Clumped isotope thermometry in foraminifera as a tool in paleoceanography: New planktic and benthic data and constraints on non-thermal effects

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27 Key Points:

- 28 We report a foraminiferal temperature calibration: $\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1744 \pm 0.0154$; 29 n = 260; >2500 analyses).
- We apply the regressions to published Cenozoic data and show they yield more reasonable temperature and seawater δ^{18} O estimates.

34 Abstract

35 The carbonate "clumped" isotope thermometer (Δ_{47}) in foraminifera is increasingly being used to

- 36 reconstruct ocean temperature. Here we address several less understood aspects of the proxy using a large
- 37 dataset comprising new and reprocessed data. The Δ_{47} -temperature relationship in foraminifera (n = 260) is
- 38 described by $\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1744 \pm 0.0154$, and in inorganic calcite (n = 118) by $\Delta_{47} = 0.0392 \pm 0.0014 \ 10^6/T^2 + 0.1547 \pm 0.0165$. Dataset-related differences explain only 11% of the variance:
- 39 $0.0392 \pm 0.0014 \ 10^6/T^2 + 0.1547 \pm 0.0165$. Dataset-related differences explain only 11% of the variance; 40 non-thermal effects explain up to 13% of the variance. We address the paucity of benthic data, establishing

41 with more certainty that temperature sensitivity is indistinguishable from planktics and inorganic calcite.

42 The large benthic dataset resolves a previously uncharacterized correlation with $[CO_3^{2-}]$ that is small 43 $(\Delta\Delta_{47}/\Delta CO_3^{2-} \text{ slope} = 0.00019 \pm 0.00004, n = 66; R^2 = 0.315, p < 0.01)$. We report a multivariate regression

44 to account for both temperature and $\Delta[CO_3^2]$ for all benthics (epifaunal and infaunal), with $\Delta_{47} = 0.152 \pm$

45 $0.049 + 0.03865 \pm 0.00376 \times 10^{6}/T^{2} + 0.000150 \pm 0.0000601\Delta[CO_{3}^{2-}]$. We apply these regressions to

46 published Cenozoic benthic Δ_{47} data, with the multivariate benthic equation yielding temperature and $\delta^{18}O_{sw}$

47 values more consistent with independent proxies, models, and the broader understanding of ocean and

- 48 cryosphere dynamics under different conditions, including across the Eocene-Oligocene Transition and the
- 49 Early Eocene Climatic Optimum. In total, this work enables the application of clumped isotopes to
- 50 for a more realistic understanding of uncertainties.

51 Plain Language Summary

52 Carbonate clumped isotopes (Δ_{47}) is an emerging proxy for temperature. Several calibration studies have 53 examined the relationship between Δ_{47} and temperature in core-top for a but have used small 54 datasets. Here we measure new samples and restandardize published data for foraminifera and synthetic 55 calcite using best practices and explore controls on the proxy using results for 260 samples (>2500 56 analyses). We confidently demonstrate with a large dataset that for a forminiferal Δ_{47} shows a temperature 57 dependence like inorganic calcite, report a robust temperature calibration relationship, and discuss evidence 58 for a possible carbonate ion effect on benthic foraminifera that merits further examination. We then apply 59 the new calibrations to address some puzzling aspects of published Cenozoic benthic foraminiferal 60 temperature and $\delta^{18}O_{sw}$ reconstructions.

61 **1 Introduction**

62 Accurate constraints on past ocean temperatures are critical to understanding ocean-climate 63 interactions and Earth's climate history. However, employing proxies for temperature requires frameworks 64 for disambiguating thermal effects from chemical, biotic, and/or diagenetic imprints. Therefore, a large body of work has focused on calibrating and applying temperature proxies, including ¹⁸O/¹⁶O ratios (δ^{18} O) 65 in foraminifera (Zachos et al., 1994, 2001, 2008), Mg/Ca (Anand et al., 2003; Elderfield and Ganssen, 2000; 66 67 Lea et al., 1999; Haynes et al., 2023; Mashiotta et al., 1999; Nürnberg et al., 1996; Pak et al., 2004), the 68 alkenone unsaturation index (Conte et al., 2001, 2006; Müller et al., 1998; Sachs et al., 2000), and TEX₈₆ (Kim et al., 2008; Leider et al., 2010; Powers et al., 2010; Schouten et al., 2007; Tierney and Tingley, 2014). 69 70 Each of these approaches leverages a temperature tracer recorded in sedimentary material and compounds 71 produced by different types of organisms. Although each temperature proxy has been applied to provide 72 information on the history of the oceans (e.g., Bard, 2000; Bard et al., 1997; de Garidel-Thoron et al., 2005; 73 Koutavas, 2002; Lea et al., 2002; Martin and Lea, 2002), limitations exist, such as kinetic effects, 74 unconstrained species-specific vital effects, the need for regional and latitudinal specific calibrations, the 75 influence of changing seawater chemistry, and/or poor preservation potential.

76 More recently, a temperature proxy has emerged as a potentially transformative tool in the ocean 77 sciences is carbonate clumped isotope thermometry, which may be robust to some of these limitations 78 (Came et al., 2007; Ghosh et al., 2006; Leutert et al., 2019; Meckler et al., 2022; Meinicke et al., 2020; 79 Peral et al., 2018; Piasecki et al., 2019; Thiagarajan et al., 2011; Tripati et al., 2010, 2014). The clumped 80 isotope thermometer stands out because of its basis in thermodynamics (Ghosh et al., 2006; Schauble et al., 81 2006; Tripati et al., 2015). In calcite, aragonite, dolomite, other carbonates, and different chemical species of dissolved inorganic carbon (DIC), heavier isotopes of C and O (¹³C, ¹⁸O) are rare relative to the lighter 82 83 isotopes (¹²C, ¹⁶O). Carbonate "clumped" isotope geochemistry examines the proportion of these heavy 84 isotopes that are bound to each other forming heavy isotope "pairs" or clumps, for example, the multiply-85 substituted isotopologues ${}^{13}C^{16}O$ (the predominant species used for temperature reconstructions, to date) or ¹²C¹⁸O¹⁸O. 86

Since the first studies of clumped isotope systematics in foraminifera were published (Grauel et al.,
2013; Tripati et al., 2010), the use of the temperature proxy in paleoceanography has been limited because
of the large sample sizes needed for the analyses (Table 1), and the presence of interlaboratory offsets (see

Bernasconi et al., 2018, 2021; Dennis et al., 2011; Petersen et al., 2019; Tang et al., 2014 for discussion).
The work of multiple labs has been addressing these issues, including the use of consistent standardization
and more accurate methods for isotope ratio calculations (Dennis et al., 2011; Bernasconi et al., 2021;
Daeron et al., 2016; Upadhyay et al., 2021; Lucarelli et al., 2023) and sample size reduction (Grauel et al.,
2013; Upadhyay et al., 2021). Building on these advances, downcore records have been published for
foraminifera (e.g., Tripati et al., 2014; Leutert et al., 2019; Meckler et al., 2022).

96 However, recent work has highlighted the need for additional work to improve clumped isotope 97 calibrations for foraminifera, particularly for benthics (Daeron and Gray, 2023; Rohling et al., 2024). To 98 date, the exploration of potential chemical and biological biases has been limited, and the scope of non-99 thermal effects that we know confound other proxies are unresolved for clumped isotopes. For example, 910 studies have shown for oxygen isotopes and Mg/Ca ratios in foraminifera, carbonate ion effects can be 910 important (Bemis et al., 1998; Elderfield et al., 2006; Saenger and Evans, 2019; Spero et al., 1997).

102 Thus, in this study, we build on these developments in the systematics of other proxies to similarly 103 advance the foundation for more widespread use of clumped isotope thermometry in foraminifera as a tool 104 in paleoceanography. We recognize that a key aspect underpinning the nuanced application of temperature 105 proxies has been the development of large datasets (Anand et al., 2003; Elderfield et al., 2006), as well as 106 meta-analysis (e.g., Bemis et al., 1998; Daeron and Gray, 2023; Elderfield et al., 2006) to allow for the 107 intensive characterization of thermal and non-thermal effects to understand the circumstances in which 108 proxy application may be limited. Therefore, here we utilize current analytical methods and report new Δ_{47} 109 data for 124 core-top foraminiferal samples, including data on previously unstudied species. Next, we 110 reprocess data from published studies on foraminiferal samples (n = 136) using updated data handling 111 practices including use of the IUPAC parameter set (Daëron et al., 2016; Petersen et al., 2019), and 112 carbonate-based standardization onto the Intercarb-Carbon Dioxide Equilibrium Scale (I-CDES or I-113 CDES₉₀) reference frame (Bernasconi et al., 2021; Lucarelli et al., 2023; Upadhyay et al., 2021). These data 114 were generated on multiple models of mass spectrometer and carbonate preparation systems and together 115 allow us to report a novel combined dataset. We report a combined benthic foraminiferal dataset that is 116 composed of 42 monospecific samples and 25 mixed benthic foraminiferal samples. We conduct a meta-117 analysis to assess if there are interlaboratory related effects or other non-thermal effects, including evidence 118 for ecological, taxonomic, regional, or depth-dependent isotope effects. We derive calibrations for planktic 119 and benthic foraminifera (and different subsets of these taxa) and compare them to synthetic calcite. We 120 use this dataset to determine benthic multivariable regressions that account for a dependence of Δ_{47} on both 121 temperature and Δ [CO₃²]. These calibrations are used to reevaluate published reconstructions of deep ocean 122 temperature from δ^{18} O and Δ_{47} .

123 **1.1 Background on clumped isotope thermometry**

124 Carbonate clumped isotope thermometry has the potential in paleoceanography to be similarly 125 impactful as oxygen isotope thermometry. For over 50 years, the primary isotopic tool used to constrain past ocean temperatures has been the δ^{18} O of foraminifera (Shackleton and Opdyke, 1973; Zachos et al., 126 1994, 2001, 2008). For example, for aminiferal δ^{18} O has been used to provide critical constraints on how 127 128 temperature, ice volume, salinity, water column structure, and ocean circulation have responded to past 129 changes in greenhouse gas concentrations. However, a long-standing challenge in such applications has 130 been accurately partitioning the thermodynamic fractionation of oxygen isotopes in carbonates from the 131 effects of changing water δ^{18} O. In contrast, carbonate clumped isotope thermometry is not sensitive to water 132 δ^{18} O. All that is needed to determine mineral formation temperatures is the clumped isotope composition 133 of the solid, not the water from which it grew.

134 Clumped isotope thermometry relies on internal isotopic exchange between isotopes in a single 135 phase, instead of relying on an isotopic exchange reaction between different phases (e.g., $CaCO_3$ and H_2O 136 for the carbonate-water oxygen isotope thermometer). As with oxygen isotope thermometry, zero-point 137 energy differences between isotopic species (or isotopologues) form the basis of the clumped isotope 138 thermometer. For carbonate minerals that have grown in equilibrium, the paired measurement of clumped 139 isotopes and δ^{18} O in carbonate minerals can therefore yield both carbonate formation temperature and fluid 140 δ^{18} O. However, the low abundance of clumped (multiply-substituted) isotopologues ($^{13}C^{18}O^{16}O$ and 141 $^{12}C^{18}O^{18}O$, for example), and the low temperature sensitivity of the relevant isotope exchange reactions, 142 measurements need to be made with high precision and accuracy. Further, robust application of the 143 geothermometer necessitates mineral equilibrium being achieved and reliable quantification of temperature 144 dependencies, and if appropriate, corrections for kinetic isotope effects.

145 An initial study of clumped isotopes in planktic and benthic foraminifera and coccoliths was 146 published four years after the geothermometer was developed (Tripati et al., 2010). Using error-in-variable 147 regression models (a Deming regression), this study found no evidence for differences between synthetic 148 calcite and foraminifera calibrations, nor between cultured coccolithophores and inorganic calcite, and did 149 not find discernible taxon-specific fractionations (Tripati et al., 2010). Several calibration studies have 150 validated those initial findings and have also made instrumental advancements by reducing sample 151 requirements and using different methods and instrumentation for sample analysis as well as different 152 regression models (Meinicke et al., 2020; Peral et al., 2018; Piasecki et al., 2019). Multiple studies have 153 evaluated different models for estimation of seawater calcification temperature (e.g., Daëron and Gray, 154 2023; Tripati et al., 2010). Although foraminiferal measurements are challenging, there is a growing 155 literature of applied studies using clumped isotope thermometry to reconstruct ocean temperature and δ^{18} O 156 (e.g., Agterhuis et al., 2022; Evans et al., 2018; Leutert et al., 2019, 2020; Meckler et al., 2022; Tripati et 157 al., 2014; Taylor et al., 2023).

158 1.1.1 Challenges in applying clumped isotopes to foraminifera

159 While the growing body of work is demonstrating the power of clumped isotope thermometry when 160 applied to foraminifera, proxy use in paleoceanography is still limited because of several issues. Recent 161 major technological advances in applying this proxy to foraminifer have resolved the issue of large sample sizes (Table 1). While initially 3-8 mg CaCO₃ were used to perform a single replicate analysis and at least 162 163 three replicates required, with typical sample sizes of 9 to 30 mg (Ghosh et al., 2006; Tripati et al., 2010), 164 a major step forward has involved reduction in sample sizes that are similar to Mg/Ca analyses with 0.1 to 165 0.5 mg required for a single replicate, and 3 to 50 replicates, with typical total sample sizes of 2.5 to 5 mg (Table 1). These small sample measurements (total sample sizes of 2.5 to 5 mg, including replicates) have 166 167 been made on a MAT 253 isotope ration mass spectrometer (IRMS) with a kiel device (Grauel et al., 2013; 168 Meckler et al., 2014; Meinecke et al., 2020, Piasecki et al., 2019), the Nu-Perspective IRMS and NuCarb 169 or in-house carbonate reaction and purification devices (Anderson et al., 2021; Defliese and Tripati, 2020; 170 Upadhyay et al., 2021). These advances make employment of the proxy on foraminifera more feasible.

