <400 µatm; Badger et al. (2019, 2013a) and compare Badger et al. (2013b) and Pagani et al. (2009) with Martínez-Botí et al. (2015)). Whilst the longstanding difference between alkenone (Pagani et al., 1999), $\delta^{11}B$ (Foster et al., 2012) and stomatal proxies (Kürschner et al., 2008) in
- 40 the Miocene CO₂ reconstructions appears to be resolved with new SST records (Super et al., 2018), differences remain in the Pliocene (Pagani et al., 2009; Badger et al., 2013b; Martínez-Botí et al., 2015) and Pleistocene (Badger et al., 2019).

One plausible reason for these discrepancies is the action of active carbon concentrating mechanisms (CCMs) in the haptophytes. These are potentially important as $CO_{2(\varepsilon_p-alk)}$ assumes purely passive uptake of carbon into the haptophyte cell (Laws et al., 1995; Bidigare et al., 1997). The potential for CCMs to be active and to effect $CO_{2(\varepsilon_p-alk)}$ has long been known

45 (Laws et al., 1997, 2002; Cassar et al., 2006) and recent work has refocussed efforts on understanding CCMs in $CO_{2(\varepsilon_p-alk)}$ (Bolton et al., 2012; Bolton and Stoll, 2013; Stoll et al., 2019; Zhang et al., 2019, 2020). In this study I use the now large number of published $CO_{2(\varepsilon_p-alk)}$ records which overlap with ice core records of atmospheric CO_2 (Tables 1 & 2) to explore the relationship between $CO_{2(\varepsilon_p-alk)}$ and CCMs in the Pleistocene, where our understanding of atmospheric CO_2 is best.

Table 1. Sites with Pleistocene $CO_{2(\varepsilon_p-alk)}$ records. Note that the MANOP Site C record was generated to track changes in surface water– atmosphere equilibrium not atmospheric pCO_2 , so although included here for completeness, is not included in the analysis

Site	Latitude	Longitude	Reference
ODP 999	12° 44.639' N	78° 44.360' W	Badger et al. (2019)
ODP 925	4° 12.249' N	43° 29.334' W	Zhang et al. (2013)
DSDP 619	27° 11.61' N	91° 24.54' W	Jasper and Hayes (1990)
MANOP Site C	0° 57.2" N	138° 57.3' W	Jasper et al. (1994)
NIOP 464	22° 15.4' N	63° 31.1'E	Palmer et al. (2010)
GeoB 1016-3	11° 46' S	11° 40' E	Andersen et al. (1999)
05PC-21	37° 30.11'N	129° 42.86'E	Bae et al. (2015)

 Table 2. Sources of ice core data, as compiled by Bereiter et al. (2015). WAIS - West Antarctic Ice Sheet, TALDICE - TALos Dome Ice

 CorE, EDML - EPICA Dronning Maud Land

Age interval (kyr)	Ice core location	Reference
-0.051 - 1.8	Law Dome	Rubino et al. (2013)
1.8–2	Law Dome	MacFarling Meure et al. (2006)
2–11	Dome C	Monnin et al. (2001, 2004)
11–22	WAIS	Marcott et al. (2014)
22–40	Siple Dome	Ahn and Brook (2014)
40-60	TALDICE	Bereiter et al. (2012)
60–115	EDML	Bereiter et al. (2012)
105–155	Dome C Sublimation	Schneider et al. (2013)
155–393	Vostok	Petit et al. (1999)

2 Materials and Methods

50 2.1 Calculating CO₂ from alkenone δ^{13} C: The CO_{2(ε_p -alk)} proxy

Multiple records of $CO_{2(\varepsilon_p-alk)}$ have been published for the Pleistocene (Figure 2, Table 1) allowing direct comparision with ice core based CO_2 records (Table 2). These records are globally distributed in longitude, but are concentrated at low latitude sites, largely as there is a general preference for sites which have (in the modern ocean) surface waters close to equilibrium with the atmosphere (Figure 2, Table 1). In longer term palaeoclimate studies there has also been a preference for low latitude,

55 gyre sites in the belief that these sites are more likely to be oceanographically stable over long time intervals (Pagani et al., 1999). Most of the records included here (Table 1, Figure 2) were generated with the aim to reconstruct atmospheric CO₂, however one, the MANOP C Site of Jasper et al. (1994), was used to explicitly reconstruct changing disequilibrium due to oceanographic frontal changes over time, and so is excluded from the following analysis.



