Review of

Using a foraminiferal ecology model to test if tropical Pacific planktonic foraminifera are suitable recorders of ENSO, version 4 by Metcalfe et al.

Recommendation: Reject

Summary

The manuscript uses a number of forward models to assess the possibility of recovering distinct ENSO states from individual foraminiferal analyses (IFAs). While the goal is excellent and the general approach sound, the presentation is so abstruse that it removes all credibility from the paper's claims. After what looks like 4 trials, it is unclear that another round of revisions could fix these fatals flaws.

1 General Comments

As a prelude, I note that the paper has received prior reviews, which I refrained from reading to avoid biasing my judgment. I apologize in advance if the following comments are redundant, or if they contradict the recommendations of previous referees – I know firsthand that one cannot please every referee.

1.1 Statistics

As said above, the general approach is sound, though the tediousness of the exposition leads the authors to belabor obvious points at the expense of critical explanations.

My biggest question mark is on the kernel density estimation (section 2.4). The authors use an Epanechikov kernel with the following justification "The use of an Epanechnikov kernel distribution to fit the data, as opposed to other types of distribution, represents a trade-off between keeping as many parameters constant whilst mimicking the underlying dataset for a large number of grid points.". I fail to see how this justifies an Epanechikov kernel, as opposed to any other kernel. While the choice of kernel is typically unimportant, the fact that it is so awkwardly justified raises a red flag. If your results depend sensitively on the choice of kernel, you are in deep, deep trouble. It would be important to include (as a supplement) an analysis with a different kernel choice (e.g. Gaussian).

The authors also mention a variable bandwidth, which is fine, but do not explain how it is chosen (e.g. Silvermann criterion). Given that the entire premise of the paper is to compare distributions, this is a crucial detail that needs to be better explained, possibly with a sensitivity analysis.

On the broader point of reporting the results of the Anderson-Darling test, the authors rely

exclusively on whether the p-values are above or below 5%. As emphasized by *Wasserstein and* Lazar [2016], the American Statistical Association explicitly warns against relying exclusively on p-values, and recommends additional metrics like effect sizes and confidence intervals. I think mapping effect sizes, possibly stippled to indicate whether the p-values are above or below 5%, would be better practice.

Finally, and though I would be the last reader to request that the paper get any lengthier, I am surprised that the authors did not focus on the most obvious Achille's heel of IFAs, as practiced, for instance, by *White et al.* [2018]. Contrary to the author's claim, there is nothing inherently wrong with using quantile-quantile plots to compare distributions – indeed it can be a fine idea. The one issue with QQ plots as applied in studies like *White et al.* [2018] is that it is a handful of extreme values that determine the slope, making the results extremely brittle to outliers. To my mind, this is the most urgent statistical issue to address about the way IFAs are currently presented.

1.2 Structure

The paper follows the classic structure of Introduction/Methods/Results/Discussion. The only issue here is that, because they consider 3 distinct questions, the methods are varied and lengthy, and by the time the reader gets to the end of Section 3, they have largely forgotten the relevant methods. It would seem more natural to me to structure one section per question, with relevant methods introduced where needed. This would look like: 1 introduction 2) distinguishing variance statistics 3) distinguishing distributions, 4) sensitivity to input parameters, 5) impacts of dissolution and bioturbation 6) discussion. One thing is for sure: the present structure is extremely indigestible, and squanders any goodwill that the reader might still have after reading that pompous introduction.

1.3 Grammar

I have reviewed dozens of papers over my career, but this one takes the prize for the most abstruse writing coming from native English speakers. A few times I had to look up whether some of the quirks might be differences between British and American English, but I could find no justification in any grammar book for spelling figures "figure's" (P14L30), for starting sentences by "Whilst" followed by a comma, for writing Proustian run-on sentences, or for being generally so incoherent that, after reviewing the paper on my iPad, the number one suggestion from my autocomplete is "incoherent grammar" (see annotated manuscript). I am surprised that experienced scientists like Didier Roche or Claire Waelbroeck let the paper be submitted once, let alone four times, with such flaws.

1.4 Figures

The figures are a piece of work. First, this is the first time I have seen figures so large that they make the text pages of the PDF look like microfilm. To add insult to injury, they are all of different sizes, making the document's navigation extremely tedious. The substance is no better than the style, unfortunately, as (apart from Fig 10), they are all so poorly designed that I would tell my students to redo them. It seems like the authors cannot decide what point to make, so they bombard the reader with lots of similar, overloaded figures. It is imperative to focus the design around the key points, and put the other figures in an appendix/supplement.

2 Line by line Comments

see annotated manuscript.

In summary, this paper is not appropriate for publication in present form, and it is unclear if, after this many trials, it can ever be brought up to that standard.