For carbonate-hosted tracers, it has also been important for the community to investigate inter-lab differences (Greaves et al., 2008; Rosenthal et al., 2004) and understand whether they are associated with sample preparation or with the specific analytical technique used (Barker et al., 2003, 2005; Bian and Martin, 2010; Martin and Lea, 2002). In clumped isotopes, interlaboratory offsets have been reduced through updated data handling practices including use of the IUPAC parameter set (Daëron et al., 2016; Petersen et al., 2019), and carbonate-based standardization onto the I-CDES reference frame (Bernasconi et al., 2021; Lucarelli et al., 2023; Upadhyay et al., 2021; Anderson et al., 2024).

178 **1.2 Overview of this study**

In this study, we leverage advances, best practices, and instrumentation that have been shown to yield highly precise, accurate, and intercomparable data (Bernasconi et al., 2021; Upadhyay et al., 2021; Lucarelli et al., 2023; Daeron and Gray, 2023). First, we utilize current analytical methods that have been shown to yield highly precise, accurate, and intercomparable data, with all data reported on the I-CDES reference frame (Bernasconi et al., 2021; Upadhyay et al., 2021; Lucarelli et al., 2023). We report Δ_{47} data for new foraminiferal samples (n = 124; 82 planktic foraminiferal samples and 42 benthic foraminiferal samples) which includes data on previously unstudied species. We reprocess data from published studies

- 187 combined dataset comprised of 260 core-top planktic and benthic foraminiferal samples, with 29 species
- and 2569 replicate analyses, from several major ocean basins (Table 1).
- 189 Figure 1. Locality information for core-top samples used in this study. Map shows sites and histogram
- 190 shows water depths. Site information is in Table S1.



191 We evaluate the use of multiple calcification temperature methods, including different δ^{18} O-based 192 calcification temperature equations. We conduct a meta-analysis to assess if there are interlaboratory related 193 effects or other non-thermal effects, including evidence for ecological, taxonomic, regional, or depth-194 dependent isotope effects, and show non-thermal effects can explain only 13% of the data variance. We 195 derive calibrations with significant data representation for both planktic and benthic foraminifera - where 196 previously relatively little benthic data has been published - and compare results to synthetic calcite. We 197 derive Δ_{47} -temperature calibrations for foraminiferal samples to compare material specific differences and 198 report limits for accuracy and precision of reconstructions. We validate prior findings of no significant 199 deviation from foraminiferal test Δ_{47} temperature sensitivity to inorganic calcite precipitates (118 samples; 200 641 analyses; grown in experiments at temperatures between 4 and 50 °C; Table 2). The analysis of benthic 201 foraminiferal data indicates a weak carbonate ion effect, and we report a new multivariate regression 202 relating Δ_{47} to both temperature and carbonate saturation. Finally, we apply our novel calibrations. We 203 analyze profile temperature reconstructions and show that modern hydrographic profiles are reasonably 204 constrained using data from multiple taxa, show the impact of the use of our new calibrations on 205 reconstructed temperatures on Cenozoic temperatures, and make recommendations for future studies.

206 2 Materials and Methods

207 2.1 Sample and locality information

208 Samples are from the four major ocean basins (Arctic, Atlantic, Indian, Pacific), the Gulf of 209 Mexico, and the Mediterranean Sea (Figure 1, Table S1). Figure 1 shows locality information for the 49 210 sites. A total of 29 foraminifera taxa were studied (13 benthic, 15 planktic), and assemblages of mixed 211 benthic species. The species included in this study live in a range of habitats, spanning Atlas temperatures

212 from -1.5 °C to 29.6 °C (Levitus et al., 2010). Data for 124 foraminifera samples are newly reported and

- 213 were run in the Eagle-Tripati clumped isotope laboratory at UCLA, while 136 samples from five previous
- publications (Breitenbach et al., 2018; Meinicke et al., 2020; Peral et al., 2018; Piasecki et al., 2019; Tripati
 et al., 2010) were projected into the I-CDES reference frame for this work (Table 1, Table S2).
- et al., 2010) were projected into the I-CDES reference frame for this work (Table 1, Table S 216

217 Table 1. Sources of clumped isotopic data from foraminifera used in this study. Data for 260 core-top 218 foraminifera are reported (based on 2569 analyses), comprised of 124 new foraminifera samples measured 219 for this study, and 136 samples reported in prior publications (Tripati et al., 2010; Peral et al., 2018; 220 Breitenbach et al., 2018; Meinicke et al., 2020; Piasecki et al., 2019). The dataset from Grauel et al. (2013) 221 was excluded based on the PI from the measuring laboratory not being confident in the standardization of 222 the dataset. Coccolith bulk measurements from Tripati et al. (2010) were not included in this study. 223 Foraminifera calcification temperatures are recalculated with multiple methods as described in section 224 2.4.2. All published datasets were updated to the same standardization reference frame (I-CDES₉₀).

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Publication	Ecology	# samples	# of analyses	Mass/replicate (mg)	# replicates	Integration time (seconds)
Tripati et al., 2010	planktic, benthic.	34	58	4-8	1-5	640-1280
Grauel et al., 2013	planktic, benthic	42	618	0.15-0.2	2-47	156
Peral et al., 2018	planktic, benthic	27	248	2-3	4	480
Breitenbach et al., 2018	planktic	19	158	0.12-0.24	1-12	208
Meinicke et al., 2020	planktic	43	830	0.1-0.13	15-30	400
Piasecki et al., 2019	benthic	13	582	0.1-0.4	1-43	400
This study New measurements	planktic, benthic	124	720	0.25-0.52	3-20	1200
This study Foraminifera meta-analysis	planktic , benthic	260	2569			

226

227 2.2 Methods for new samples measured in the Eagle-Tripati laboratory

228 2.2.1 Sample Preparation

229 Core-top sediment samples from 47 different sites were suspended in deionized (DI) water for a 230 minimum of 24 hours using a rotating wheel (Table S1). Fine and coarse fraction were separated using a 231 63 μ m sieve and dried overnight at 30 °C (\leq 50 °C to prevent bond reordering). A total of 16 different taxa, 232 both planktic and benthic, and mixed benthic species, were hand-picked from different size fractions at 233 these sites. Tests were translucent and intact. We aimed for a minimum weight of picked foraminifera of 3 234 mg to have sufficient material for several replicate analyses.

235 Picked samples underwent a cleaning procedure to remove potential contaminants, adapted from 236 the cleaning procedure for Mg/Ca analysis from Barker et al. (2003), as described in Tripati et al. (2010), 237 with a focus on removing infill and adhering sediment. Briefly, foraminiferal shells are cracked open, rinsed 238 and ultrasonicated using methanol and DI water until the cleaning water is clear. Each ultrasonication step 239 consisted of 4 intervals lasting 15 seconds to ensure that the water did not heat up. Samples were then 240 placed inside an oven to dry overnight at 30 °C (< 50 °C). Samples were homogenized using a mortar and 241 replicates were weighed and stored in a desiccator until the day of analysis. Additional details on sample 242 preparation are provided in the Supplement (S2).

243 2.2.2 Measurement Procedure

Aliquots of cleaned samples that ranged from 0.46 to 0.52 mg were measured in single analyses, with 3 to 20 replicates measured per sample. A total of 720 new analyses of 124 foraminifera samples were made. Samples were analyzed from 2017 to 2023 on two Nu Perspective Dual inlet mass spectrometers with secondary electron suppression. These were run on instrument configurations "2" (Nu Perspective

- 248 2014, acid bath 90 °C), "3A" (Nu Perspective 2014, Nu Carb, 70 °C) and "3B" (Nu Perspective 2016, Nu
- Carb, 70 °C) that were shown in Upadhyay et al. (2021) and Lucarelli et al. (2023) to yield statistically indistinguishable results from each other.

251 Every 4 to 5 sample measurements, a carbonate standard was measured and Δ_{47} values are reported 252 on the Inter-Carbon Dioxide Equilibrium Scale (I-CDES₉₀, Bernasconi et al. 2021) at 90 °C. A total of 253 seven standards were used, including ETH-1, ETH-2, ETH-3, ETH-4, CM Tile, Carmel Chalk, and 254 Veinstrom (Upadhyay et al., 2021; Lucarelli et al., 2023). ETH-1 and ETH-2 were used for non-linearity 255 corrections while ETH-1, ETH-2, and ETH-3 and the remaining in-house standards were utilized for δ^{13} C 256 and δ^{18} O drift corrections and empirical transfer function (ETF) calculations. ETH-4 was used as a 257 consistency standard and not included in any standard corrections. Depending on standard drift during a 258 correction interval, we utilized either 10 or 20-point moving averages for drift, nonlinearity, and empirical 259 transfer function corrections for each correction interval, following published studies (Meckler et al., 2014; 260 Upadhyay et al., 2021; Lucarelli et al., 2023). The Brand parameter set was used for calculations (Daëron et al., 2019; Petersen et al., 2019). Calculations to determine Δ_{47} , δ^{13} C, and δ^{18} O use publicly available 261 262 software (Easotope - John and Bowen, 2016) and further details on clumped isotopes are provided in the 263 Supporting Information (Section S1).

264 2.2.3 Accuracy and precision of data and data archiving

265 Standard values for each instrumental configuration are typically within 0.005‰ of values reported 266 by Bernasconi et al. (2021) as described in Upadhyay et al. (2021) and Lucarelli et al. (2023). Long-term 267 absolute mean error in standard Δ_{47} across instruments is 0.0002% (or 0.03% in percent terms), and is 268 normally distributed (Upadhyay et al., 2021; Lucarelli et al., 2023). The average Δ_{47} standard deviation (1 269 s.d.) among standard replicates for all instruments is 0.021‰ (Upadhyay et al., 2021; Lucarelli et al., 2023). 270 Similar values are reported for samples. For samples, average Δ_{47} reproducibility is 0.022‰ (1 s.d.) 271 and 0.009‰ (1 s.e.) Supporting Information Table S1). The ranges for samples are 0.001-0.049‰ (1 s.d.) 272 and 0.000-0.029‰ (1 s.e.). Standard and sample data for new measurements are provided using the 273 recommended EarthChem template (Petersen et al., 2019) and archived online with manuscript acceptance.

274 2.2.4 Quality control

275 Data quality control followed published procedures described in Parvez et al. (2023). Although in 276 the laboratory, samples are routinely screened based on Δ_{48} or Δ_{49} values that are indicative of high organic 277 content, all samples reported for this study fell within initial screening bounds, so no replicates were excluded based on this criteria. Replicates with incomplete gas transfer and Δ_{47} , δ^{13} C, and δ^{18} O values that 278 fall outside of a SD range of $\pm 0.5\%$ for δ^{13} C and δ^{18} O and $\pm 0.05\%$ for Δ_{47} are also flagged. Similar 279 280 screening thresholds for Δ_{47} were reported by Meckler et al. (2014), Upadhyay et al. (2021), and Lucarelli 281 et al. (2023). Data were screened based on instrument source stability, pumpdown times and leaks in the 282 acid digestion system, and standard data quality during correction intervals as described in Parvez et al. 283 (2023). Of the new analyses conducted for this study, 720 replicates were included, and 52 replicates 284 excluded.

285 2.3 Reprocessed datasets from publications

286 2.3.1 Core-top foraminiferal datasets

Five published foraminiferal datasets (Breitenbach et al., 2018; Meinicke et al., 2020; Peral et al., 2018; Piasecki et al., 2019; Tripati et al., 2010) were reprocessed onto the I-CDES₉₀ reference frame as part of this study. The foraminiferal data include 24 different taxa (10 benthic, and 14 planktic) and mixed benthic samples from Tripati et al. (2010) and Piasecki et al. (2019) (Table 1, S2, S3). After discussion with

291 S. Bernasconi about Grauel et al. (2013) and the reprocessing of their dataset, data from the study were not

- included, as the measurements only included the use of two marble (high-temperature) standards with low Δ_{47} values, with no carbonate standard with high Δ_{47} values used for their corrections.
- 294 2.3.2 Synthetic calcite datasets

We compare our foraminiferal data to synthetic calcite to assess if there are differences between the two groups. Since growth temperatures for foraminiferal samples occupy a much narrower range of temperature than many synthetic calcite precipitates used for calibration purposes, we limit the data used for synthetic calcite to 4 to 50 °C.

299 A total of 118 synthetic calcite samples comprising 641 replicates were utilized for this comparison: 300 we include 11 new measurements of synthetic calcite analyzed in the same manner as foraminiferal samples 301 on the I-CDES₉₀ reference frame (Table 2, Supporting Information Table S4), 19 samples were reprocessed 302 as part of this study (Defliese and Tripati, 2020; UCLA samples from Tang et al., 2014; Tripati et al., 2015) 303 and 85 additional synthetic calcite samples from other studies (Table 2, Supporting Information Table S4). 304 Synthetic calcite calibration samples from other studies include samples that were reprocessed using Brand 305 parameters in the Petersen et al. (2019) study (Defliese and Lohmann, 2015; Kelson et al., 2017; Kluge et 306 al., 2015; Tulane samples from Tang et al., 2014), as well as Anderson et al. (2021), Lucarelli et al. (2023), 307 Jautzy et al. (2020), and Swart et al. (2021).

Table 2. Sources of clumped isotopic data from synthetic calcite used in this study. Data for 118 samples

309 of synthetic calcite grown from 0 to 50 °C are reported (based on 641 analyses), comprising 11 new samples

- 310 measured for this study, and 104 samples reported in prior publications. All published datasets were updated
- 311 to the same standardization reference frame (I-CDES₉₀).
- 312

Publication	# samples	# analyses	Mass/replicate (mg)	# replicates	Integration time (seconds)
Anderson et al., 2021	10	87	0.4 - 0.6	8-10	1800
Defliese et al., 2015	3	12	4-6	4	640
Defliese and Tripati 2015	2	19	0.25-0.5	7-12	640-1600
Jautzy et al., 2020	4	85	0.08 - 0.1	19-24	700
Kelson et al., 2017	30	108	6-9	2-7	1560
Kluge et al., 2015	6	18	5 – 8	1-5	1456
Lucarelli et al., 2022	4	36	0.48-0.52	5-12	1200
Swart et al., 2021	6	73	5-8	3-14	2436
Tang et al., 2014	32	101	25	1-24	1200
Tripati et al., 2015	7	13	5 – 8	1-4	1200
This study					
New measurements	11	89	0.48 - 0.52	6-13	1200
This study Synthetics meta-analysis	115	641			

313 2.3.3 Data reprocessing for published datasets

314 The Tripati et al. (2010) dataset for foraminifera and Tripati et al. (2015) dataset for synthetic calcite 315 were reprocessed for use in this study in the Easotope software (John and Bowen, 2016) with updated 316 parameter values (Daëron et al., 2016) and standard values (Bernasconi et al., 2021; Lucarelli et al., 2023) 317 on the CDES absolute reference frame. Gas standards are used for the nonlinearity correction step and a 318 combination of gas and carbonate standards for the ETF to project onto the CDES reference frame (see 319 methods section 2.1.2). We use these new Δ_{47} (CDES) values, along with the transfer functions (Supporting 320 Information Table S5) constructed from Bernasconi et al. (2021) to project values into the I-CDES₉₀ 321 reference frame.