Figure 2. Study sites relative to mean annual surface ocean CO_2 disequilibrium for 2005. Sites are globally distributed in longitude but restricted in latitude, as generally sites are choosen to be close to surface water equilibrium with the atmosphere. Sites used for this study are indicated, over the mean annual surface ocean disequilibrium for 2005 calculated from Takahashi et al. (2014). The MANOP C Site (Jasper et al., 1994) was choosen to study the disequilbrium at that site, so is shown here but not used in the following analyses. Site symbols are used throughout the figures: ODP 999 - circle, 05PC-21 - triangle, ODP 925 - inverted triangle, DSDP 619 - hexagon, MANOP Site C - square, NIOP 464 - star, GeoB 1016-3 - diamond.

Whilst these sites do only span a relatively small latitudinal extent, the diversity of settings does allow for investigation of any secondary controls on alkenone δ^{13} C (δ^{13} C_{alkenone}). In particular, differences in oceanographic setting and SST to test the hypothesis that low [CO₂]_(aq) breaks the relationship between δ^{13} C_{alkenone} and atmospheric CO₂, as might be expected if haptophytes are able to actively uptake carbon from seawater to meet metabolic demand – i.e. activate CCMs.

To facilitate fair comparision between sites and consistent comparision with the ice core records, all $CO_{2(\varepsilon_p-alk)}$ records I recalculated using a consistent approach. The approach is based on Bidigare et al. (1997) which updated the initial approach of Jasper and Hayes (1990) to $CO_{2(\varepsilon_p-alk)}$.

This approach removes some additional corrections used in the original publication of the records (such as growth rate adjustment for NIOP 464 (Palmer et al., 2010)) but does allow for direct comparison to be made.

An overview of how $CO_{2(\varepsilon_p-alk)}$ data are typically generated is given in Badger et al. (2013b).

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Briefly, to calculate ε_p requires the carbon isotopic composition of the dissolved CO₂ ($\delta^{13}C_{CO_{2(aq)}}$) and haptophyte biomass 70 ($\delta^{13}C_{org}$). The isotopic fractionation between $\delta^{13}C_{alkenone}$ and $\delta^{13}C_{org}$ is first corrected assuming a constant fractionation ($\varepsilon_{alkenone}$) of 4.2 % (Popp et al., 1998; Bidigare et al., 1997):

$$\varepsilon_{\text{alkenone}} = \frac{\delta^{13} C_{alkenone} + 1000}{\delta^{13} C_{org} + 1000} - 1 \tag{1}$$

The isotopic composition of dissolved inorganic carbon (DIC) is estimated using (ideally) the δ^{13} C of planktic foraminifera and the temperature-dependant fractionation between calcite and $[CO_2]_{(g)}$ experimentally determined by Romanek et al. (1992), where T is sea surface temperature in degrees Celsius (SST):

$$\varepsilon_{calcite-CO_{2(a)}} = 11.98 - 0.12T\tag{2}$$

The value the cabon isotopic composition of $CO_{2(g)}$ ($\delta^{13}C_{CO_{2(g)}}$) can then be calculated:

$$\delta^{13}C_{CO_{2(g)}} = \frac{\delta^{13}C_{carbonate} + 1000}{\varepsilon_{calcite-CO_{2(g)}}/1000 + 1} - 1000$$
(3)

From this $\delta^{13}C_{CO_{2(aa)}}$ can be calculated using the relationship experimentally determined by Mook et al. (1974):