References

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Using a foraminiferal ecology model to test if tropical Pacific planktonic foraminifera are suitable recorders of ENSO

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- 10 A complete understanding of past El Niño-Southern Oscillation (ENSO) fluctuations is important for the future predictions of regional climate using climate models. One approach to reconstructing past ENSO dynamics uses planktonic foraminifera as recorders of past climate to assess past spatiotemporal changes in upper ocean conditions. In this paper we utilise a model of planktonic foraminifera populations, *Foraminifera as Modelled Entities* (FAME), to forward model the potential monthly average δ¹⁸O_c and temperature signal proxy values for *Globigerinoides ruber*,
- 15 Globigerinoides sacculifer and Neogloboquadrina dutertrei from input variables covering the period of the instrumental record. We tested whether the modelled foraminifera population $\delta^{18}O_c$ and T_c associated with El Niño events statistically differ from the values associated with other climate states. For these foraminiferal species our results suggest that the values of El Niño events can be differentiated from other climate states. Our model computes the proxy values of foraminifera in the water, suggesting that, in theory, water locations for a large proportion of the
- 20 Tropical Pacific should be suitable for differentiating El Niño events from other climate states. However, in practice it may not be possible to differentiate climate states in the sediment record. Specifically, comparison of our model results with the sedimentological features of the Pacific Ocean shows that a large proportion of the hydrographically/ecologically suitable regions, coincide with low sediment accumulation rate at the sea floor and/or regions that lie below the critical water depths for calcite preservation (lysocline and CCD).

25 1. Introduction

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1.1 El Niño-Southern Oscillation (ENSO)

Predictions of short-term, abrupt changes in regional climate are imperative for improving spatiotemporal precision and accuracy when forecasting future climate. Coupled ocean-atmosphere interactions (wind circulation and sea surface temperature) in the tropical Pacific, collectively known as the El Niño-Southern Oscillation (ENSO) on interannual timescales and the Pacific Decadal Oscillation on decadal timescales, represent global climate's largest source (Wang et al.,

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2017) of inter-annual climate variability (Figure 1). Due to ENSO's major socio-economic impacts upon pan-Pacific nations, which, depending on the location, can include flooding, drought and fire risk, it is imperative to have an accurate understanding of both past and future behaviour of ENSO (Trenberth and Otto-Bliesner, 2003; Rosenthal and Broccoli, 2004; McPhaden et al., 2006). The instrumental record of the past century provides important information (that can be

- 5 translated into the Southern Oscillation Index; SOI), however, detailed oceanographic observations of the components of ENSO (both the El Nino and Southern Oscillation), such as the Tropical Oceans Global Atmosphere (TOGA; 1985-1994) experiment only provide information from the latter half of the twentieth century (Wang et al., 2017). To acquire longer records, researchers must turn to the geological record using various archives that are available from the (pan-)Pacific region, including: corals (Cole and Tudhope, 2017); foraminifera (Ford et al., 2015; Garidel-Thoron et al., 2007; Koutavas et al.,
- 10 2006; Koutavas and Joanides, 2012; Koutavas and Lynch-Stieglitz, 2003; Leduc et al., 2009; White et al., 2018); stalagmite (Asmerom et al., 2007; Zhu et al., 2017b); nsn detritus (Patterson et al., 2004; Skrivanek and Hendy, 2015); lake (Anderson et al., 2005; Bennett et al., 2001; Benson et al., 2002; Conroy et al., 2008; Enzel and Wells, 1997; Higley et al., 2018; Loubere et al., 2013; Zhang et al., 2014); terrestrial (Barron et al., 2003; Barron and Anderson, 2011; Caramanica et al., 2018; Hendy et al., 2015; Staines-Urías et al., 2015); and sedimentological parameters (Moy et al., 2002) including varves
- (Du et al., 2018; Nederbragt and Thurow, 2001, 2006) to reconstruct long-term variations in proxies, linked to climate, that may provide clues to ENSO and its impact upon both regional and global climate. An integrated approach combining palaeoclimate proxies (Ford et al., 2015; Garidel-Thoron et al., 2007; Koutavas et al., 2006; Koutavas and Joanides, 2012; Koutavas and Lynch-Stieglitz, 2003; Leduc et al., 2009; White et al., 2018) and computer models (Zhu et al., 2017a) can help shed light on the triggers of past ENSO events, their magnitude and their spatiotemporal distribution. Yet, the
 si mation of past ENSO using climate models has been fraught with difficulties due to the associated feedbacks of ENSO upon model boundary conditions (*e.g.*, SST, *p*CO₂) (Ford et al., 2015). One way to deduce the relative impact and importance of various feedbacks and, in turn, reduce model-dependent noise in our predictions, is to compare model output

with proxy data (Roche et al., 2017; Zhu et al., 2017a).

1.2 Foraminiferal Proxies

- 25 Such an approach, however, requires an abundance of reliable spatiotemporal proxy data from the entire Pacific Ocean. Moreover, such proxy reconstructions are subject to several unknowns, uncertainties and biases. For the specific case of foraminifera populations in the water, it particularly arises from the species-specific ecological niche. The mapping of proxy value to climate value can therefore be skewed, a major factor governing the spatiotemporal distribution of a given planktonic foraminiferal species is the presence of an ideal water temperature. Proxies of past ENSO and Pacific SST (Ford
- 30 et al., 2015; Koutavas et al., 2006; Koutavas and Joanides, 2012; Koutavas and Lynch-Stieglitz, 2003; Leduc et al., 2009; Sadekov et al., 2013; White et al., 2018) are based upon the biomineralisation of the calcite, or a polymorph such as verite (Jacob et al., 2017), shells of foraminifera (Emiliani, 1955; Evans et al., 2018; Zeebe and Wolf-Gladrow, 2001). In general, there are three major types of foraminifera-based palaeoceanographic proxies:

(1) those associated with the faunal composition and their abundance within deep-sea sediments that utilises either a qualitative approach (Phleger et al., 1953; Schott, 1952); a weighted average (Berger and Gardner, 1975; Jones, 1964; Lynts and Judd, 1971); a selected species approach (e.g. coiling direction, or warm-water species presence; Ericson et al., 1964; Ericson and Wollin, 1968; Hutson, 1980b; Parker, 1958; Peeters et al., 2004; Ruddiman, 1971; Schott, 1966); a regression analysis (Hecht, 1973; Imbrie and Kipp, 1971; Williams and Johnson, 1975); or, a transfer function (CLIMAP Project Members, 1976; McIntyre et al., 1976; Williams, 1976; Williams and Johnson, 1975) that compares the down-core records with a dataset of 'modern' values and their associated water column parameters (Hutson, 1977, 1978);

(2) those associated with the stable oxygen isotope composition of a whole shell analysed either individually (Ganssen et al., 2011; Koutavas et al., 2006; Koutavas and Joanides, 2012; Leduc et al., 2009) or pooled (Garidel-Thoron et al., 2007; Koutavas et al., 2002; Stott et al., 2002, 2004), herein $\delta^{18}O_c$ (c = calcite), which can be used to reconstruct SST and past oxygen isotope values in seawater $\delta^{18}O_{sw}$ (sw = seawater) when paired with a proxy that can either reconstruct temperature or salinity;

(3) those associated with trace metal geochemistry (e.g., Ford et al., 2015; Sadekov et al., 2013; Stott et al., 2002, 2004; White et al., 2018), more specifically the natural logarithm of the relative concentration of Mg and Ca (ln(Mg/Ca), of the shell, based upon the temperature dependent (Elderfield and Ganssen, 2000; Nürnberg et al., 1996) incorporation and substitution of a Mg cation into the calcite lattice (Branson et al., 2013, 2016).

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The interpretation of these proxies, however, is not straightforward, for example, calibration of foraminiferal assemblage based transfer functions with surface temperatures as opposed to a deeper temperature signal may in fact skew the
 reconstructed temperature (Telford et al., 2013); δ¹⁸O_c can be affected by species-specific size effects (Feldmeijer et al., 2015; Metcalfe et al., 2015; Pracht et al., 2018). Assequilibria or vital effects, which clouds the accurate reconstruction of past SST and δ¹⁸O_{sw}. There is also no simple bijective function between δ¹⁸O_c and the oceanic variables δ¹⁸O_{sw} and temperature used in its calculation, with variability in δ¹⁸O_{sw} limiting the use of δ¹⁸O_c as a pure temperature proxy. Likewise, researchers have not been able to discount the impacts of the ambient salinity (Allen et al., 2016; Gray et al., 2018; Zeebe and Sanyal, 2002) on the Mg/Ca content of foraminifera, nor biological effects such as growth banding (Eggins et al., 2003; Hori et al., 2018; Sadekov et al., 2008, 2009; Vetter et al., 2013). Foraminifera are also not passive recorders of environmental conditions such as SST, in that the very ambient environment that researchers wish to reconstruct also

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modifies the foraminiferal population as well (Mix, 1987; Mulitza et al., 1998). Sensitivity to the variable being

reconstructed may increase or decrease the relative contribution of individual events through modulation of the flux to the seafloor, increasing or decreasing the chance of sampling such occurrences (Mix, 1987; Mulitza et al., 1998). Culture experiments have identified temperature (Lombard et al., 2009, 2011), light (Bé et al., 1982; Bé and Spero, 1981; Lombard et al., 2010; Rink et al., 1998; Spero, 1987; Wolf-Gladrow et al., 1999), carbonate ion concentration [CO₃²⁻] (Bijma et al.,

5 2002; Lombard et al., 2010) and ontogenetic changes (Hamilton et al., 2008; Wycech et al., 2018) as variables that drive, alter or induce changes in foraminiferal growth.

Computation of the influence of biological and vital effects upon physiochemical proxies, such as those based on foraminifera should be a fundamental consideration for any accurate data-model comparison. Recent attempts at circumnavigating poxy related problems have employed isotope-enabled models (Caley et al., 2014; Roche et al., 2014; Zhu

- 10 et al., 2017a), proxy models (Dolman and Laepple, 2018; Jonkers and Kučera, 2017; Roche et al., 2018) or uncertainty analysis (Thirumalai et al., 2013; Fraass and Lowery, 2017; Dolman and Laepple, 2018) to predict both the potential $\delta^{18}O_c$ values in foraminifera and/or the probability of detection of a climatic event. The use of ecophysiological models (Kageyama et al., 2013; Lombard et al., 2009, 2011) can help circumvent some of the problems associated with a purely mathematical approximation (e.g., Caley et al., 2014) of the translation of an ambient signal into a palaeoclimate proxy. They are not
- 15 limited to foraminifera and can provide an important way to test whether proxies used for palaeoclimate reconstructions are suitable for the given research question. Several studies have investigated the response of planktonic foraminifera from core material or computed pseudo foraminiferal distributions, their proxy values, and the resultant (likely) distribution of these
- proxy values with respect to ENSO (e.g., Leduc et al., 2009; Thirmulai et al., 2013; Ford et al., 2015; Zhu et al., 2017).