We also reprocessed data from the Petersen et al. (2019) study (Defliese et al., 2015; Kelson et al., 2017; Kluge et al., 2015; Tang et al., 2014). These datasets were initially brought onto the CDES reference frame at 25 °C (Dennis et al., 2011) using ETH carbonate standards (Bernasconi et al., 2018) and acid fractionations associated with Brand parameters (Daëron et al. 2016). We use these originally published Δ_{47} CDES values in concert with the methodology described in Bernasconi et al. (2021) to project values into the I-CDES₉₀ reference frame.

We reprocessed data from Defliese and Tripati (2020) and Tang et al. (2014) that were measured at UCLA as part of this study in Easotope on I-CDES₉₀. Data from Anderson et al. (2021), Lucarelli et al. (2022), Jautzy et al. (2020), and Swart et al. (2021) were already in I-CDES₉₀ and were not reprocessed.

331 2.4 Ecological and Hydrographic data

Foraminiferal Δ_{47} values were compared to estimates of calcification temperature, that in turn are constrained by independent datasets (i.e., habitat, ocean atlas-derived calcification temperatures, oxygenisotope derived calcification temperatures). We also use multiple methods to probe secondary (non-thermal effects). We use Δ_{47} -temperature regression models to estimate vertical profiles of temperature and compare results to ocean atlas temperatures.

337 2.4.1 Habitat

338 Species were classified by depth habitats including mixed-layer, mixed-layer to thermocline, and 339 thermocline for planktic species, and epifaunal and infaunal for benthic species. Specific depth habitats and 340 ranges for different taxa were taken from Rippert et al. (2016) and Schiebel and Hemleben (2017) and 341 references therein. We note that calcification depths of foraminifera vary seasonally and can evolve with 342 ontogeny, as well as depend on factors influencing feeding patterns (Schiebel and Hemleben, 2017). Ocean 343 basins sampled in the complete dataset include the Arctic, Atlantic, Indian, and Pacific Oceans, as well as 344 the Mediterranean Sea and the Gulf of Mexico.

345 2.4.2 Calcification temperatures

346 Calcification temperatures for each sample were calculated for new and reprocessed samples, so 347 that we could ensure the same methods were consistently used (Supporting Information Table S3). Three 348 methods were utilized and a comparison of results is provided in the Supporting Information material 349 (Section S3). Briefly, method 1 uses the literature on calcification depths of foraminifera species in 350 geographical regions in conjunction with WOA temperature data (version 13V2; Levitus et al., 2010) to 351 estimate calcification temperatures. Method 2 consists of combining measured δ^{18} O of foraminiferal tests 352 with seawater δ^{18} O (at assumed for a database 353 (LeGrande and Schmidt, 2006), in combination with published taxon-specific δ^{18} O-temperature 354 relationships. For the latter, we referred to the most data-rich study on planktic taxon-specific δ^{18} O-355 temperature relationships from Malevich et al. (2019) that compiles data from >2600 core top samples from five taxa and reports Bayesian regression models that are taxon-specific as well as a pooled calibration. 356 357 They examined both annual and seasonal data, with similar errors for each. For our calcification temperature 358 estimates, in the case of taxa for which species-specific calibrations were provided, we used the seasonal 359 hierarchical model to estimate calcification temperatures and used the pooled calibration for all other 360 planktic species. To estimate calcification temperatures of benthic taxa, we implemented equation 9 of 361 Marchitto et al. (2014) for epifaunal species and mixed benthics, and the same equation modified by 0.47 362 ‰ for infaunal species as the authors suggest. Method 2, presented in the figures and the text, uses assumed 363 calcification depths and oxygen isotope measurements, to estimate calcification temperatures for planktic 364 foraminifera. Additionally, isotopic calcification temperatures were estimated using Method 2, but with equations from Kim and O'Neil (1997) and Shackleton (1974). Method 3 uses theoretical values of $\delta^{18}O_{calcite}$ 365 to develop vertical profiles and compares them to measured $\delta^{18}O_{\text{calcite}}$ values to determine the minimum and 366 367 maximum calcification depths.

368 2.4.3 Salinity

Estimated salinity is calculated using published decadal averaged seawater salinity from the WOA18 gridded dataset (Zweng et al., 2019). Salinity is averaged for each species at each site over their site-specific assumed calcification depths.

372 2.4.4 Bottom water saturation state

373 Bottom water saturation state was compared to benthic foraminiferal data to assess carbonate ion 374 effects on calcification, with the caveat that infaunal taxa may inhabit environments with values that differ 375 in composition from bottom waters. For these comparisons, calcite saturation state (Δ [CO₃²⁻]) is calculated as the difference between the in-situ carbonate-ion concentration ($[CO_3^{2-}]$) and the carbonate-ion 376 377 concentration at saturation ($[CO_3^{2-}]_{sat}$). Carbon system data was obtained from the nearest neighbor sites 378 from the World Ocean Circulation Experiment and the Global Ocean Data Analysis Project (Olsen et al., 379 carbonate-ion concentration calculated via Ocean Data Viewer 2019) and was 380 (http://woce.nodc.noaa.gov/wdiu) using best practices by Dickson et al. (2007).

381 2.5 Statistical Models

We used multiple statistical methods for this work. One method was used for developing regressions between Δ_{47} and temperature. Multiple approaches were applied for comparing subsets of data to assess if different models are needed (i.e., for foraminifera relative to synthetic calcite, or planktics relative to benthics). We used different approaches to assess if non-thermal effects are substantial (i.e., carbonate ion effects).

387 2.5.1 Regressions between Δ_{47} and calcification temperature

We fit regressions between Δ_{47} and calcification temperature utilizing Deming models. Among multiple analyzed frequentist regression models, Deming models are known to outperform other errors in variable models (i.e., York models), with higher levels of accuracy and precision for datasets of these sizes and similar error structures (Román-Palacios et al., 2021). Regression parameters from these models are generally congruent with Bayesian-derived estimates (Román-Palacios et al., 2021). We use Deming models to investigate potential non-thermal effects and differences between data subsets through confidence interval (CI) comparison.

395 2.5.2 Use of R-based model selection for investigation of non-thermal effects

396 We used a model selection approach to determine whether non-thermal factors substantially impact 397 isotopic signatures in foraminifera. Deming regressions are strictly univariate, so we substituted 398 multivariate ordinary least squares (OLS) linear models in R version 4.1.2 (R Core Team, 2022). The 399 potential suite of non-thermal variables consisted of lab/dataset, salinity, depth, region, photosymbiont 400 presence, habitat, genus, ocean, and bottom water saturation state. Because this constitutes a large number 401 of potential variables (set within a model as predictor variables) that could feasibly impact the final Δ_{47} 402 value, with an unknown potential for interactions between variables, we employed automated model 403 selection using the multithreaded implementation of the 'dredge' function in R package MuMIn version 404 1.47.1 (Bartoń, 2022). All numerical predictor variables were scaled and centered (i.e., z-scored) prior to 405 building models, as differences in the absolute magnitude of predictor variables can lead to erroneous 406 results in effect size calculations. Scaling the response variable was found to be unnecessary.

407 We report coefficients from linear models fit to z-scored data, which provides an estimate of the 408 per mil change in Δ_{47} for each 1 standard deviation (1 σ) increase in the predictor variable(s) of interest. 409 Uncertainty around coefficients is given in standard error in permil units. The effect size, partial eta squared 410 (n^2) , is used to evaluate how much variation in foraminiferal Δ_{47} is explained by each model component (fit 411 to z-scored data) and determine whether those effects are small (partial $\eta^2 = 0.01$), medium (partial $\eta^2 =$ 0.06), or large (partial $\eta^2 \ge 0.14$) when all other variables in the model are held constant. The calculation 412 for η^2 is mathematically similar to the calculation for R^2 and may therefore be interpreted in a similar 413 414 manner for individual model components. Note that we report effect sizes calculated using Type II (non-415 sequential) sums-of-squares as implemented in the car package for R (Fox and Weisberg, 2019). This 416 means that the impact of each predictor variable on foraminiferal Δ_{47} is evaluated individually in turn, while 417 all other predictor variables are held constant. Effect sizes may therefore add up to >1. As η^2 is calculated 418 using a one-sided hypothesis, we report one-sided 95% confidence intervals. Where final models include 419 only a single predictor variable (temperature), partial η^2 is equivalent to η^2 , and so η^2 is reported. Linear 420 model F-statistics are reported with the numerator and denominator degrees of freedom listed as subscripts 421 in that order.

422 A 'full' model was constructed for each subset of data (by lab/dataset, habitat, etc.) which contained 423 all relevant candidate predictor variables. The 'dredge' function then examined every possible combination 424 of predictor variables and determined the most likely reduced models. Candidate reduced models were 425 manually checked for goodness-of-fit and extraneous terms were dropped if warranted. To determine 426 whether dropping extraneous terms improved model fit, a hypothesis testing approach, followed by 427 examination of the Akaike Information Criterion (AIC), was used to select between the reduced model 428 identified by 'dredge' and a simplified model with terms dropped. If hypothesis testing and AIC indicated 429 that two reduced models were equally likely to fit the data, we defaulted to the simpler model (e.g., the 430 model with fewer terms or without interactions). Effect sizes and confidence intervals for final model 431 components were calculated using R package effectsize version 0.7.0.5 (Ben-Shachar et al., 2020), with 432 non-sequential (Type II) sums-of-squares calculated in R package car version 3.1-1 (Fox and Weisberg, 433 2019). Full details of this method are available in the Supporting Information (Section S5). We also use 434 OLS for directly investigating non-thermal dependences (i.e., regressions between Δ_{47} residuals and bottom 435 water Δ [CO₃²⁻]). Individual comparisons between data subsets are done using ANCOVA-type analyses 436 (performed in Graphpad Prism). Tests for multicollinearity confirm PCA results, with the additional ability 437 to include categorical predictors, and show no evidence of problematic degrees of correlation between 438 predictor variables including temperature, dataset, depth, carbonate saturation, salinity, and photosymbiont 439 presence/absence.

440 2.6 Temperature reconstruction

We reconstruct temperatures using core-top data for multiple species of foraminifera at several locations to evaluate the fidelity of using this approach to constrain hydrographic profiles in different oceanographic regimes. Temperatures were estimated using the full foraminiferal regression from the metaanalysis (labelled "All Foraminifera" in Table 3).

445 **3 Results and Discussion**

446 3.1 Sensitivity of regression to calcification temperatures estimated using different methods

447 Figure 2 shows Δ_{47} values for calibration samples compared to δ^{18} O-based calcification 448 temperatures estimated using the species-specific Bayesian model (method 2). Supporting Information 449 Table S3 contains the calcification temperature estimates from different methods, and water isotope values 450 and the supplement compaFres results from these different methods in more detail. Briefly, although all 451 three methods yield results that are broadly similar, there are some notable differences. Calcification 452 temperatures derived from method 1 (the WOA13 values) show more scatter than those estimated using 453 either methods 2 or 3 (oxygen isotope-based estimates) as is evidenced by a lower Pearson's correlation 454 coefficient (see Supporting Information S4). Differences between the Atlas (method 1) and isotopic 455 temperatures (method 2) reach up to 13.4 °C for the same sample, which is a Trilobatus sacculifer from the 456 Equatorial Pacific from Tripati et al. (2010). In contrast, calcification temperatures derived from method 2 457 and method 3 yield small differences for most samples (max = 5.1 °C). Method 2 allows for the use of the species-specific Bayesian model for carbonate δ^{18} O temperature estimates and decreases the temperature 458 459 estimate error for planktic samples. If we solely examine the new dataset generated at UCLA on 124 460 for a miniferal samples, we reach similar conclusions with the largest Pearson's correlation from Method 2 461 (r = 0.898) and relatively small differences between Method 2 and 3 (mean = 0.0077 °C, SD = 1.7 °C).

We conclude from our meta-analysis that the differences between methods are generally small, with the smallest offsets between Method 2 and Method 3 and the largest offsets between Method 1 and 464 Method 3. Results from Method 2 exhibited the strongest correlation between temperature and Δ_{47} , both in 465 the new dataset generated at UCLA and the overall meta-analysis. The method is similar to what was used 466 by Meinicke et al. (2020) and what is recommended by Daeron and Gray (2023). Given Method 2 has the 467 strongest correlation between temperature and Δ_{47} , both in the new dataset generated at UCLA and the 468 overall meta-analysis, it was the method utilized for calcification temperature estimation in this study.

469

470 Figure 2: Calibration data from this work, including new calibration data (blue) and a meta-analysis of all 471 data (grey) standardized to be on the same reference frame (I-CDES). (a) New calibration data for small 472 for a miniferal samples from the Eagle-Tripati Lab (blue; n = 124) compared to a meta-analysis of all data 473 (grey; n = 260), (b) New foraminiferal data compared to a meta-analysis of synthetic calcite from a similar 474 temperature range (0 to 50 °C; n = 118). Also shown are 95% confidence bands for the regressions. 475 Individual point errors and coefficient errors are reported as SE. The regression through all foraminiferal data from UCLA (n = 124) is: $\Delta_{47} = 0.0353 \pm 0.0015 \ 10^6/T^2 + 0.1973 \pm 0.0187$ while a regression for the 476 full dataset (n = 260) is: $\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1745 \pm 0.0154$. For synthetic calcite (n = 118), 477 the regression between temperature and Δ_{47} yields the equation with standard error: $\Delta_{47} = 0.0393 \pm 0.0014$ 478 479 $10^{6}/T^{2} + 0.1547 \pm 0.0165$. Regression parameters for all groups of data are in Table 3.