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$$\varepsilon_{CO_{2(aq)}-CO_{2(g)}} = \frac{-373}{T+273.15} + 0.19$$
 (4)

and

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$$\delta^{13}C_{CO_{2(aq)}} = \left(\frac{\varepsilon_{CO_{2(aq)}} - CO_{2(g)}}{1000} + 1\right) \cdot \left(\delta^{13}C_{CO_{c(g)}} + 1000\right) - 1000$$
(5)

Finally $\varepsilon_{\rm p}$ can be calculated:

$$\varepsilon_p = \left(\frac{\delta^{13} C_{CO_{2(aq)}} + 1000}{\delta^{13} C_{org} + 1000} - 1\right).1000\tag{6}$$

and from that $[CO_2]_{(aq)}$ is calculated using the isotopic fractionation during carbon fixation (ε_f) and 'b', which represents the summation of physiological factors:

$$[CO_2]_{(aq)} = \frac{b}{\varepsilon_f - \varepsilon_p} \tag{7}$$

Here $\varepsilon_{\rm f}$ is assumed to be a constant 25 % (Bidigare et al., 1997). In the modern ocean the 'b' term, which accounts for physiological factors such as cell size and growth rate, shows a close correlation with $[PO_4^{3-}]$ (Bidigare et al., 1997; Pagani et al., 2009). However, the relationship between b, growth rate and $[PO_4^{3-}]$ has recently been questioned (Zhang et al., 2019,

2020) but for the purposes of this analysis is assumed to hold. This is discussed further below. Values for SST, $\delta^{13}C_{alkenone}$, $\delta^{13}C_{carbonate}$, salinity and $[PO_4^{3-}]$ are either taken from the original publications or estimated from modern ocean estimates (Takahashi et al., 2009; Antonov et al., 2010; Garcia et al., 2013; Locarnini et al., 2013).

Providing that the atmosphere is in equilibrium with surface water, the concentration of atmospheric CO_2 can be calculated 95 from $[CO_2]_{(aq)}$, (and vice versa if atmospheric CO_2 is known) using Henry's law:

$$pCO_2 = \frac{[CO_2]_{(aq)}}{K_{\rm H}}$$
(8)

The solubility coefficient (K_H) is dependent on salinity and SST, and here is calculated following the paramterization of Weiss (1970, 1974).

3 Results

100 3.1 Multi-site comparisons between $CO_{2(\varepsilon_p-alk)}$ and the ice core records

Across the six sites included in this analysis, there are 217 $CO_{2(\varepsilon_p-alk)}$ -based estimates of atmospheric CO_2 over the past 260 Ka for comparison with the ice core records (Table 2; Bereiter et al. (2015)). When all $CO_{2(\varepsilon_p-alk)}$ estimates are considered together over 260 Ka, this compilation of proxy-based records fails to replicate the ice core record (Figure 3). This has already been noted at specific sites (e.g. Site 999 in the Caribbean Badger et al. (2019)) but this is the first time that all available records

- 105 coincident with the Pleistocene ice core records have been compiled using a common methodology. Notably the $CO_{2(\varepsilon_p-alk)}$ based estimates are rarely lower than time-equivalent ice core estimate, but frequently higher. Given that haptophytes require carbon to satisfy metabolic demand, this is perhaps unsurprising; if at times of low carbon availabily haptophytes can switch from passive to active uptake to satisfy metabolic demand, it would be times of low atmospheric CO_2 (and so lower $[CO_{2(aq)})$ when the active uptake is most likely to be needed. As $CO_{2(\varepsilon_p-alk)}$ -based estimates of atmospheric CO_2 rely on the assumption
- 110 of a purely difusive uptake of carbon, it is therefore likely that the proxy would perform least well at times of low atmospheric CO_2 .

The haptophyse do not directly interact with the atmosphere, obtaining their carbon from dissolved carbon. As it is not only atmospheric CO_2 which controls the concentration of dissolved carbon ($[CO_2]_{(aq)}$), but also temperature, alkalinity and other oceanographic factors which control the equilibrium state between surface waters at the atmosphere, (Figure 2) the multiple

115 sites in different settings now give the opportunity to test whether other factors are important in controlling the accuracy of $CO_{2(\varepsilon_p-alk)}$.