1.2 Aims and Objectives

- Here, we investigate whether living planktonic foraminifera can be theoretically used in ENSO reconstructions, differing from previous research by using a foraminiferal growth model, *Foraminifera as modelled entities* (FAME; Roche et al., 2018), to tackle the dynamic seasonal and depth habitat of planktonic foraminifera (Wilke et al., 2006; Steinhardt et al., 2015; Mix, 1987; Mulitza et al., 1998). To be a valid proxy for the reconstruction of ENSO, the proxy values of populations of planktonic foraminifera associated to different climatic states (*i.e.*, El Nino, Neutral, La Nina) should be significantly different from one another. In order to test our research question, '*are the distributions of proxy values associated with El Niño months statistically different from distributions of proxy values associated with neutral or La Niña months?*', our methodology follows a forward modelling approach in which the computed values of the temperature recorded by calcite (T_c a pseudo temperature aimed at mimicking Mg/Ca albeit one uninfluenced by secondary factors) and δ¹⁸O_c are assigned to one of these climatological states. This forward modelling approach does not pre-suppose foraminifera can record ENSO
- 30 variability ('Can we detect?') i.e., what is done when inverting the core top pooled δ^{18} O or individual foraminiferal δ^{18} O distributions and infer changes in ENSO ('How could we detect?'). A secondary objective is to compare the output of this approach with secondary factors that further podulate the climatic signal through post-mortem processes. We identify regions in the Pacific Ocean where the sedimentation rate (Berger, 1970a, 1971; Boltovskoy, 1994; Lougheed et al., 2018;

Olson et al., 2016) may be too low or the water depth (Berger, 1967, 1970b; Boltovskoy, 1966; Lougheed et al., 2018) too deep (causing dissolution of carbonate sediments) thus preventing the capture and preservation of the foraminiferal signal.

2. Methods

2.1 Input variables (Temperature; Salinity and $\delta^{18}O_{sw}$)

- 5 For input variables, temperature and salinity of the ocean reanalysis data product (Universiteit Hamburg, DE) ORA S4 (Balmaseda et al., 2013) were extracted at one-degree resolution for the tropical Pacific (-20°S to 20°N and 120°E to -70°W), with each single grid cell comprised of data for 42 depth intervals (5 5300 m water depth) and 696 months (January 1958 December 2015). For computation of the oxygen isotope of seawater ($\delta^{18}O_{sw}$), a global 1-degree grid was generated, and each grid cell was classified as belonging to one of 27 distinct ocean regions, as defined by either societal and
- 10 scientific agencies, for identifying regional $\delta^{18}O_{sw}$ salinity relationships (LeGrande and Schmidt, 2006). Using the $\delta^{18}O_{sw}$ database of LeGrande and Schmidt (2006) a regional $\delta^{18}O_{sw}$ salinity relationship was defined, of which the salinity is the salinity measured directly at the isotope sample collection point (included within the database). Two matrices were computed; one giving values of the slope (*m*) and the other of intercept (*c*) of the resultant linear regression equations, these were used as look-up tables to define the monthly $\delta^{18}O_{sw}$ from the monthly salinity Ocean reanalysis product ORAS S4
- 15 (Balmaseda et al., 2013), which was used for the calculation of $\delta^{18}O_{eq}$, *i.e.* the expected $\delta^{18}O$ for foraminiferal calcite formed at a certain temperature (Kim and O'Neil, 1997). The $\delta^{18}O_{eq}$ is calculated from a rearranged form of the following temperature equation:

$$T = T_0 - b \cdot (\delta^{18} O_c - \delta^{18} O_{sw}) + a \cdot (\delta^{18} O_c - \delta^{18} O_{sw})^2, (1)$$

Specifically, we used the quadratic approximation (Bemis et al., 1998) of Kim and O'Neil (1997), where $T_0 = 16.1$, a = 0.09, 20 b = -4.64 and converted from V-SMOW to V-PDB using a constant of -0.27 ‰ (Hut, 1987; Roche et al., 2017):

$$\Delta = b^2 - 4a \cdot (T_0 - T_{sw}), (2)$$

$$\delta^{18}O_{c,eq} = \frac{-b - \sqrt{\Delta}}{2a} + \delta^{18}O_{sw} - 0.27, (3)$$

The difference between the constant of Hut (1987) and the dynamic value (Brand et al., 2014) is minor.

2.2 Foraminifera as modelled entities (FAME)

- 25 Foraminifera as modelled entities has been degradoped as a tool for translating, a climatic input (typically a reanalysis dataset or climate model output) into a (simulated-) climatic signal, a signal that aims to approximate the depth integrated growth of foraminifera (e.g., Pracht et al., 2019; Wilke et al., 2006; Steindhardt et al., 2015). Data-model comparison studies suffer from an inability to directly compare like with like so that there are differences in (i) the units used *i.e.*, most proxies reconstructing temperature do not give values of temperature in degrees °C or K but in their own proxy units (e.g., per mil ‰;
- 30 mmol/mol; species abundance or ratio) necessitating a conversion; and (ii) there is a reduction in scales, *i.e.*, models give a

wealth of information (multiple depth layers and high resolution time slices) in the time-depth domain. A number of models and modelling studies exist to determine the foraminiferal responses to present (Fraile et al., 2008, 2009; Kageyama et al., 2013; Kretschmer et al., 2017; Lombard et al., 2009, 2011; Roy et al., 2015; Waterson et al., 2016; Žarić et al., 2005, 2005), past (Fraile et al., 2009; Kretschmer et al., 2016) and future (Roy et al., 2015) climate scenarios, FAME uses the associated temperature and $\delta^{18}O_{eq}$ at each grid cell to compute a time averaged $\delta^{18}O_c$ and T_c for a given species. FAME was produced as