480

481 We also compare multiple models for estimating isotopic calcification temperatures utilizing 482 Method 2. Figures S1 and S2 compare calcification temperatures for planktics (from Malevich et al., 2019) 483 and benthics (derived from Marchitto et al., 2014) estimated using Method 2 to atlas temperatures (Method 484 1), and are modelled after Figure 7 from Daeron and Gray (2023). Figures S3-S6 are similar to these but 485 utilize Method 2 but show isotopic calcification temperatures derived using the equations of Kim and 486 O'Neil (1997) or Shackleton (1974). Figure S7 shows a histogram of the differences between isotopic 487 temperatures and atlas temperatures. In our analysis, there are fewer benthic foraminiferal samples with 488 large offsets between isotopic and atlas temperatures when the isotopic equation of Marchitto et al. (2014) 489 is used (Figure S2), in contrast to when equations from Kim and O'Neil (1997) or Shackleton (1974) are 490 utilized (Figures S4 and S6). We find the use of the temperature relationships from Malevich et al. (2019). 491 in contrast to Kim and O'Neil (1997) and Shackleton (1974), yields relatively few planktic samples from 492 cold waters (<10 °C) that have isotopic temperatures that are not within uncertainty of atlas temperatures 493 (Figure S1 compared to Figures S3 and S5). At warmer temperatures, there are more planktic foraminiferal 494 samples that are offset, irrespective of choice of δ^{18} O-calibration (Figure S1); such samples are described by Daeron and Gray (2023) as being isotopically "discordant" and in this analysis, are dominated by P. 495 496 obliquioculata and G. ruber (Figure S1). We speculate that the offset could reflect possible non-thermal 497 effects on planktic foraminiferal Δ_{47} . We note that the data for *P. obliquloculata* are for sites at a range of 498 water depths, and thus dissolution could be a contributing factor, but that is not the case for samples of G. 499 *ruber*, as these are from shallower sites.

500 501 $3.2 \Delta_{47}$ -Temperature calibration

502 Figure 2 and Table 3 show the relationship between temperature and Δ_{47} for the UCLA only full 503 for a miniferal dataset (n = 124), for the meta-analysis of the for a miniferal data, and the synthetic calcite 504 data. Figure 3 and Table 3 also report results for different ecological subgroups of data that are well-defined 505 (e.g., planktic, benthic, mixed-layer planktics, epifaunal vs infaunal benthics) (also reported in Supporting 506 Information Table S5 with other subgroups of data). Results are compared to a regression for synthetic 507 calcite grown at ≤ 50 °C (Tables 3, S3). Pearson's correlation results for all of the populations indicate a 508 strong correlation between temperature and Δ_{47} , consistent with multiple prior studies (e.g., Ghosh et al., 509 2006; Hill et al., 2014; Petersen et al., 2019; Schauble et al., 2006; Tripati et al., 2015). For example, the 510 meta-analyses of the foraminiferal data and synthetic calcite support a correlation that is significant at the 511 >99% level (r = 0.873, t = 28.7, p < 0.001, and r = 0.917, t = 24.4, p < 2.2e-16, respectively).

512

Datasets included	Ν	slope	s.e.	CI range	intercept	s.e.	CI range
UCLA Foraminifera	124	0.03527	0.00154	0.03222 to 0.03833	0.1973	0.019	0.1602 to 0.2344
All foraminifera	260	0.03739	0.00128	0.03487 to 0.03991	0.1745	0.015	0.1441 to 0.2049
All foraminifera, no mixed benthics	235	0.0383	0.00128	0.03579 to 0.04084	0.1635	0.015	0.13310 to 0.19390
Synthetic calcite <50 °C	118	0.03926	0.00143	0.03643 to 0.04208	0.1547	0.017	0.1219 to 0.1874
All benthics	67	0.028	0.00264	0.02267 to 0.03324	0.2948	0.034	0.2272 to 0.3625
All benthics, no mixed benthics	42	0.0342	0.00336	0.02740 to 0.04099	0.2149	0.043	0.12770 to 0.30210
All infaunal benthics	23	0.0365	0.00552	0.02505 to 0.04799	0.186	0.07	0.04126 to 0.3307
All epifaunal benthics	19	0.0317	0.00515	0.02081 to 0.04254	0.2474	0.067	0.1053 to 0.3895
All planktics	193	0.0423	0.00191	0.03854 to 0.04609	0.1166	0.023	0.07203 to 0.1612
All mixed-layer planktics	99	0.0423	0.00242	0.03746 to 0.04706	0.1194	0.028	0.063466 to 0.1754
All foraminifera and synthetic calcite	381	0.03894	0.00092	0.03713 to 0.04075	0.1566	0.011	0.1350 to 0.1782

513 **Table 3:** Regression parameters for foraminiferal calibration dataset and synthetic calcite dataset.

514

515 When comparing foraminiferal results to synthetic carbonates, we see no significant differences between

516 synthetic carbonates and foraminifera overall or between synthetic carbonates and different subgroups of

517 foraminifera excluding mixed benthics (Figures 2 and 3, Table 3). The regression through all

518 for a for a from UCLA (n = 124) is: $\Delta_{47} = 0.0353 \pm 0.0015 \ 10^6/T^2 + 0.1973 \pm 0.0187$ while a

519 regression for the full dataset (n = 260) is: $\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1745 \pm 0.0154$. For epifaunal

520 benthics (n = 19) the regression is $\Delta_{47} = 0.0317 \pm 0.0052 \ 10^6/T^2 + 0.2474 \pm 0.0674$. For infaunal benthics

521 (n = 23), the regression is $\Delta_{47} = 0.0365 \pm 0.0055 \ 10^6/T^2 + 0.186 \pm 0.0696$. For mixed layer planktics, the

522 regression is $\Delta_{47} = 0.0423 \pm 0.0024 \ 10^6/T^2 + 0.1194 \pm 0.0282$. For synthetic calcite (n = 118), the

regression between temperature and Δ_{47} yields the equation with standard error: $\Delta_{47} = 0.0393 \pm 0.0014$ $10^{6}/T^{2} + 0.1547 \pm 0.0165$.

525

Figure 3: Slope and intercept for Δ_{47} -temperature regressions for different data subsets from this study. Results show that regression parameters for datasets that are well constrained (i.e., taxonomically welldefined, with precise estimates of calcification temperature, and with a high n) and other datasets are indistinguishable from inorganic calcite (horizontal grey bar). These datasets include results for samples that were measured at UCLA, the full foraminiferal dataset, mixed-layer planktics, epifaunal benthics, and infaunal benthics, epifaunal and infaunal benthics, and all foraminiferal data excluding mixed benthics. 532 Panels A and B show slope and intercept for samples that are well constrained, respectively. Panels C and

533 D show slope and intercept for regressions through datasets that contain more poorly constrained samples.

534 Poorly constrained sample groups include groups with low n, groups that are taxonomically more variable, and groups that contain mixed benthic samples. 1 SE is shown in black, 95% CI in gray. Horizontal gray

- 535
- 536 bar is the SE band of the synthetic regression parameters.



537 538

539 3.3 Non-thermal effects in foraminifera

540 To constrain the potential scope of non-thermal effects in this proxy, we build on the body of 541 literature derived from in-depth studies of other foraminiferal proxies such as δ^{18} O and Mg/Ca (e.g., Anand 542 et al., 2003; Gray and Evans, 2019; Lea et al., 1999; Russell et al., 2004; Stirpe et al., 2021). One group of 543 effects are biological in origin. In some cases, taxon-specific calibrations have been argued to be a better 544 fit (e.g., Bemis et al., 1998; Malevich et al., 2019; Skinner and Elderfield, 2005), while others suggested 545 the use of pooled calibrations (e.g., Anand et al., 2003; Malevich et al., 2019). Recent work surveying 546 cryptic species of foraminifera has shown there are genotype-specific biological controls on shell 547 geochemistry, leading to suggestions that regional calibrations of proxies may also potentially be useful 548 (Darling et al., 2017; Sadekov et al., 2016). Additionally, basinal differences have been identified for Mg/Ca 549 (i.e., a local intercept for Mg/Ca-temperature calibration equations are adjusted to yield modern 550 temperatures from core-top values; Skinner and Elderfield, 2005).

551 Seawater chemistry has been identified in some studies as being important for proxy systematics. 552 In particular, carbonate chemistry and salinity have been described in the literature as affecting 553 for a miniferal δ^{18} O and/or Mg/Ca. Carbonate ion effects on oxygen isotopes (Spero et al., 1997) and Mg/Ca 554 (Lea et al., 1999; Russell et al., 2004) have been noted from culturing of planktic foraminifera. Salinity 555 effects have also been reported on Mg/Ca (Ferguson et al., 2008; Gray and Evans, 2019; Hoogakker et al., 556 2009; Lea et al., 1999). Bottom water carbonate saturation is thought to impact Mg/Ca ratios in benthic 557 foraminifera (Dawber and Tripati, 2012; Elderfield et al., 2006; Yu and Elderfield, 2008), and potentially 558 oxygen isotopes (Bemis et al., 1998).

559 These factors have mostly not yet been examined for clumped isotopes in foraminifera. Studies of 560 Δ_{47} have concluded that taxonomic effects have not been detectable (Tripati et al., 2010; Peral et al., 2018; 561 Meinecke et al., 2020, Daëron and Gray, 2023). Similarly, carbonate ion effects have been explored in 562 benthic foraminifera in two studies using relatively small sample sizes; they were not detectable (Tripati et 563 al., 2010; Piasecki et al., 2019). Another study probed for and found no evidence for salinity effects in 564 foraminifera (Peral et al., 2018). The potential scope of carbonate ion and salinity effects in foraminifera 565 has been explored with theory and the mechanisms associated with clumped isotope effects linked to DIC 566 speciation shown to, in principle, be limited across most oceanic conditions (Hill et al., 2014; Tripati et al., 567 2015). Thus, we hypothesize that non-thermal effects are likely to be weak.

568 3.3.1 Methodological

569 Our meta-analysis shows that despite methodological differences, for foraminiferal data, there is 570 strong agreement between most labs when data are placed on the I-CDES reference frame. Clumped isotope 571 measurements involve examination of low abundance isotopic species, and thus are counting statistics 572 limited, and necessitate accurate and precise measurements. Thus, part of our work for this study has 573 involved recalculating available data determined using different methods on the same I-CDES reference 574 frame using published standard values.

575 Between studies, the instrumentation used varies - including carbonate digestion systems and 576 associated reaction temperatures, mass spectrometers (with distinct signal to noise, and mass spectrometric 577 corrections), and frequency of standards run and standardization approaches. Different labs have variable 578 measurement procedures, with a range of integration times (less than 400 seconds - to 1200 seconds or 579 more), sample sizes (sub-mg or larger for a single analysis), and replication (3 or less to >8). Standard 580 corrections differ, with some labs averaging standard values over multiple days or weeks for corrections, 581 while other labs characterize short-term drift. Some datasets necessitated the use of transfer functions to 582 convert from CDES to I-CDES reference frames (Tripati et al., 2010; Breitenbach et al., 2018; Peral et al., 583 2018), which uses average standard δ^{47} values. However, this should contribute < 3 ppm for δ^{47} values 584 within \pm 18 ‰ of ETH-1 (Bernasconi et al., 2021).

These foraminiferal calibration data come from six clumped isotope laboratories (Table 1) that utilize different instruments and analytical methods. There is particularly strong agreement between data from Tripati et al. (2010), Breitenbach et al. (2018), Meinicke et al. (2020), and this study (Eagle-Tripati lab), that overlap with each other including at the extremes of the highest and lowest temperatures. All of these datasets show more variability at warmer temperatures, which could reflect the larger sample density relative to cold regions.

591 Each dataset included in this study has regression slopes and intercepts that fall within the 95% 592 confidence intervals of each other (Supporting Information Table S6). Within the foraminiferal data 593 synthesis for 260 samples, dataset-related (i.e., methodological artifacts) account for up to 11% of the 594 variance from our non-thermal R-model based effect size testing. This largely comes from two datasets that 595 are amongst the three smallest in terms of sample numbers (Peral et al., 2018 - 27 samples; Piasecki et al., 596 2019 - 13 samples). The relatively small sample sizes could potentially explain these offsets, although we 597 cannot preclude that differences in instrumentation and/or standardization during the specific intervals that 598 samples were run may also be a contributing factor. At low temperatures, both the Peral et al. (2018) and reprocessed Peral et al. (2022) results are offset from the other studies, with a slightly steeper slope, while

600 the Piasecki et al. (2019) benthic foraminiferal values are positively offset from other studies, especially at 601 bigher temperatures, and exhibit a shellower clone (Supporting Information Table S6)

601 higher temperatures, and exhibit a shallower slope (Supporting Information Table S6).

- 602 3.3.2 Biological
- 603 3.3.2.1 Ecology

604 Here, if we subdivide benthic foraminiferal data by habitat (Figures 3 and 4), we see that the slopes 605 and intercepts of both the infaunal and epifaunal species individually fall within the 95% confidence 606 intervals of planktic regression slope and intercept. While the mixed benthic species regression does not 607 overlap with the planktic regression 95% CI, this is likely a result of low data density at warmer temperature, 608 an artifact of how measurements were combined between species (e.g., Piasecki et al., 2019), potential 609 mixing effects, or a combination of these factors. It is also possible that some of the data are offset due to 610 other potential non-thermal effects that the benthic population is sampling (e.g., a carbonate ion saturation 611 effect, discussed further below).

612 3.3.2.2 Taxon-specificity

Temperature proxies including for a forminiferal δ^{18} O and Mg/Ca show that interspecific offsets exist 613 for multiple taxa (e.g., Bemis et al., 1998; Jentzen et al., 2018; Marchitto et al., 2014; Regenberg et al., 614 615 2009). Previous studies looking for such effects in Δ_{47} have suggested that there are little to no vital effects 616 on this thermometer (Tripati et al., 2010; Grauel et al., 2013; Peral et al., 2018; Meinicke et al., 2020, 617 Daëron and Gray, 2023). However, this previous work on Δ_{47} has investigated relatively small numbers of 618 species, with limited numbers of localities sampled for each species (typically n < 7). Here, with our new 619 data and the meta-analysis, we assess taxon-specific effects using data for 29 species, with n>20 for five 620 species.