To produce time-equilvalent estimates of atmospheric CO_2 for comparison with the ice core records, a simple linear interpolation of the Bereiter et al. (2015) compilation was initially used (Figure 4). This assumes that both the age model of the ice core and the published age models of the sites are correct and equivalent. This is almost certainly not the case, and so for the

120 calculations below, a ± 3000 year uncertainty is included for ages of both the ice core and $CO_{2(\varepsilon_p-alk)}$ values. Figure 4 shows that $CO_{2(\varepsilon_p-alk)}$ -based atmospheric CO_2 agree with ice core CO_2 at some sites and at some times, but not througout. Sites such as 05-PC21 (Bae et al., 2015) and DSDP Site 619 (Jasper and Hayes, 1990) perform quite well, througout, whilst others only appear to agree at higher values of CO_2 , such as ODP Site 999 (Badger et al., 2019) and NIOP 464 (Palmer et al., 2010),



Figure 3. Compiled $CO_{2(\varepsilon_p-alk)}$ -based estimates of atmospheric CO_2 over the past 260 Ka (blue circles), with the ice core compilation of Bereiter et al. (2015) shown as the solid black line. Full sources for the ice core and $CO_{2(\varepsilon_p-alk)}$ records are in Tables 2 and 1

whilst at some locations there is very little overlap between the two methods of reconstructing CO_2 (such as ODP Site 925 (Zhang et al., 2013) and GEoB 1016-3 (Andersen et al., 1999)).

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To explore whether $[CO_2]_{(aq)}$ is an imporant influence on $CO_{2(\varepsilon_p-alk)}$, I calculate predicted $[CO_2]_{(aq)}$ ($[CO_2]_{(aq)-predicted}$) for each of the samples. To calculate $[CO_2]_{(aq)-predicted}$, the ice core value of atmospheric CO_2 at the equivalent time interval is used in combination with Eq. 8 to calculate $[CO_2]_{(aq)}$ at the time of alkenone production. Reconstructed estimates of SST and salinity are used as for $CO_{2(\varepsilon_p-alk)}$ above, along with any estimated surface water-atmosphere disequilbrium. Points in Figure 4 are then coloured by $[CO_2]_{(aq)-predicted}$.

Inspection of Figure 4 suggests a connection between $([CO_2]_{(aq)-predicted})$ and the skill of $CO_{2(\varepsilon_p-alk)}$ to reconstruct atmospheric CO₂, with the points clustering around the 1:1 line lighter in colour (so with higher $[CO_2]_{(aq)-predicted}$), whilst points falling away from the 1:1 line have lower $[CO_2]_{(aq)-predicted}$.

To explore this relationship, I progressively restricted the included samples on the basis of $[CO_2]_{(aq)-predicted}$, and at each stage calculated a Pearson correlation co-efficient for each subset. Under this analysis the correlation co-efficient progressively increased as more of the low $[CO_2]_{(aq)-predicted}$ samples were excluded (Figure 5). All analyses were performed in R (R Core Team, 2020) using RStudio (RStudio Team, 2020).

This suggests that the fidelity of the $CO_{2(\varepsilon_p-alk)}$ depends on the concentration of $[CO_2]_{(aq)}$, improving at higher levels of $[CO_2]_{(aq)}$.

To further investigate this potential relationship, I progressively exclude samples based on $[CO_2]_{(aq)-predicted}$ with a step size of 0.05 µmolL⁻¹, again calculating Pearson correlation coefficients between ice core and $CO_{2(\varepsilon_p-alk)}$ for each subsample of the population. The result is shown in Figure 6. Here the analysis shows, similar to Figure 5, that as the samples with lowest $[CO_2]_{(aq)-predicted}$ are progressively removed, the correlation between ice core and $CO_{2(\varepsilon_p-alk)}$ increases. Furthermore, this