- 5 temperature and $\delta^{18}O_{eq}$ at each grid cell to compute a time averaged $\delta^{18}O_c$ and T_c for a given species. FAME was produced as an attempt to reduce the error associated with data-model comparisons by (i) generating simulated-proxy time-series from model runs that can be compared with age-depth values down core; and (ii) to reduce the model information for a given timeslice into a manageable and relevant value using an integration that would make sense Θ_{e} a biological point of view (Roche et al., 2018).
- 10 The FAME model utilises the temperature-growth rate equations of Lombard *et al.* (2009) to simulate temperature-derived growth rate (Kageyama et al., 2013; Lombard et al., 2009, 2011), this growth rate is then used as a weighing to produce a growth rate-weighted proxy value (Roche et al., 2018). The original Lombard et al. (2009, 2011) equations are based upon a synthesis of culture studies, pooled together irrespective of experimental design or rationale, therefore they can be considered to conceptually represent the fundamental niche of a given foraminiferal species, *i.e.* the range in environment
- 15 that the species can survive. The basic structure of FAME is based upon temperature based Michaelis-Menton kinetics to predict growth rate, described in Lombard et al. (2009), without using the parameters (e.g., light, respiration, food) associated with FORAMCLIM (Lombard et al., 2011). The absence of known values or proxy values for the full set of parameters associated with FORAMCLIM has led us to an Occam's Razor favoured approach in model parameterism on for FAME (Roche et al., 2018). Although other processes may also impact species such as mixed layer depth and nutrients these
- 20 variables for now can be set aside, as temperature provides the dominant signal, it is worth noting that in all probability some variance will arise from these processes and deviation between observed and expected values should consider this. Using the MARGO core top δ¹⁸O_c database (MARGO Project Members*, 2009), Roche *et al.* (2018) validated and computed the optimum depth habitat (the depth habitat that exhibits the strongest correlation when comparing FAME δ¹⁸O_c and MARGO δ¹⁸O_c) for each species in the MARGO database (MARGO Project Members*, 2009). Whilst, both models,
- 25 FAME and FORAMCLIM, can compute the growth rate of eight foraminiferal species from culture studies (Kageyama et al., 2013; Lombard et al., 2009, 2011; Roche et al., 2017), the limited number of species available for a global core top comparison led to a reduction in the number of species modelled (Roche et al., 2018). Here the output of FAME is further restricted to three species that have been the main focus of foraminifera-based studies that have been used to infer ENSO variability, namely the upper ocean dwelling *Globigerinoides sacculifer* and *Globigerinoides ruber*, as well as the
- 30 thermocline dwelling *Neogloboquadrina dutertrei* (Ford et al., 2015; Koutavas et al., 2006; Koutavas and Joanides, 2012; Kuravas and Lynch-Stieglitz, 2003; Leduc et al., 2009; Sadekov et al., 2013). The MARGO database does not include *N*. *dutertrei*, meaning that we concentrate our efforts mainly on *G. ruber* and *G. sacculifer*.

In this study, ORA S4 temperature was used as the input variable, with the growth rate computations artificially constrained to arbitrary values of the upper 60; 100 and 200 m to reflect the presence of photosymbiotic algae in the various

foraminiferal species. By identifying the optimum depth habitat, Roche *et al.* (2018) established the realised niche, *i.e.* the range in environment that the species can be found, for these species for the late Holocene. Unlike some foraminiferal models, FAME does not include limiting factors such as competition, respiration or predation variables as no reliable proxy exists for such parameterisation in the geological record, therefore aspects such as interspecific competition that may limit

- 5 the niche width of a species are not computed. As these depth constraints (<60 m; <100 m; and <200 m) may induce some variability we opted to include an extreme value of <400 m that grossly exaggerates the potential depth window. It is important to note however that as the computation of FAME is based on growth occurring within a temperature window it does not necessarily mean that for a given grid point modelled foraminifera will grow down to 400 m (or whichever cut-off value is used), only that the model in theory can do so (depending if optimal temperature conditions are met) to capture the
- total theoretical niche width. As the optimised depths computed from the MARGO dataset of Roche *et al.* (2018) are shallower, and upper ocean water is more prone to temperature variability, our approach likely dampens both the modelled $\delta^{18}O_c$ and T_c. The modelled growth rate was used to compute the monthly depth-weighted oxygen isotope distribution for each species, using the aforementioned computed $\delta^{18}O_{eq}$ for a given latitudinal and longitudinal grid point (Figures 2 and 3).

This was repeated four times, during which the lower depth limit of the growth rate computation was set to 60; 100; 200 and

15 400 m. No correction for species specific disequilibria, such as vital effect, was applied to the data.