621 Our analysis indicates that taxon-specific offsets are not resolvable when subgroups of data are 622 examined (Figure 5). If grouped by genus, mean residuals (Δ_{47} observed minus Δ_{47} predicted from the 623 calibration for all foraminifera) for 14 taxa are smaller than +/-0.005 ‰, and for all but one taxa are less 624 than +/-0.01 \%. At the genus level, residuals are indistinguishable from a normal distribution at the 95% 625 confidence level for 10 out of 12 genera, the exceptions being *Trilobatus* and *Uvigerina* where there is a 626 skewness of -0.10 and -2.03 respectively. The skew in the data for *Uvigerina* (n = 9) is driven by a single 627 point that lies far outside the rest of the samples. Trilobatus comprises two species and when these are 628 looked at individually, each species is not distinguishable from a normal distribution at the 95% confidence 629 level (*T. sacculifer:* W = 0.94203831, p = 0.09396572; *T. trilobus:* W = 0.97780615, p = 0.88903859), but 630 when combined reflect a larger trend in the overall dataset of a minor skew to the left (Figure S9).

631 If considered at the species-level, mean species residuals for 17 taxa are smaller than ± -0.005 %, 632 and for all but 5 taxa are less than 0.01 ‰. Residuals are indistinguishable from a normal distribution other 633 than N. pachyderma (Shapiro-Wilk test, p > 0.05). For the five species represented at a large number of 634 localities (P. obliquiloculata, T. sacculifer, G. tumida, G. bulloides, G. ruber), mean residuals are less than 635 0.005 % (Figure 5, S9). All but one species exhibit residuals within ± -0.02 %, which is the long-term 636 standard deviation for standards reported for different instruments and by multiple labs (e.g., Meckler et 637 al., 2014; Bernasconi et al., 2021; Upadhyay et al., 2021). These residuals are normally distributed for all 638 taxa that have a large number of samples.

639

Figure 4: Regressions through subsets of core-top foraminiferal data and foraminiferal compilation. All subsets are plotted on top of the foraminiferal compilation (grey) with subsets highlighted as follows: a)
Planktic foraminifera b) Benthic foraminifera c) Mixed-layer planktics d) Mixed benthics e) Infaunal
herthics for Eniformal herthics 05% coefficience interval is also shown

643 benthics f) Epifaunal benthics. 95% confidence interval is also shown.

644

We note that the taxa with larger residuals and non-normal distributions could bias the regression, and thus, we assessed impacts of individual species on the slope and intercept of the overall calibration line by removing species individually from the calculation and reporting the resulting values, following the approach of Meinecke et al. (2020). When using this approach, both slopes and intercepts still remain well within error of the primary regression, suggesting that individual taxa are not unduly influencing the regression (Supporting Information Table S6).

We investigated other potential biological effects on Δ_{47} associated with mineralogy and photosymbionts. *H. elegans* is an aragonitic species of foraminifera and was measured in multiple studies. However, we do not observe any systematic offsets of data from this species (Figure 5, Supporting Information Table S6). Photosymbiosis in foraminifera affects both calcification rates and microenvironments (de Nooijer et al., 2014) and has the potential to cause disequilibrium isotopic values

- 656 (e.g., Spero et al., 1997). We do not observe systematic offsets when comparing non-photosymbiont bearing 657 species and photosymbiont bearing species at the 95% confidence interval (Figure S10).
- **Figure 5:** Δ_{47} residuals for different taxa calculated as $\Delta_{47\text{observed}}$ minus $\Delta_{47\text{calculated}}$ using the foraminiferal
- calibration equation from our meta-analysis. a) Species-specific residuals. b) Grouped genus residuals.
- 660

661 3.3.2.3 Oceanic region

662 Given basin-specific Mg/Ca calibrations (e.g., Skinner et al., 2007) and region-specific genotypes 663 (Sadekov et al., 2016; Darling et al., 2017), we explored whether geography impacted calibration in the 664 meta-analysis. For all regions with large numbers of samples (Atlantic Ocean, Pacific Ocean, Indian Ocean, 665 Gulf of Mexico), 95% confidence intervals overlapped with the broader foraminiferal calibration (Figure 666 S11). The Arctic Ocean contains only a small number of samples and a limited temperature range so the 667 uncertainties in a basinal calibration are large, and additional calibration material from this region (and 668 other high latitude areas) are critical.

669 3.3.3 Seawater Chemistry

670 Given evidence for seawater chemistry impacts on Mg/Ca, we explore if there are discernable 671 effects of seawater chemistry on Δ_{47} in foraminifera. We first used PCA to assess how much variance in 672 hydrographic parameters as potential predictors can be attributed to different factors, and whether any 673 potential model components are strongly correlated with one another. We find no evidence of such 674 correlations between hydrographic parameters, when investigating temperature, salinity, carbonate 675 saturation, and depth (Figure S12). Below, we describe the relationship between Δ_{47} residuals and salinity. 676 We also examine if in benthic foraminifera, Δ_{47} is correlated with saturation state.

677 3.3.3.1 Salinity

678 Salinity variation has been shown to affect Mg/Ca sensitivity by around 4.5 % per salinity unit 679 depending on the species (e.g., Ferguson et al., 2008; Hönisch et al., 2013). A small salinity effect on Δ_{47} 680 associated with the salinity-dependence of dissolved inorganic carbon (DIC) speciation has been predicted 681 from theory, but was not thought to be impactful over the salinity range of the oceans (Tripati et al., 2015). 682 Consistent with this prediction, previous studies on clumped isotopes in foraminifera that had small sample 683 sizes suggest no detectable correlation with salinity (Grauel et al., 2013; Peral et al., 2018).

684 We examined the larger dataset to see if a salinity effect was detectable. Foraminifera from this 685 study cover a salinity range of 33.0 to 38.8 PSU and as a whole similarly show no clear correlation of 686 salinity with Δ_{47} residuals (Pearson's correlation *p*-value = 0.77, Figure S13). These results support 687 theoretical predictions (Tripati et al., 2015). However, Pearson's correlation between the two variables has 688 a *p*-value of 0.01 for the benthic data. This correlation does not appear in the planktic data. Analysis of the 689 correlation between temperature and salinity ranges in the benthic data suggest that the correlation observed 690 may be due to offsets in temperature trends observed in section 3.2.2 rather than a true salinity effect (Figure 691 S13).

692 3.3.3.2 Carbonate saturation effects on benthics

693 Seawater carbonate chemistry has been shown to impact benthic foraminiferal Mg/Ca (Elderfield 694 et al., 2006). A number of approaches have been used to detect such an effect, including direct comparisons of benthic Mg/Ca to seawater $\Delta[CO_3^{2-}]$, a comparison of temperature-corrected Mg/Ca to $\Delta[CO_3^{2-}]$, as well 695 696 as the use of a multiple linear regression to examine the sensitivity to temperature and carbonate ion 697 (Elderfield et al., 2006). Based on culturing studies, a carbonate ion effect in δ^{18} O has been observed in 698 planktic foraminiferal calcite and coral aragonite (Spero et al., 1997). In benthic foraminifera, 699 disequilibrium in δ^{18} O can systematically vary between taxa, linked to carbonate ion (Bemis et al., 1998; 700 Ishimura et al., 2012; Rathman and Kuhnert, 2008). For δ^{18} O, this disequilibrium is hypothesized to be 701 related to changes in the isotope composition of the dissolved inorganic carbon pool as a function of pH 702 (Zeebe, 1999).

703 When we compare benthic Δ_{47} residuals ($\Delta_{47\text{observed}}$ minus $\Delta_{47\text{calculated}}$ using the foraminiferal 704 temperature calibration equation from our meta-analysis) to Δ [CO₃²⁻] (Figure 6), we find a significant 705 positive correlation, consistent with a weak sensitivity of benthic Δ_{47} to changing carbonate ion (Figure 6B; 706 slope = 0.00019, $R^2 = 0.315$, p < 0.01). We note that the correlation largely reflects a significant relationship 707 in the epifaunal benthic foraminiferal data significant at the >95% confidence level (Figure 6C; slope = 708 0.00018, $R^2 = 0.315$, p = 0.043, n = 19), and in the mixed benthics data (Figure 6E; slope = 0.00022, $R^2 =$ 709 0.482. p <0.01, n = 24). The data showing this trend were generated as part of this study and by Piasecki et 710 al. (2019) (Figure 6A). No significant correlation is found at the >95% confidence level in the infaunal 711 benthic data (Figure 6D).

If we directly compare benthic Δ_{47} to Δ [CO₃²⁻], a significant but weak (p=0.01, R²=0.09) is observed in the meta-analysis (Figure S14). This pattern is similar to what was reported for benthic Mg/Ca (Elderfield et al., 2006). However, different sub-groups of the benthic foraminiferal data (e.g., epifaunal benthics, infaunal benthics, etc.) do not exhibit significant linear correlations, but this may not be surprising if an effect is present but relatively weak.

We developed benthic multivariable regressions that account for a dependence of Δ_{47} on both temperature and $\Delta[CO_3^2]$, in the form of Δ_{47} (‰) = $\beta 0 + \beta 1 \times 10^6$ /Temperature (K)² + $\beta 2 \times \Delta[CO_3^{2-}]$ (Table 4 and Figure 7). These analyses also support sensitivities of Δ_{47} to $\Delta[CO_3^{2-}]$ that are generally similar in magnitude to those derived using other methods (e.g., regressing Δ_{47} residuals against $\Delta[CO_3^{2-}]$), with values of 0.0017 for all benthic foraminifera, 0.0015 for all benthic foraminifera (epifaunal and infaunal benthics), 0.0019 for epifaunal benthic foraminifera, and 0.0013 for infaunal benthic foraminifera (Figure 7, Table 4).

c) epifaunal benthics, d) infaunal benthics, e) mixed benthics, f) all benthics (infaunal + epifaunal).

727

728 3.3.4 R-based model selection

729 3.3.4.1 Evaluation of non-thermal effects in UCLA dataset

730 The benthic foraminiferal dataset from the Eagle-Tripati lab at UCLA (n = 42) is best explained by 731 a model with temperature as the sole predictor ($F_{1,40} = 64.6$, SE = 0.01, Adjusted R² = 0.608, p < 0.00001; 732 Supporting Information Table S7). If planktic foraminiferal Δ_{47} from the UCLA dataset is considered, the 733 signal is best explained by an additive model containing temperature, bottom water carbonate saturation (i.e., a dissolution effect), and photosymbiont presence/absence ($F_{3,78} = 50.9$, SE = 0.01, Adjusted $R^2 =$ 734 735 0.649, p <0.00001; Supporting Information Table S7). Temperature explains the vast majority of the variation in Δ_{47} (Partial $\eta^2 = 0.57$, lower 95% CI 0.46; Supporting Information Table S7), followed by 736 bottom water carbonate saturation and photosymbiont presence/absence, which both have a "medium" 737 738 effect, and return the same effect size and lower 95% CI (Partial $\eta^2 = 0.06$, lower 95% CI 0.0). Temperature 739 has a large positive effect of 0.029 \pm 0.003 ‰ on Δ_{47} for each 1 σ increase. Bottom water carbonate 740 saturation in the linear model exhibits a slight positive effect of 0.006 ± 0.003 ‰ on Δ_{47} per 1 σ increase in

- carbonate ion saturation. Photosymbiont presence produces a slight positive offset in Δ_{47} relative to taxa
- 742 who lack photosymbionts of 0.009 ± 0.004 ‰.

743 744 Figure 7: Predicted Δ_{47} compared to measured Δ_{47} for benthics. Predicted Δ_{47} is from a multivariate regression of the form Δ_{47} (‰) = $\beta 0 + \beta 1 \times 10^6/T^2 + \beta 2 \times \Delta[CO_3^{2-}]$ (regression parameters in Table 4). a) 745 746 all benthics, b) all benthics (infaunal + epifaunal), c) infaunal benthics, and d) epifaunal benthics. There are 747 at least two different ways, in theory, that Δ_{47} could be affected by bottom water carbonate saturation. First, a small effect has been predicted from theory and observed in experiments for Δ_{47} , with a difference in the 748 749 equilibrium isotopic composition of DIC species at pH between 7 and 11 of ~ 0.03 to 0.05 ‰, that was 750 predicted to give rise to effects of less than 0.02 % for the Cenozoic, and lower Δ_{47} at elevated pH (Hill et 751 al., 2014; Tripati et al., 2015). A second effect, with increasing Δ_{47} associated with increased pH, could 752 arise from a possible CO₂ hydrolysis effect that increases with precipitation rate (Guo, 2020; Lucarelli et 753 al., 2022; Tripati et al., 2015). Experimental studies have observed this effect only at low temperature and 754 elevated pH (pH > 9.5) with decreased effect size in the presence of carbonic anhydrase recently shown to 755 be active in some benthic foraminiferal species (de Goeyse et al., 2021; Lucarelli et al., 2022).