Figure 4. Crossplots of $CO_{2(\varepsilon_p-alk)}$ -based atmospheric CO_2 (y-axes) vs the time-equivalent estimate from ice core records (x-axes; Bereiter et al. (2015); Table 2)). The large panel compiles all sites, with the exception of MANOP Site C, as explained in the text. Symbols are coloured by predicted $[CO_{2(aq)}]$ for each site and time as explained in the text. Full sources for alkenone data are shown in Table 1. A 1:1 line is included in all plots for comparison.

continues only up until $[CO_2]_{(aq)-predicted}$ reaches 7 μ molL⁻¹. Above this, the correlation coefficient plateaus, until the subsample reaches such a small size that spurious correlations become important (Figure 6b).

3.2 Sensitivity and Uncertainty Tests

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As it is not impossible that that a similar pattern could emerge if the dataset were particularly shaped so that there was increased density surrounding the 1:1 correlation line, I ran a series of sensitivity experiments. In these, rather than reducing the sample by filtering by $[CO_2]_{(aq)-predicted}$, the whole dataset (Table 1) was randomly ordered, and then stepwise subsampled so that the number of samples equalled the number of values for each value of $[CO_2]_{(aq)-predicted}$ (ie for each point in Figure 6, an equivalently sized but randomly selected sample was made such that for any equivalent value of $[CO_2]_{(aq)-predicted}$ the randomly ordered sample had an equivalent n as shown in Figure 6b). Pearson correlation coefficients were calcluated for each subsample as above. To allow for possible age model uncertainties, a 3000 year (1σ) uncertainty was also applied to the age of each sample prior to sampling of the ice core record, and is



Figure 5. Crossplots of $CO_{2(\varepsilon_p-alk)}$ -based atmospheric CO_2 (Table 1; y-axes) vs the time-equivalent estimate from ice core records (x-axes; Bereiter et al. (2015); Table 2)). The sample of published vales of $CO_{2(\varepsilon_p-alk)}$ was progressively restricted by $[CO_2]_{(aq)-predicted}$, indicated by the subplot titles. Individual values are coloured by $[CO_2]_{(aq)-predicted}$, and Sites indicated by shape (see key). Pearson correlation coefficients and equations of best fit are shown in each panel, along with a 1:1 line.

- applied as a normally distributed uncertainty. Uncertainty in CO_{2(εp-alk)} measurements is typically calculated using Monte Carlo modelling of all the parameters (i.e Pagani et al. (1999); Badger et al. (2013a, b)), however this was not done in all the published work (Table 1), and some differences in approach was found accross the published work. Therefore to create CO_{2(εp-alk)} uncertainty estimates for each value in this study, I emulate the uncertainties based on the CO_{2(εp-alk)} value. I built a simple emulator (Figure 7) by running Monte Carlo uncertainty estimates for all of the included datasets (Table 1)
 using the same estimates of uncertainty for each variable in the CO_{2(εp-alk)} calculation as applied in Badger et al. (2013a, b).
- This then allows the uncertainty to be included in the $[CO_2]_{(aq)-predicted}$ calcuation as well as $CO_{2(\varepsilon_p-alk)}$, and allowed for uncertainty estimates to be site-ambivalent.

The result is shown in Figure 8, and suggests that the 7 μ molL⁻¹ break point remains valid. The absolute value of r^2 is reduced, even at higher $[CO_2]_{(aq)-predicted}$, but this would be expected given the addition of uncertainty in age model, as the

165 published age is most likely to align with the ice core. Given the rapid rate of change at deglaciations, this effect is likely to be particularly pronounced in this dataset as many records have high temporal resolution around deglaciations in order to attempt to resolve them. Any small age model offset introduced by the error modelling in these intervals also clearly has the potential



Figure 6. Pearson correlation coefficient (panel a) of a reducing sample of all compiled $CO_{2(\varepsilon_p-alk)}$ (Table 1) vs the time-equivalent estimate from ice core records (Bereiter et al. (2015); Table 2). The sample reduces stepwise by 0.05 µmolL⁻¹, and the number of records in each subsample is shown in panel b.

to induce large differences between the $CO_{2(\varepsilon_p-alk)}$ and ice core values. Figure 8 clearly demonstrates that it is the filtering by $[CO_2]_{(aq)-predicted}$ rather than any spurious correlations which determine the shape of the data in Figures 6 and 8.