2.4 Statistical analysis

The tropical Pacific Ocean is divided into four Niño regions based on historical ship tracks, from east to west: Niño 1 and 2-(0° to -10°S, 90°W to 80°W), Niño 3 (5°N to -5°S, 150°W to 90°W), Niño 3.4 (5°N to -5°S, 170°W to 120°W) and Niño 4 (5°N to -5°S, 160°E to 150°W). Pan-Pacific meteorological agencies differ in their definition (An and Bong, 2016, 2018) of

- 20 an El Niño, with each country's definition reflecting socio-economic factors. Therefore, for simplicity we use the Oceanic Niño Index (ONI), based upon the Niño 3.4 region (because of the region's importance for interactions between ocean and atmosphere) which is a 3-month running mean of SST anomalies in ERSST.v5 (Huang et al., 2017). We utilise a threshold of χ ≥ +0.5°C (where χ is the value of ONI) as a proxy for El Niño, -0.5°C ≤ χ ≥ +0.5°C for neutral climate conditions and -0.5°C ≤ χ for a La Niña in the Oceanic Niño Index. Many meteorological agencies consider that five consecutive months of χ
- 25 ≥ +0.5°C must occur for the classification of an El Niño event. However, here the only difference is that we consider that any single month falling within our threshold values as representative of El Niño, neutral or La Niña conditions (grey bars in Figure 1). This simplification reflects the lifecycle of planktonic foraminifera (~4 weeks) seeing that the population at time step *t* does not record what happened at *t*-*1* or what will happen at *t*+*1*. As we are producing the mean population growth weighted δ^{18} O values, 'almost' El Niño or 'almost' La Niña would be indistinguishable from the build-up and subsequent
- 30 climb-down of actual El Niño and La Niña events. Therefore, these 'almost' El Niño or 'almost' La Niña are placed within their respective climatological pools as El Niño or La Niña.

Each time-step for the entirety of the Pacific was classified as one of three climate states (El Niño; Neutral; and La Niña), where after the resultant $\delta^{18}O_c$ and T_c at each timestep produced by FAME for each grid-point were binned into their

respective categories. An Epanechnikov-kernel distribution was first fitted to the binned monthly output of a single climate state, the bandwidth varies between grid-points to provide for an optimal kernel distribution. The use of an Epanechnikov-kernel distribution to fit the data, as opposed to other types of distribution, represents a trade-off between keeping as many parameters constant whilst mimicking the underlying dataset for a large number of grid points. The conversion of the data

- 5 from dataset to distribution may induce some small error induced by: rounding to whole integers; the use of a δ¹⁸O_{mid-point} which gives an error associated with the bin size (±0.05 ‰) that is symmetrical close to the distributions measures of central tendency but asymmetrical at the sides; and finally, the associated rounding error at the bin edges within a histogram (±0.005 ‰). Subsequently any two desired distributions can be compared for (dis)similarity using an Anderson-Darling test (1954). Here, all values, i.e., the population, associated with a climatological state are compared with the other populations.
- 10 representing the different climatological state, the results plotted here are Neutral climate state vs. El Niño climate state.

2.5 Test of input data (Temperature and calculated $\delta^{18}O_{eq}$)

Foraminifera as modelled entities produces a modulated response that seeks to replicate how foraminifera modify the climate signal, several studies have approximated the foraminiferal signal in a different way (e.g., Thirmulai et al., 2013; Zhu et al., 2017a). In order to understand how FAME has altered the signal, and the degree to which the conclusions drawn depend upon the modelled growth rates, the input datasets of the sea water properties (Temperature and calculated δ¹⁸O_{eq}), underwent a similar statistical test (Figure 4). Unlike FAME, which integrates over several depth levels using the computed growth rate, the test of the input datasets was with fixed depths without any growth rate weighting. These fixed depths are 5, 149 and 235 m, giving a Eulerian view (Zhu et al., 2017a) in which to observe the implications of FAME's dynamic depth habitat. As per the FAME output, each timestep value was placed into its climate state and an Anderson-Darling test performed to compare the (dis)similarity of on the resultant distributions.

2.6 Alternative statistical tests

In order to compare our results with previously published studies using planktonic foraminifera we employed a series of simple statistical tests, mimicking those applied to sediment archives by the palaeoclimate community. A chief parameter that has been employed in previous ENSO proxy work using foraminiferal analysis (more specifically, individual foraminiferal analysis; IFA) is the measure of individual foraminifera downcore standard deviation (σ(δ¹⁸O_c)). Increased σ(δ¹⁸O_c) is considered to correlate to increased variation in SST and, in turn, increased ENSO incidence and/or magnitude (Leduc et al., 2009; Zhu et al., 2017a) or increased interannual variance (Thirmulai et al., 2013). The variance (σ²(δ¹⁸O_c)) of the timeseries were computed both as the total variance and as the interannual variance, the latter is computed as outlined in Zhu et al. (2017a). For the interannual variance, the mean morthly climatology is subtracted from the dataset, producing monthly anomalies and a linear trend removed (using the detrend function of MatLab 2019a) – the resultant data was left unfiltered (*i.e.*, Zhu et al., 2017a used a 1-2-1 filter). Four 'picking' experiments were performed, as FAME computes the average value for a given time step and given the single foraminiferal isotope variance for an equivalent time step (e.g.,

weeks: Steinhard al., 2015) it is more than likely that this computation reduces the real spread in values. Therefore, rather than use the terminology specimen we prefer to use months. Given the complexity in reconstructions of trace metal geochemistry (Elderfield and Ganssen, 2000; Nürnberg et al., 1996): the potential error associated with determining which carbonate phase is first used when foraminifera biomineralise (Jacob et al., 2017); growth-band integration; secondary factors (e.g., salinity, carbonate ion) the focus of the picking here has been on the $\delta^{18}O_c$. Irrespective of which experiment, 60