3.3.4.2 Statistical model investigation of non-thermal effects in meta-analysis of benthic and planktic
 foraminifera

758 Benthic foraminiferal Δ_{47} in the meta-analysis (n = 67) is best explained by the additive effects of 759 temperature, bottom water carbonate saturation, and salinity ($F_{3,63} = 54.4$, SE = 0.01, Adjusted R² = 0.708, 760 p <0.00001). Temperature again explains the vast majority of the variation in Δ_{47} (Partial $\eta^2 = 0.60$, lower 761 95% CI 0.47; Supporting Information Table S7), followed by carbonate saturation (Partial $\eta^2 = 0.25$, lower 762 95% CI 0.11), and salinity (Partial $\eta^2 = 0.08$, lower 95% CI 0.01). Temperature has a positive effect on Δ_{47} 763 of 0.022 ± 0.002 % for each 1 σ increase; carbonate saturation has a slight positive effect of 0.009 ± 0.002 764 ‰; and salinity has a slight negative effect of 0.005 ± 0.002 ‰. We found no evidence of a significant 765 effect of benthic foraminiferal habitat (epifaunal versus infaunal) on the final Δ_{47} signal (Estimate = -0.0001, Std err = 0.004, t = -0.033, p = 0.97). 766

For the meta-analysis, planktic foraminiferal Δ_{47} (n = 193) is explained by multiple variables. Model selection identified three-way interactions between temperature, dataset, and bottom water carbonate saturation; and temperature, bottom water carbonate saturation (i.e., dissolution), and photosymbiont presence/absence (F_{23,169} = 31.6, SE = 0.01, Adjusted R² = 0.786, p <0.00001). Temperature explains the vast majority of the variation in Δ_{47} (Partial η^2 = 0.62, lower 95% CI 0.55; Supporting Information Table S6), followed by the interaction between dataset and bottom water carbonate saturation (Partial η^2 = 0.21, 173 lower 95% CI 0.11); dataset alone and the three-way interaction between temperature, dataset, and 174 carbonate saturation (Partial $\eta^2 = 0.15$, lower 95% CI 0.07); (Both: Partial $\eta^2 = 0.14$, lower 95% CI 0.06);

carbonate saturation (Partial $\eta^2 = 0.10$, lower 95% CI 0.07), (Both, Partial $\eta^2 = 0.14$, lower 95% CI 0.07), carbonate saturation alone (Partial $\eta^2 = 0.10$, lower 95% CI 0.04) and the two-way interaction between

temperature and carbonate saturation (Partial $\eta^2 = 0.06$, lower 95% CI 0.01), with all other combinations

having an effect size of 0.04 or less.

Table 4: Multivariable regressions between Δ_{47} (‰), temperature T (K) and Δ [CO₃²⁻] (µmol/kg).

$\Delta_{47} (\%) = \beta 0 + \beta 1 \times 10^{6} / T^{2} + \beta 2 \times \Delta [CO_{3}^{2-}]$									
All benthics									
Parameter estimates	Variable	Estimate	Standard error	P value	R ²	n			
				< 0.000					
β0	Intercept	0.1968	0.0415	1					
				< 0.000	0.6926	67			
β1	$10^{6}/T^{2}$	0.03504	0.003138	1	0.0720	07			
		0.000167							
β2	$\Delta[CO_3^{2-}]$	3	0.00004572	0.001					
	All benthics	s (epifaunal a	nd infaunal benth	ics)					
Parameter estimates	Variable	Estimate	Standard error	P value	R ²	n			
β0	Intercept	0.1518	0.04932	0.0038					
				< 0.000					
β1	$10^{6}/T^{2}$	0.03865	0.00376	1	0.7427	42			
•		0.000149							
β2	$\Delta[CO_3^{2-}]$	5	0.00006014	0.0173					
	· · · · · · · · · · · · · · · · · · ·	All infaunal	benthics						
Parameter estimates	Variable	Estimate	Standard error	P value	R ²	n			
ß0	Intercept	0.1441	0.06748	0.0453					
F*				< 0.000					
β1	Temperature	0.03936	0.005239	1	0.743	23			
		0.000127							
β2	$\Delta[CO_3^{2-}]$	1	0.00007825	0.1199					
•		All epifauna	l benthics						
Parameter estimates	Variable	Estimate	Standard error	P value	R ²	n			
β0	Intercept	0.152	0.09113	0.1147					
				< 0.000	0.0000	10			
β1	Temperature	0.03851	0.006841	1	0.6838	19			
β2	Δ [CO ₃ ²⁻]	0.000188	0.0001059	0.0949					

780

781 3.3.4.3 Statistical model investigation of non-thermal effects in full foraminiferal dataset

782 R-based model selection indicates that temperature effects vastly outweigh non-thermal signals in 783 the full foraminiferal dataset (n = 260) (Supporting Information Table S6). Non-thermal effects can account for up to 13% of the variance in the dataset if all other variables are held constant. For a miniferal Δ_{47} is best 784 785 explained by temperature, plus the effects of three two-way interactions (foraminiferal type and bottom 786 water carbonate saturation; dataset and depth; dataset and photosymbiont presence/absence; $F_{20,239} = 60.8$, SE = 0.01, Adjusted R^2 = 0.822, p < 0.00001; Supporting Information Table S6). Temperature has a positive 787 788 effect on Δ_{47} of 0.028 ± 0.001 ‰ for each 1 σ increase. The interaction between dataset and depth accounts for 13% of the variance (Partial $\eta^2 = 0.13$, lower 95% CI 0.06, followed by dataset alone, which accounts 789 790 for 10% of the variance in Δ_{47} (Partial $\eta^2 = 0.10$, lower 95% CI 0.03). All other model terms have a partial 791 η^2 of 0.05 or less.

792 3.4 Vertical profiles of temperature reconstructed from core-top planktics

793 We assess the utility of using foraminiferal Δ_{47} -derived temperature estimates in reconstructing 794 oceanic water column thermal properties in different oceanographic regions. For planktics, Δ_{47} -temperature

- estimates are calculated using the full foraminiferal regression from the meta-analysis. We compared WOA temperature profiles with Δ_{47} -reconstructed temperatures from multiple species we analyzed from the Arctic Ocean, North Atlantic, Indian Ocean, and Western Equatorial Pacific (Figure 8; Figures S15-S17). Reconstructed temperatures are plotted at assumed calcification depths for each taxa at each site (Figure 8, Supporting Information Table S2). A cross comparison of Δ_{47} -temperature and expected temperature based on calcification depth shows good agreement (slope of 0.99 ± 0.03 , $R^2 = 0.75$ and p < 0.001).
- 801

Figure 8: Reconstructed Δ_{47} -temperatures a) Comparison of Δ_{47} -temperature with δ^{18} O-based calcification

temperatures (method 2) for all core-top samples. Linear regression has a slope of 0.99 ± 0.03 , $R^2 = 0.75$

- and p < 0.001 (n = 260). Reconstructed temperature profiles based on clumped isotopes temperature for b)
- 805 the Indian Ocean, c) Western Pacific, d) North Atlantic and e) Arctic Ocean.

806 Most Δ_{47} -temperatures plot within error of measured water temperature profiles. With the exception 807 of the Arctic data, there is little indication of seasonality being resolvable using this method because 808 typically, the uncertainty in Δ_{47} -temperatures is greater than the seasonal extremes of temperatures at the 809 sites chosen. Interestingly, in the case of near freezing temperatures, our results show no systematic offset 810 from the measured site temperatures which suggests that there is no observed isotopic offset introduced by

- 811 calcification at these cold temperatures in foraminifera suggested in other studies (Tripati et al., 2010),
- although it should be noted that samples in this region are limited only to the species *N. pachyderma*.
- 813 3.5 Recalculating Cenozoic bottom water temperatures using new calibrations
- 814

5.5 Recalculating Cenozoic bottom water temperatures using new canorations

815 Daeron and Gray (2023) and Rohling et al. (2024) discuss some of the differences between published reconstructions of bottom water temperatures from benthic for a miniferal Δ_{47} and δ^{18} O, including 816 817 discrepancies between proxy predictions for the Late Paleocene, Eocene, Eocene, Oligocene transition, and 818 Pleistocene, highlighting the importance of utilizing new calibration data and constraints from carbonate 819 chemistry in such calculations. Thus, here we apply the new calibrations to benthic foraminiferal data and 820 evaluate Δ_{47} -temperatures and seawater water δ^{18} O. We recalculate Cenozoic temperatures with low-821 resolution benthic Δ_{47} -data from Meckler et al. (2022) for the North Atlantic from IODP Expedition 342 822 (Sites 1406, 1407, 1409, and 1410) on the CENOGRID timescale (Westerhold et al., 2020). We recalculate 823 Eocene-Oligocene temperatures using benthic Δ_{47} -data from Taylor et al. (2023) for ODP Site 1218 and 824 IODP Sites U1333 and U1334 from the Pacific and the ages reported therein.

825 Below, we discuss results from the different calibration approaches for the Pleistocene (Section 826 3.5.1), Eocene-Oligocene transition (Section 3.5.2), Early Eocene Climatic Optimum (Section 3.5.3), and 827 Paleocene (Section 3.5.4). We also calculate the potential scope of impacts on benthic reconstructions for 828 the PETM (Section 3.5.5). In each section, we compare the originally published benthic Δ_{47} -reconstructions 829 that utilize the planktic calibration of Meinicke et al. (2020) to values calculated using (1) the new Δ_{47} -830 temperature calibrations and (2) the new benthic multivariate Δ_{47} calibration that also factor changes in 831 carbonate chemistry into estimates of temperature and seawater δ^{18} O (Supporting Information Table S8-832 10). For the latter, to constrain changes in carbonate chemistry, one can use site- or region-specific estimates 833 of bottom water carbonate saturation based on Li/Ca or B/Ca data from benthic foraminifera (Lear and 834 Rosenthal, 2006, Yu et al., 2007), which is the approach we use in Section 3.5.2 (the Eocene-Oligocene 835 transition) because Li/Ca data were available. Another approach that may be less accurate is to utilize a 836 combination of constraints on two different carbonate system parameters using proxy and/or carbon cycle 837 model calculations (e.g., Meckler et al., 2022; Roberts and Tripati, 2009; Tyrell and Zeebe, 2004). This 838 first-order calculation of the impact of $[CO_3^{2-}]$ on Δ_{47} -temperatures is what we utilize for the other sections. We calculated $[CO_3^2]_{in situ}$ using CO₂sys excel (v. 2.3. Pierrot et al. 2016) and published constraints on pCO₂ 839 840 and Total alkalinity (Supporting Information Table S8). pCO₂ was interpolated from a 50 points average 841 record from Hönisch et al. (2023) based phytoplankton and boron isotopes. Total alkalinity was interpolated 842 using the published record from Tyrell and Zeebe (2004). The precipitation constant for calcite (Ksp, Mucci 843 et al. 1983) was calculated using a constant salinity of 35 and the published temperature from clumped 844 isotopes. $[CO_3^{2^-}]_{sat}$ was then determined for $\Omega=1$ and $[Ca^{2^+}]$ interpolated from Horita et al. (2002). 845

846 3.5.1 Cooler Pleistocene temperatures calculated using new benthic Δ_{47} -temperature calibrations

847 Pleistocene temperatures estimated by Meckler et al. (2022) range from -0.3 °C to 3.4 °C, and 848 average 1.6 °C (Figure 9, Supporting Information Table S9). All of the new Δ_{47} -temperature calibrations 849 yield cooler estimates of Pleistocene temperatures (Figure 9, Supporting Information Table S9). For 850 example, the "all foraminifera-MB" regression gives similar values to the temperature calibration from 851 And erson et al. (2021) for the Pleistocene, with Δ_{47} -temperatures ranging from -1.8 to 1.6 °C for the "all 852 for aminifera-MB" and -1.5 to 1.9 °C using the Anderson et al. (2021) regression. Temperatures calculated 853 using the "all benthics" calibration range from -5.1 to 0.5 °C and the "infaunal" regression from -2.3 to 1.3 854 °C. The "all benthics (infaunal + epifaunal" and "epifaunal" regressions yield similar results to each other 855 for the Pleistocene, with temperatures ranging from -3.1 to 0.6 °C for "all benthic – MB" and from -3.8 to 856 0.3 °C for "epifaunal". The full foraminifera calibration supports temperatures in between -2.2 to 1.3 °C.

857 Estimates of $\delta^{18}O_{sw}$ from these calibrations are lower than those estimated from Meckler et al. 858 (2022), with values of 0.44 ‰ for the "all foraminifera - MB" calibration, 0.71 ‰ for the "All benthic 859 (infaunal + epifaunal)" regression, 0.85 ‰ for the "epifaunal" regression, and 0.54 ‰ for the "infaunal" 860 regression (Figure 9, Supporting Information Table S9). For the "compiled benthic", "All benthic (infaunal

- 861 + epifaunal)", and "epifaunal" regressions, the average $\delta^{18}O_{sw}$ is negative (Figure 9, Supporting Information
- Table S9). Note that negative $\delta^{18}O_{sw}$ values for the Pleistocene are not likely given that average global icevolumes are expected to be between interglacial and glacial extremes, which should lead to positive $\delta^{18}O_{sw}$
- values (Cramer et al., 2011).
- 865 3.5.1.1 Pleistocene temperatures from the new benthic multivariate Δ_4 -calibration
- Figure 11 shows Pleistocene values of temperature that are calculated using the "All benthic (infaunal + epifaunal)" multivariate calibration. Pleistocene Δ_{47} -temperatures for the Atlantic range from -
- $0.7 \circ C$ to $3.0 \circ C$, in better agreement with the temperature estimates from Meckler et al. (2022) (Figure 11,
- Supporting Information Table S8). Estimates of $\delta^{18}O_{sw}$ range from -0.09 to 1.14 ‰ and average 0.55 ‰.
- 870 The positive $\delta^{18}O_{sw}$ predicted by the multivariate regression is in better agreement with estimates of ice
- 871 volumes from the Pleistocene (Cramer et al., 2011).
- 872
- 873 3.5.2 Cooler Eocene-Oligocene bottom water temperatures from the new benthic Δ_{47} -temperature calibrations
- 875 We recalculate tropical Pacific Ocean bottom water temperatures using benthic Δ_{47} data from 876 Taylor et al. (2023) (that used the planktic calibration of Meinicke et al., 2020) for ODP Site 1218 and 877 IODP Sites U1333 and U1334. These sites were located at a paleodepth of 3500 to 4000 m. This transition 878 is associated with the permanent establishment of large ice sheets on Antarctica and a ~1 km deepening of 879 carbonate compensation depth in the Pacific Ocean (Coxall et al., 2005). It is thought that the deepening of 880 the carbonate compensation depth was global (Tripati et al., 2005) and associated with a rise in $[CO_3^{2-}]$ 881 (Pusz et al., 2011). Taylor et al. (2023) report late Eocene Δ_{47} -temperatures of 11.0 °C and Earliest 882 Oligocene temperatures of 9.7 °C, with a change across the Early Oligocene Oxygen Isotope Step (EOIS, 883 absolute temperature of 6.5 °C) of -4.7 ± 0.9 °C.
- We apply the new foraminiferal Δ_{47} -temperature calibrations, with recalculated values shown in Supporting Information Tables S9-10 and Figure 9. Overall, all of the new (non-multivariate) benthic Δ_{47} -T calibrations result in absolute temperatures that are cooler by ~2-2.5 °C, compared to what was originally reported by Taylor et al. (2023) using the planktic calibration of Meinicke et al. (2020). The amplitude of changes in temperature and $\delta^{18}O_{sw}$ are slightly larger or similar to what was published by Taylor et al. (2023).
- 890

891 3.5.2.1 Lower $\delta^{18}O_{sw}$ values using the new benthic Δ_{47} -temperature calibrations

The Taylor et al. (2023) estimates of late Eocene $\delta^{18}O_{sw}$ are 0.35 ‰ to 0.89 ‰ for the early Oligocene. They report an increase of $\delta^{18}O$ seawater of 0.07 ‰ over EOIS and a change of 0.54 ‰ across the transition. As reference, a change of 0.6 ‰ across the transition is estimated to lead to a change of ice volume on the order of 70 %–110 % relative to modern. The new benthic Δ_{47} -temperature calibrations yield absolute $\delta^{18}O_{sw}$ values are ~0.4 to 0.6 ‰ lower, and the changes in $\delta^{18}O_{sw}$ are slightly larger or similar to what was published (Supporting Information Table S9).