170 4 Discussion

The plateau in r^2 in Figures 6a and 8a suggest that below a $[CO_2]_{(aq)-predicted}$ of ~ 7 μ molL⁻¹ $CO_{2(\varepsilon_p-alk)}$ is no longer as good a predictor of ice core CO_2 as when $[CO_2]_{(aq)-predicted} > 7 \mu$ molL⁻¹. This is clear from comparing the relationship between samples where $[CO_2]_{(aq)-predicted} < 7 \mu$ molL⁻¹ with those where $[CO_2]_{(aq)-predicted} > 7 \mu$ molL⁻¹ in Figure 9. Here the r^2 for the former of 0.15 is substantially less than the latter of 0.55. I suggest that this is because below this threshold, the

175 fundamental assumption of $CO_{2(\varepsilon_p-alk)}$; that carbon is passively taken up by haptophytes, no longer holds true. One obvious explanation for why this would be the case is that at low levels of $[CO_2]_{(aq)}$ haptophytes have to actively uptake carbon in order to satisfy metabolic demand.



Figure 7. Emulated uncertainty in $CO_{2(\varepsilon_p-alk)}$, generated by running Monte Carlo uncertainty models for all sites in Table 1 applying the same approach to uncertainty as Badger et al. (2013a, b). Estimates used in this study are highlighted in blue.

Similar behaviour has been recognised in some culture studies (Laws et al., 1997, 2002; Cassar et al., 2006), with some evidence that the diatom *Phaeodactylum tricornutum* even has a similar CCM threshold (Laws et al., 1997) but this study is the clearest evidence of the behaviour in alkenone based studies of the environment.

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By applying a threshold value for [CO₂]_{(aq)-predicted} of 7 µmolL⁻¹ to the published records (Table 1) values of CO_{2(εp-alk)} which are influence by active CCMs can be eliminated.Recognition of this new threshold value of [CO₂]_{(aq)-predicted} allows for a new record of Pleistocene CO_{2(εp-alk)} to be compiled. This compilation then much better replicates the glacial-interglacial pattern of CO₂ change over the last 260 Ka (Figure 10). Whilst this present compilation does rely on ice core CO₂ records to estimate [CO₂]_{(aq)-predicted}, and therefore has little direct utility as a CO₂ record, it does demonstrate that recognition of the CCM threshold allows accurate CO₂ reconstruction using CO_{2(εp-alk)}. This, and the behaviour shown in Figures 6a and 8a suggests the CCMs may effectively be considered either active or not, and that when [CO₂]_(aq) is plentiful passive uptake dominates, at least sufficiently in most oceanographic settings that CO_{2(εp-alk)} can accurately record atmospheric CO₂. This implies that if areas of the ocean (or intervals of time) with low [CO₂]_(aq) can be avoided, accurate reconstructions of atmospheric CO₂ can be acquired using CO_{2(εp-alk)}.

As $[CO_2]_{(aq)}$ is effected by both SST and atmospheric CO_2 , for $CO_{2(\varepsilon_p-alk)}$ to be effective in reconstructing atmospheric CO_2 , areas of warm water (i.e. tropical or shallow shelf regions) under relatively low atmospheric CO_2 must be avoided. However, as the atmospheric CO_2 control renders the global surface ocean sufficiently replete in $[CO_2]_{(aq)}$ at Pliocene-like levels



Figure 8. Pearson correlation coefficient of a reducing sample of all compiled $CO_{2(\varepsilon_p-alk)}$ (Table 1) vs the time-equivalent estimate from ice core records (Bereiter et al. (2015); Table 2). As in Figure 6 the sample reduces stepwise by 0.05 µmolL⁻¹. Panel a shows a 1000 member Monte Carlo analysis, whereby uncertainty in $CO_{2(\varepsilon_p-alk)}$ and age is considered, as detailed in the text. Panel b shows a similar 1000 member Monte Carlo analysis, but with random sampling of the whole $CO_{2(\varepsilon_p-alk)}$ population so that the number of samples is equivalent to the dataset shown in panel a, ie the size of the sample follows that shown in Figure 6b. Means and one σ uncertainties are shown as the bold lines.