- 5 factors (e.g., salinity, carbonate ion) the focus of the picking here has been on the $\delta^{18}O_c$. Irrespective of which experiment, 60 months were drawn, with replacement, and the number of Monte Carlo iterations is set at 10,000. We assume that the 'picker' is taxonomically well-trained and/or has a procedure in which species can be checked taxonomically post-analysis (e.g. photographing all specimens prior to analysis, Pracht et al. (2019)) and therefore do not include an error that deals with incorrect identification. Although we note that parameterisation of misidentification would be difficult, as it requires
- 10 understanding of the variability in both standard deviation and absolute values for species co-occurring downcore (Feldmeijer et al., 2015; Metcalfe et al., 2015; 2019). For each run of experiment's (i) to (iii) the drawn months were saved in order to perform (iv):

(i) In Picking Experiment-I (Figure 3D), the months drawn for each iteration of the Monte Carlo were selected and each grid-point was sampled (i.e., there are 10000.60 selected months). This assumes that the same months are selected at grid point A as point B.

(ii) In Picking Experiment-II (Figure 3E), at each grid-point a Monte Carlo was run (*i.e.*, there are $170 \cdot 40 \cdot 10000 \cdot 60$ selected months). This assumes that different months could be selected between grid point A and point B.

(iii) In Picking Experiment-III (Figure 3F), at each grid-point a Monte Carlo was run using the growth rate weighting for each month (*i.e.*, there are 170.40.10000.60 selected months), this assumes that in periods of higher growth there will be a higher flux of the species and therefore a greater chance of selecting that month. The rationale being that not only are different months selected between grid point A and point B, but if A and B differ climatologically there may be an over subscription of ecologically beneficial habitats in one core location compared to the other.

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(iv) In Picking Experiment-IV (Figures 3G to 3I), the experiment of (ii) was re-run but with the addition of two sources of error: The first error is based upon FAME producing the average value for a given time slice, therefore short-term variability in temperature and/or the spread in the population (*i.e.*, variance in depth of an individual; variance in chamber growth per individual), as evidenced by single foraminiferal analysis of sediment trap samples (*e.g.*, Steinhardt et al., 2015), is potentially lost. Therefore, for each picked month between -0.40 and 0.40 ‰ is added to the picked month value (in intervals of 0.02 ‰), this is approximately

 $\pm 2^{\circ}$ C (*i.e.*, 4° C). The second error is the analytical error that an individual measurement will have. Machine measurement error is assumed to lie between -0.12 and 0.12 ‰ (in intervals of 0.005 ‰ – the 3rd decimal place is an exaggeration of machine capabilities although it will have repercussions for rounding) the 1 σ of within run (as opposed to long-term average) of international stable isotope standards. The intervals of both errors (0.02 ‰ and 0.005 ‰) were chosen to give a similar number (n = 41 and 49) of potential randomly selected error for each picked month. Each picked month has their own error. The values for within month variability (Figures 3G) and machine error (Figure 3H) are calculated separately and then combined (Figure 3I), as they may have a corresponding or conflicting signs, either 'cancelling' out each other or amplifying the difference.

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An associated statistical methodology is the graphical summary (as opposed to a numerical summary via a test value) of plotting the quantiles of two probability or the quantiles of sample probability distribution against a theoretical distribution distributions also referred to as a Quantile-Quantile, or Q-Q plot (*e.g.*, Ford et al., 2015; White et al., 2018). A complimentary (*i.e.*, used in association with, not as replacement, Filliben 1975) test metric, the Probability plot correlation coefficient (Filliben, 1975) can be used as a numerical summation of this approach, which bases its rationale on near linearity between the two tested distributions. This graphical technique is not used here for the following reasons, (i) the climatic categories (*i.e.*, El Nino, Neutral, La Nina) imposed upon the data give uneven sized sample distributions requiring an interpolated quantile estimate; and (ii) the large graphical computation required (170.40).

2.7 An approximation of sedimentary archives: Water depth & Sedimentation Rate

20 Discrete sediment intervals retrieved from systematically bioturbated deep-sea sediment cores contain foraminifera with ages spanning many centuries (Lougheed et al., 2018; Peng et al., 1979). This **c** h contrast to other proxies such as corals (Cole and Tudhope, 2017), speleotherms (Chen et al., 2016) and molluscs (Butler et al., 2013; Milano et al., 2017), where distinct time-specific banding is present (true 'time-series' proxies). The ambient signal following translation into a foraminiferal signal within the water is therefore further modulated by several post mortem processes, which include: the latitudinal-25 longitudinal shift in position of sinking foraminifera - the so-called 'funnel affect' (van Sebille et al., 2015; Deuser et al., 1981); dissolution of calcium carbonate either in the water (Schiebel et al., 2007), at the seafloor, or due to pore fluids; and bioturbation. As mentioned, mixing by bioturbation, results in an apparent smoothing of the downcore, discrete-depth multispecimen signal (Hutson, 1980a; Löwemark, 2007; Löwemark et al., 2005, 2008; Löwemark and Grootes, 2004; Cole and Tudhope, 2017; Mix, 1987), thus leading to the possibility of interpreting single outlying foraminifera values within a 30 specific depth as representing an 'extreme' climate, when they may in fact represent climate from a different time or epoch. This is especially apparent in $\delta^{18}O_c$ where there is a difference temporally of $\delta^{18}O_{sw}$ (e.g., the ice volume effect in glacial and interglacial cycles ~1.25 ‰) meaning that the same temperature can have radically different $\delta^{18}O_c$ values, a consequence of

this is that a series of high magnitude, but low frequency El Niño events could be disturbed in a discrete-depth record.