A puzzling pattern emerges when looking at calculated changes in $\delta^{18}O_{sw}$ in more detail, 898 899 specifically if we look at the values of $\delta^{18}O_{sw}$ associated with the EOIS from the different calibrations. We would expect a positive change in $\delta^{18}O_{sw}$ associated with the oxygen isotope step, given the growth of ice, 900 901 yet some of the new regressions, from "all benthic-MB", "all benthic", "infaunal" and "epifaunal", yield 902 negative changes in $\delta^{18}O_{sw}$ associated with the EOIS, although they do give positive changes across the 903 broader EOT. For example, the "all benthic-MB" calibration gives late Eocene mean temperatures of 9.3 904 °C, a temperature change associated with the EOIS of -4.6 °C and an associated $\delta^{18}O_{sw}$ change of 0.04 ‰, 905 and a change across the EOT of 0.53 ‰ (Supporting Information Table S9). The "epifaunal" calibration 906 yields 9.6 °C, with a -5.5 °C change and a -0.18 ‰ $\delta^{18}O_{sw}$ change across the EOIS, and a 0.47 ‰ change 907 across the transition. 908

909 3.5.2.2 Reasonable Δ_{47} -T and $\delta^{18}O_{sw}$ values from the new benthic multivariate Δ_{47} -calibration

910 The most robust absolute values and patterns of change in temperature and $\delta^{18}O_{sw}$ are reconstructed

911 using the All benthic (infaunal + epifaunal) multivariate equation. To apply the multivariate calibration, we 912 utilize a series of constraints. Estimates of the increase in $[CO_3^{2-}]$ derived from Li/Ca ratios in benthic

- 913 for a sociated with the Oi-1 glaciation range from ~36 µmol/kg at ODP Site 1218 in the equatorial
- Pacific Ocean (Lear et al., 2010; Lear and Rosenthal, 2006) to $\sim 29 \,\mu$ mol/kg at ODP Site 1263 in the South
- 915 Atlantic on Walvis Ridge (Peck et al., 2010). Modern carbonate profiles in this region have a bottom water
- 916 $[CO_3^{2-}] = 81 \,\mu\text{mol/kg}$ (GLODAP database corrected from anthropogenic inputs, modern profiles), $[CO_3^{2-}]$
- 917]_{sat} = 87 μ mol/kg, and a Δ [CO₃²⁻] = -6 μ mol/kg. If we apply the estimated change in [CO₃²⁻] from ODP Site
- 918 1218 across the Eocene-Oligocene Transition, we calculate a Δ [CO₃²⁻] from -6 to 30 µmol/kg.
- 919

Figure 9: Reconstructed Δ_{47} -based estimates of bottom water temperatures and $\delta^{18}O_{sw}$ for the Eocene-Oligocene Transition. Temperatures are cooler for all of the new regressions from this study (colored lines), and most $\delta^{18}O_{sw}$ values lower, compared to Taylor et al. (2023) (grey line). The multivariate regression

922 and most $\delta^{18}O_{sw}$ values lower, compared to Taylor et al. (2023) (grey line). The multivariate regression 923 shows the largest $\delta^{18}O_{sw}$ increase across the EOT. Light blue and pink lines show Δ_{47} -temperatures based

- 924 on the all benthic (infaunal + epifaunal) multivariate equation at different $\Delta[CO_3^{2-}]$. The dashed line
- combines these lines to correct for $\Delta[CO_3^{2-}]$ across the event. The dark blue line shows the non-multivariate
- 926 Δ_{47} -temperatures with no correction for $\Delta[CO_3^{2-}]$. The grey line shows original reconstruction from Taylor
- 927 et al. (2023) that used planktic calibration from Meinicke et al. (2020) for comparison. All calculations use
- 928 Δ_{47} data from Taylor et al. (2023).

930 Compared to Taylor et al. (2023), the absolute temperatures are cooler, the amplitude of cooling 931 associated with the EOIS is reduced, and changes in $\delta^{18}O_{sw}$ with the EOIS are positive and reasonable in 932 magnitude (Supporting Information Table S10). At ~33.65 Ma, the coolest temperatures of the transition 933 are reconstructed, with absolute bottom water temperatures estimated as 3.9 °C, instead of 6.5 °C. We 934 estimate a cooling of 4.1 °C associated with the EOIS (Figure 9), rather than the 4.7 °C change reported by 935 Taylor et al. (2023). These calculations use a pre- EOIS value of $\Delta [CO_3^{2-}] = -6 \mu mol/kg$, and EOIS value 936 of $\Delta[CO_3^{2-}] = 30 \,\mu\text{mol/kg}$. The recalculated $\delta^{18}O_{sw}$ values (Figure 9, Supporting Information Table S10) 937 support a positive excursion associated with the EOIS, as expected based on evidence for a major expansion 938 of continental ice. The absolute values change from an average of -0.47 to 0.21 ‰ post-transition, with a 939 change of 0.67 ‰, which is similar to the change in Taylor et al. (2023). The calculations do change the 940 $\delta^{18}O_{sw}$ prior to the EOIS which is still closer to but still higher than anticipated based on theoretical 941 calculations of and ice-free world (e.g., -0.9 to -1.2 ‰, Cramer et al., 2011), that could support the presence 942 of some ice storage before this time.

943

944 3.5.3 Cooler Early Eocene Climatic Optimum temperatures from the new benthic Δ_{47} -temperature 945 calibrations

946 Some of the warmest temperatures in the Cenozoic occurred during the Early Eocene Climatic 947 Optimum (EECO; ~52 Ma), based on benthic foraminiferal δ^{18} O (Zachos et al. 2001). For the interval from 948 53 to 51 Ma, published (Meckler et al., 2022) Δ_{47} -temperatures for the Atlantic Ocean range from 10.4 to 949 20.6 °C and $\delta^{18}O_{sw}$ values range from -1.6 to 0.7 ‰, while average EECO temperatures estimated using 950 benthic Δ_{47} are 16.2 °C and δ^{18} O_{sw} estimates average -0.3 %. These Δ_{47} -temperatures use the Meinicke et 951 al. (2020) calibration and are 2-3 °C warmer than for a for a formula δ^{18} O-based estimates, prompting studies to 952 suggest there may be uncertainties in the calibration used, effects relating to carbonate chemistry, and/or 953 warm saline deep water (Meckler et al., 2022; Daeron and Gray, 2023; Rohling et al., 2024).

954 Using the new benthic Δ_{47} -temperature regressions yields temperatures that are about ~1°C cooler. 955 with temperatures ranging from 8.6 to 23.7 °C for the Early Eocene (Figure 10, Supporting Information 956 Table S9). Average EECO temperatures of 14.6 °C and seawater $\delta^{18}O_{sw}$ values of -0.64 ‰ are estimated 957 using the "all foraminifera" calibration, while values of 15.3 °C and -0.51 ‰ are reconstructed using the 958 "all benthics (infaunal + epifaunal)" calibration, values of 16.1°C and -0.34 ‰ from the "epifaunal" 959 calibration and 15.0 °C and -0.58 ‰ from the "infaunal" calibration (Figure 9, Supporting Information 960 Table S9). With a pH correction for carbonate $\delta^{18}O$ as described in Meckler et al. (2022), the $\delta^{18}O_{sw}$ 961 decreases further and is closer to expected ice-free predictions of the early Eocene of ~ 0.9 ‰ (Figure 10, 962 Supporting Information Table S9) (Cramer et al., 2011). These new temperature estimates are slightly 963 higher than Mg/Ca-derived bottom water temperatures at ODP Site 1263 in the Atlantic, where an average 964 bottom water temperature of 14°C was reported (Lauretano et al., 2018) and are still warmer than δ^{18} O-965 based temperatures (Meckler et al., 2022; Daeron and Gray, 2023; Rohling et al., 2024).

966 3.5.4 Cooler Paleocene temperatures from the new benthic Δ_{47} -temperature calibrations

967 Similar to the early Eocene, the Paleocene is characterized by warm global temperatures and ended 968 with the first of a series of hyperthermals defining the transition into the Eocene, the Paleocene-Eocene 969 Thermal Maximum (PETM) (Zachos et al. 2001). For the interval from 65 to 56 Ma, published (Meckler et 970 al., 2022) Δ_{47} -temperatures for the Atlantic Ocean range from 10.9 to 19.9 °C and $\delta^{18}O_{sw}$ values range from 971 -0.35 to 1.33 ‰, while average temperatures estimated using benthic Δ_{47} are 15.2 °C and $\delta^{18}O_{sw}$ estimates 972 average 0.45 %. These Δ_{47} -temperatures use the Meinicke et al. (2020) planktic calibration and are also 2-973 3 °C warmer than foraminiferal δ^{18} O-based estimates (Meckler et al., 2022; Daeron and Gray, 2023; 974 Rohling et al., 2024).

975 Applying the new benthic Δ_{47} -temperature regressions to the published Δ_{47} data yields temperatures 976 that are about ~1.5 °C cooler than what was originally reconstructed (Figure 10, Supporting Information 977 Table S9). Average Paleocene temperatures of 13.6 °C and seawater $\delta^{18}O_{sw}$ values of 0.10 ‰ are estimated 978 using the "all foraminifera" calibration, while values of 14.2 °C and 0.61 ‰ are reconstructed using the "all

979 benthic-MB" calibration, values of 14.9 °C and 0.22 ‰ from the 'epifaunal" calibration and 13.9 °C and

980 0.37 ‰ from the "infaunal" calibration (Figure 10). With a pH correction for carbonate δ^{18} O as described 981 in Meckler et al. (2022), the δ^{18} O_{sw} values decrease to be closer to ice-free predictions (Figure 11e, 982 Supporting Information Table S9). These new estimates are similar to Mg/Ca-derived bottom water 983 temperatures at during the Paleocene (e.g. Cramer et al., 2011) although still warmer than δ^{18} O-based 984 temperatures (Meckler et al., 2022; Daeron and Gray, 2023; Rohling et al., 2024).

985

Figure 10: Comparison of recalculated bottom water temperatures and $\delta^{18}O_{sw}$ with clumped isotope data from Meckler et al. (2022). A) Bottom water temperature (BWT) reconstructions. B) $\delta^{18}O_{sw}$ reconstructions. Temperatures calculated with the new regressions from this study temp to decrease temperatures relative to the Meinicke et al. (2020) calibration. $\delta^{18}O_{sw}$ estimates also decrease with the new calibrations, however the decrease does not itself reconcile the deviation from ice-free predicted $\delta^{18}O_{sw}$ in the early Cenozoic.

991

992 3.5.5 Potential effects on Paleocene-Eocene Thermal Maximum (PETM) Δ_{47} -temperatures

993 Although there currently are no clumped isotope reconstructions published for the PETM, here, we 994 conduct a preliminary assessment of the potential magnitude of effects on Δ_{47} -temperatures. We used [CO₃²⁻] from Zeebe and Zachos (2007) from nine sites, $[CO_3^{2-}]_{sat}$ based on site paleodepths, and the epifaunal 995 benthics multivariable regression to explore the magnitude of potential temperature biases (Table 5). 996 997 Temperature biases from site-specific Δ [CO₃²⁻] values were calculated for three temperatures (15, 20, 35) 998 °C), and are presented in Table 5. For example, the bias in temperature from Δ [CO₃²⁻] ranging from -25 to 999 -5 µmol/kg could lead to a PETM temperature overestimation by 3.2 to 4.3 °C, at a reconstructed 1000 temperature of 15°C (Table 5). Coupling Δ_{47} with site-specific estimates of Δ [CO₃²⁻] would facilitate the 1001 most accurate determinations of temperature.

1002 **Figure 11:** Comparison of recalculated bottom water temperatures and $\delta^{18}O_{sw}$ with clumped isotope data 1003 from Meckler et al. (2022). pCO2 estimates and TA estimates used to calculate $\Delta[CO_3^{2-}]$ B) $[CO_3^{2-}]$ values 1004 calculated as described in section 3.5.4a C) BWT reconstructions using the All benthic (infaunal + 1005 epifaunal) equations where non-corrected are the results of the Deming regression, and corrected are results 1006 from the multivariate regression. D) $\delta^{18}O_{sw}$ reconstructions with E) $\delta^{18}O_{sw}$ reconstructions with a pH 1007 correction in $\delta^{18}O_{carb}$ as described in Meckler et al. (2022).

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1009

1010 3.6 Key takeaways and recommendations for future studies

1011 By roughly doubling the amount of foraminiferal calibration data through new analyses where we 1012 measured small samples using the latest analytical procedures and instrumentation, and utilizing the I-1013 CDES reference frame, we are able to rigorously determine temperature calibrations for foraminiferal Δ_{47} 1014 with reduced uncertainties and also compare results to a large compilation of synthetic calcite Δ_{47} data over 1015 a similar temperature range. We observed no significant difference between the synthetic calcite regression 1016 and the composite foraminiferal calibration, consistent with prior findings.