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of atmospheric CO_2 and above (Martínez-Botí et al., 2015) at all but the warmest surface ocean temperatures, $CO_{2(\varepsilon_p-alk)}$ is likely to be a reliable system for most of the Cenozoic. It is only in the Pleistocene that atmospheric CO_2 is low enough for CCMs to be widely active accross the surface ocean, with the low CO_2 glacials providing the most difficulty (Badger et al., 2019). This finding aligns well with evidence that CCMs developed in coccolithophores as a reponse to declining atmospheric



Figure 9. Correlations between $CO_{2(\varepsilon_p-alk)}$ and ice core CO_2 where $[CO_2]_{(aq)-predicted} > 7 \ \mu molL^{-1}$ (black symbols) and $[CO_2]_{(aq)-predicted} < 7 \ \mu molL^{-1}$ (red symbols).

 CO_2 through the Cenozoic, and were developing in $[CO_2]_{(aq)}$ -limited parts of the ocean in the late Miocene at the earliest, and likely not widespread until the Plio-Pleistocene (Bolton et al., 2012; Bolton and Stoll, 2013).

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Recent has attempted to correct for the existance of CCMs in palaeo-records of atmospheric CO_2 (Zhang et al., 2019; Stoll et al., 2019; Zhang et al., 2020). However, these assume that CCMs are always active, and that Pleistocene records can be used to correct for them throughout the Cenozoic. If, as suggested by the analyses presented here, CCMs *only* act at low $[CO_2]_{(aq)}$, and largely only in conditions prevalent through the late Pliocene and Pleistocene, it is plausible that corrections based on Pleistocene records could over-compensate for CCMs in the rest of the Cenozoic, when the assumption of passive carbon uptake inherent in $CO_{2(\varepsilon_p-alk)}$ as traditionally applied may still be valid.

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5 Conclusions

Reconstructions of past atmospheric CO₂ with proxy tools like $CO_{2(\varepsilon_p-alk)}$ are critical to understanding how the Earth's climate system operates, so long as the tools used can be relied upon to be accurate and precise. This re-analysis of existing Pleistocene $CO_{2(\varepsilon_p-alk)}$ records reveals that below a critical threshold of $[CO_2]_{(aq)}$ of 7 μmoL^{-1} the relationship between $s^{13}C$

210 $\delta^{13}C_{alkenone}$ and atmospheric CO₂ breaks down, plausibly because below this threshold haptophytes are able to actively update carbon using CCMs in order to satisfy metabolic demand.



Figure 10. Revised compilation of Pleistocene $CO_{2(\varepsilon_p - alk)}$ vs ice core records. The compiled published records (Table 1) are shown as circles, coloured red where $[CO_2]_{(aq)-predicted}$ is below a threshold of $7 \mu moL^{-1}$, and blue where $[CO_2]_{(aq)-predicted} > 7 \mu moL^{-1}$. The solid blue line is a loess filter (span 0.1) through the $[CO_2]_{(aq)-predicted} > 7 \mu moL^{-1}$ values, with 95 % confidence intervals (dashed blue line). The black line is the ice core compilation of Bereiter et al. (2015) (Table 2).

Although reconstructing the low levels of atmospheric CO_2 in the Pleistocene glacials and areas of the global ocean where $[CO_2]_{(aq)}$ is less than 7 μmoL^{-1} will be impossible, for much of the Cenozoic the $CO_{2(\varepsilon_p-alk)}$ proxy retains utility. If care is taken to avoid regions and oceanographic settings where $[CO_2]_{(aq)}$ may be expected abnormally low, $CO_{2(\varepsilon_p-alk)}$ remains an important and useful proxy to understand the Earth system.

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