1017 We investigated non-thermal effects using multiple statistical methods and found evidence for 1018 weak dependencies on carbonate chemistry. R model selection analysis indicated the carbonate ion 1019 saturation of bottom waters has an impact on the Δ_{47} of planktic foraminifera which could result from 1020 dissolution. For benthic foraminiferal reconstructions, we found epifaunal and infaunal data have 1021 overlapping confidence bands. The benthic foraminiferal data analyzed here shows a correlation between

1022 Δ_{47} residuals and carbonate ion saturation within the overall dataset and in epifaunal benthic foraminifera.

1023

Table 5: Estimates of potential PETM temperature biases on temperatures, assuming a carbonate ion effect

1025 on calcification as described by the multivariable regression for epifaunal benthics. Δ [CO₃²⁻] for individual

sites over the PETM estimated from Zeebe and Zachos (2007). The bias in temperature is based on, and is calculated using the difference in temperatures calculated using the $\Delta[CO_3^{2-}] = 50 \,\mu\text{mol/kg}$ (reference) and

1028 the calculated $\Delta[CO_3^{2-}]$ is specific to each site.

Sites	1266C	999	1001	690	1209B	1210B	1211C	1220B	1221C
[CO ₃ ²⁻] (µmol/kg)	37	<37	<37	48	52	54	58	49	50
[CO ₃ ²⁻]sat (µmol/kg)	62	55	57	57	59	62	63	63	67
Δ [CO ₃ ²⁻] (µmol/kg)	-25	-18	-20	-9	-7	-8	-5	-14	-17
Possible biases:									
T = 15 (°C)	4.3	3.9	4.0	3.4	3.3	3.3	3.2	3.7	3.8
T = 20 (°C)	4.5	4.1	4.2	3.6	3.4	3.5	3.3	3.9	4.0
T = 35 (°C)	5.2	4.7	4.9	4.1	4.0	4.1	3.9	4.5	4.7

1029 1030

For planktic reconstructions, we recommend use of the composite foraminiferal calibration for 1031 1032 temperature reconstructions due to the large sample size (n = 260) and the relevant temperature range for 1033 paleoceanographic reconstructions ($\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1744 \pm 0.0154$). We applied this 1034 calibration (and others) to core-top planktics in order to assess the ability of the thermometer to accurately 1035 constrain vertical hydrographic profiles for several regions, and this equation typically yields results that 1036 are the most accurate. Future studies should investigate any potential species-specific dissolution effects 1037 through analysis of transect data or dissolution experiments, and planktic reconstructions could explore the 1038 possible scope for dissolution impacts.

1039 For benthic reconstructions, we recommend using the all foraminiferal calibration (no mixed benthics) ($\Delta_{47} = 0.0383 \pm 0.0013 \ 10^6/T^2 + 0.1635 \pm 0.0154$) or the all benthics calibration (epifaunal and 1040 infaunal) ($\Delta_{47} = 0.0342 \pm 0.0034 \ 10^6/T^2 + 0.2149 \pm 0.0431$), given the uncertainties associated with the 1041 mixed benthic samples. If $\Delta[CO_3^{2-}]$ can be constrained, through modeling and/or with proxy constraints 1042 1043 from Li/Ca or B/Ca, the multivariate all benthics (epifaunal and infaunal) calibration can be used (Δ_{47} = $0.03865 \pm 0.00376 \ 10^{6}/T^{2} + 0.1518 \pm 0.0493 + 0.0001495 \pm 0.0000601$). We also suggest combining 1044 estimates of Δ_{47} -temperature derived using this approach with carbonate δ^{18} O estimates that factor in pH 1045 effects on carbonate δ^{18} O, similar to the approach described in Meckler et al. (2022) to calculate seawater 1046 1047 δ^{18} O.

1048

1049 4 Conclusions

1050 This study presents new clumped isotope data for foraminiferal core-tops made using recently 1051 established best practices with a meta-analysis of published data reported on the same reference frame. By 1052 performing a meta-analysis of planktic and benthic foraminiferal species, we determine possible factors 1053 that contribute to variance in Δ_{47} . The lab-specific data and meta-analysis do not show any species-specific 1054 offsets. Notably, non-thermal effects, including foraminiferal type and bottom water saturation, dataset and 1055 depth, and dataset and photosymbiont presence/absence, only explain ~13% of the variance. However, we 1056 show there is evidence for a weak carbonate ion effect on benthic foraminifera.

1057 The regressions described in this study allow for benthic Δ_{47} -based temperature reconstructions to 1058 align more reasonably with published δ^{18} O estimates of bottom water temperatures and inferred 1059 climatologies throughout the Cenozoic. In particular, the application of a multivariate regression accounting 1060 for a weak carbonate ion effect in benthic foraminifera allows for a first-order investigation of how changes 1061 in Δ [CO₃²⁻] affect Δ ₄₇-based reconstructions for the Pleistocene, Eocene-Oligocene Transition, Early 1062 Eocene Climatic Optimum, and late Paleocene. We find that applying this calibration brings both 1063 temperature and calculated δ ¹⁸O_{sw} into better agreement with model-based deconvolution of benthic 1064 foraminiferal carbonate δ ¹⁸O.

Based on our reconstructions we recommend the use of the composite foraminiferal calibration due to the sample size for planktonic foraminifera reconstructions ($\Delta_{47} = 0.0374 \pm 0.0013 \ 10^6/T^2 + 0.1744 \pm 0.0154$), and the regression of the benthic foraminiferal calibration with mixed benthics removed for benthic reconstructions ($\Delta_{47} = 0.0342 \pm 0.0034 \ 10^6/T^2 + 0.2149 \pm 0.0431$). For benthic foraminifera, when constraints on $\Delta[CO_3^{2-}]$ are available, we recommend the multivariate all benthic (infaunal + epifaunal) calibration for reconstructions ($\Delta_{47} = 0.03865 \pm 0.00376 \ 10^6/T^2 + 0.1518 \pm 0.0493 \ +0.0001495 \pm 1071 \ 0.0000601$).

1072

1073 Author contribution statement

A.T., H.T., and A.V. are joint first authors. Conceptualization: A.T., Data curation: A.T., H.T.,
A.A., C.R.P., H.C., Formal analysis: A.T., H.T., A.V., A.A., B.Z., C.T., C.B., D.B., D.S., F.C., H.C., I.M.,
J.C., M.G., M.K., R.F., R.E., R.C., Funding acquisition: A.T., R.E., R.C., Investigation, Methodology, and
Visualization: A.T., H.T., A.V., C.T., C.B., H.C., I.M., M.G., R.F., R.E., R.C., Resources: A.T., D.S., F.C.,
J.L.S., M.K., R.E., R.C., T.M., Supervision: A.T., R.C., Validation: A.T., Writing – original draft: A.T.,
H.T., A.V., C.T., C.B., D.B., H.C., I.M., M.G., Writing – reviewing and editing: All co-authors.

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1093 Open Research

All data are in the Supporting Information associated with this paper. In addition, on publication, all data will be archived in Pangaea and EarthChem.

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Site	Latitude °N	Longitude °E	Region	Depth [m]
A14	8.0	113.4	Pacific Ocean (West tropical)	1911
CD107 A MC	52.9	-16.9	Atlantic Ocean (North)	3569
CD107 B MC	57.4	-42.0	Atlantic Ocean (North)	1100
CD107 C MC	57.1	-31.0	Atlantic Ocean (North)	1925
CD145 A3200	20.0	65.6	Indian Ocean (Central tropical)	3190
CD145 Al50	23.3	66.7	Indian Ocean (Central tropical)	151
CD94 6B (OMEX)	56.6	9.8	Atlantic (North)	1683
E035	26.7	125.3	Pacific Ocean (North western)	326
ELT47-017-PC	-53.4	72.2	Indian Ocean (South)	958
EN066 GGC 10	6.6	-21.9	Atlantic Ocean (Eastern equitorial)	3527
EN066 GGC 16	5.5	-21.1	Atlantic Ocean (Eastern equitorial)	3160
EN066 GGC 21	4.2	-20.6	Atlantic Ocean (Eastern equitorial)	3792
EN066 GGC 38	4.9	-20.5	Atlantic Ocean (Eastern equitorial)	2937
EN066 GGC 51	5.4	-22.7	Atlantic Ocean (Eastern equitorial)	4263
KNR 166-2-26JPC	24.3	-83.3	Gulf of Mexico	546
KNR166-2-110MC	24.5	-79.2	Atlantic Ocean (Southwest sub tropical)	390
KNR166-2-11MC	24.2	-83.3	Gulf of Mexico	750
KNR166-2-28MC	24.2	-83.3	Gulf of Mexico	648
ML1208-13BB	0	-156	Pacific Ocean (Central tropical)	3050
ML1208-36BB	7	-161	Pacific Ocean (Central tropical)	2855
MW91-9 GGC 15	0.0	158.9	Pacific Ocean (Western equitorial)	2311
MW97 MC-28G	0.0	162.2	Pacific Ocean (Western equitorial)	4325
MW97 MC-45A	0.0	161.0	Pacific Ocean (Western equitorial)	3368
OCE 205-2 BC 51	26.2	-77.7	Atlantic Ocean (Western tropical)	830
OCE 205-2 BC 53	26.2	-77.7	Atlantic Ocean (Western tropical)	1038
OCE 205-2 BC 55	26.2	-77.7	Atlantic Ocean (Western tropical)	1140
OCE 205-2 GGC 43	26.3	-77.7	Atlantic Ocean (Western tropical)	479
MW0691 1BC7	-2.2	157.0	Pacific Ocean (West tropical)	1614
MW0691 1BC3	-2.18	157	Pacific Ocean (West tropical)	1616
MW0691 1.5BC11	-1.0	157.8	Pacific Ocean (West tropical)	2016
MW0691 1.5BC33	-1.0	157.8	Pacific Ocean (West tropical)	2015
MW0691 2.5BC37	0	159	Pacific Ocean (West tropical)	2445
MW0691 3BC16	0	160.4	Pacific Ocean (West tropical)	2959
MW0691 3BC24	0	160.45	Pacific Ocean (West tropical)	2965

MW0691 4.5BC53	0	161.4	Pacific Ocean (West tropical)	3711
MW0691 4BC51	0.0	161	Pacific Ocean (West tropical)	3411
MW0691 5BC54	0	161.4	Pacific Ocean (West tropical)	4025
MW0691 5.5BC58	0	162.22	Pacific Ocean (West tropical)	4341
MW0691 6BC74	0	162.7	Pacific Ocean (West tropical)	4438
VM18-222	-38.6	140.6	Indian Ocean (Southeast)	1904
VM20-133	33	140	Pacific Ocean (North western)	1503
VM21-1	38.7	-72.7	Atlantic Ocean (North)	2186
VM28-230	-5	167	Pacific Ocean (West tropical)	2992
VM34-157	-41	26	Atlantic Ocean (Southeast)	3636
WIND 10B	-29.1	47.5	Indian Ocean	3520
WIND 33B	-11.2	58.8	Indian Ocean	2871
WP7_01	-3.9	156.0	Pacific Ocean (West tropical)	1800

Species

Globorotalia menardii, Trilobatus sacculifer

Globoconella inflata, Neogloboquadrina pachyderma, Orbulina universa

Globigerina bulloides

Globigerina bulloides

Globorotalia menardii, Mixed benthic species

Neogloboquadrina dutertrei, Orbulina universa, Trilobatus sacculifer, Cassidulina species, Uvigerina species, Mixed benthic species

Globigerina bulloides

Pulleniatina obliquiloculata

Globigerina bulloides, Neogloboquadrina pachyderma

Cibicidoides species, Hoeglundina elegans, Mixed benthic species

Cibicidoides species, Hoeglundina elegans, Mixed benthic species

Cibicidoides species, Hoeglundina elegans

Cibicidoides species, Mixed benthic species

Cibicidoides species, Hoeglundina elegans

Globigerinoides ruber, Trilobatus sacculifer

Cibicidoides species, Hoeglundina elegans, Planulina ariminensis, Uvigerina species

Cibicidoides species, Hoeglundina elegans, Uvigerina species

Cibicidoides species, Hoeglundina elegans, Uvigerina species

Globigerinoides ruber, Trilobatus sacculifer

Globigerinoides ruber, Trilobatus sacculifer

Mixed benthic species

Mixed benthic species

Mixed benthic species

Cibicidoides species, Mixed benthic species

Hoeglundina elegans, Mixed benthic species

Hoeglundina elegans, Mixed benthic species

Cibicidoides species, Mixed benthic species

Globerginella siphonifera, Globorotalia tumida, Neogloboquadrina dutertrei, Pulleniatina obliquiloculata, Trilobatus sacculifer, Mixed benthic species

Globerginella siphonifera, Pulleniatina obliquiloculata, Trilobatus sacculifer

Neogloboquadrina dutertrei, Pullentina obliquiloculata, Trilobatus sacculifer

Globerginella siphonifera, Globorotalia tumida, Pulleniatina obliquiloculata, Trilobatus sacculifer

Globorotalia tumida, Pulleniatina obliquiloculata, Trilobatus sacculifer, Mixed benthic species

Globorotalia tumida, Pulleniatina obliquiloculata

Globorotalia tumida, Neogloboquadrina dutertrei, Pulleniatina obliquiloculata, Trilobatus sacculifer

Globerginella siphonifera, Globorotalia tumida, Pulleniatina obliquiloculata

Pulleniatina obliquiloculata, Mixed benthic species

Pulleniatina obliquiloculata

Globorotalia tumida, Pulleniatina obliquiloculata

Globorotalia tumida, Pulleniatina obliquiloculata

Globoconella inflata

Neogloboquadrina dutertrei

Globigerina bulloides, Neogloboquadrina dutertrei

Globigerinoides ruber, Trilobatus sacculifer

Trilobatus sacculifer

Trilobatus sacculifer, Cibicidoides species

Trilobatus sacculifer

Globorotalia menardii, Globorotalia tumida, Neogloboquadrina dutertrei, Pulleniatina obliquiloculata, Trilobatus sacculifer