Therefore, in order to reliably extract short-term environmental information from foraminiferal-based proxies, the signal that one is testing or aiming to recover must have a large enough magnitude, be largely unaffected by dissolution (*i.e.*, above the lysocline) so as not to adversely affect the population and the sedimentation rate must be high enough to give sufficient temporal coverage and rule out upwards bioturbation of single foraminifera from significantly different climate periods.

- 5 In our first step in consideration of post-mortem signal alteration we focus on dissolution. The lysocline, the depth at which dissolution first becomes apparent (Berger, 1968; 1970), and the Calcite (or Calcium Carbonate) Compensation Depth (CCD; Bramlette, 1961) vary between the different ocean basins; the Atlantic Ocean in which deep water forms has a relatively deep CCD as a by-product of 'young' well ventilated bottom waters whereas the Pacific Ocean the final section of the thermohaline circulation conveyor belt, has a shallower CCD. In order to highlight the potential for dissolution, the
- 10 bathymetry of the Pacific was extracted from the General Bathymetric Chart of the Oceans GEBCO 2014 30 arc-second grid (version 20150318, <u>www.gebco.net</u>) between -20°S to 20°N and 120°E to -70°W (Figure 8). Depths of 3500 m below sealevel (bsl), 4000 m bsl and 4500 m bsl arg used here as cut-off values, these depths represent multiple possible depths under which there is the potential for noticeable dissolution (i.e., lysocline) or be dissolved (i.e., CCD). Whilst our intention here is a generalised view to be used as an approximate guide, it is important to note that the Pacific Ocean has the largest
- 15 proportion globally of >1 km tall seamounts that are smaller than <100 km (Wessel, 1997). Which may have important, relatively shallow-water sedimentary sequences, which may also be of sufficient sediment accumulation rite therefore we supplement the GEBCO bathymetric data with the locations of seamounts. However, whilst there are an estimated 50,000 seamounts in the Pacific that are taller than a km (Menard, 1964; Wessel and Lyons, 1997), only 12,000 have been documented on charts (Batiza, 1982), and approximately 291 have been dated (Koppers et al., 2003; Clouard and Bonneville,
- 2005; Hillier, 2007). It is these 291, <1% of the estimated seamounts, we have overlain onto the bathymetric data (Figure 8b), although this number is further reduced as we only plot between 20°S and 20°N. The second step when considering post-mortem signal alteration is the sediment accumulation rate (SAR). We first plot the time-averaged deep-sea SAR (Olson et al., 2016), adapted by Lougheed *et al.* (2018) for the Tropical Pacific (Figure 9). New geochronological tools, such as dual ¹⁴C-δ¹⁸O measurements on single foraminifera (Lougheed et al., 2018), show that low
- 25 sedimentation rate cores can have large variances in age between individual foraminifera present within a discrete 1 cm depth interval (Berger and Heath, 1968; Lougheed et al., 2018). In order to model bioturbation, a number of papers have used a diffusion style approach that reduces the parameters down to sediment mixing intensity and sediment mixing depth (herein referred to as bioturbation depth, BD), although this may be an artificial division purely driven by mathematical need rather than biological constraints (Boudreau, 1998). The bioturbation depth has been shown to have a global average of 9.8
- 30 cm (1σ : ± 4.5 cm) that is independent of both water depth and sedimentation rate (Boudreau, 1998), though likely controlled as a result of the energy efficiency of foraging, *e.g.* deeper burrows may cost more energy to produce than can be offset in extracted food resources, and potential decay in labile food resources with sediment depth. It is not possible to carry out a transient bioturbation model upon the temperature and salinity ocean reanalysis data that we used for FAME, as it only covers half a century of data, whereas thousands of years of input data are required to force a transient bioturbation model.

To investigate how much temporal signal is integrated into discrete-depth intervals for typical tropical Pacific SAR, we, therefore, utilised the single foraminifera sediment accumulation simulator (SEAMUS, Lougheed, 2019) to bioturbate, as the input climate signal (Figure's 9 to 11), 0-40,000 year $\delta^{18}O_w$ of NGRIP (North Greenland Ice Core Project Members, 2004; Rasmussen et al., 2014; Seierstad et al., 2014). The ice core time series is an ideal input for a bioturbation simulator, as it

- 5 represents a highly temporally resolved climate input signal. SEAMUS simulates foraminifera in 10-year timesteps. The use of the NGRIP timeseries here is purely as an input parameter to investigate the effect of bioturbation upon a given climate signal - it is important to stress that by using NGRIP as an input signal for SEAMUS we are neither implying that tropical Pacific cores should have signal similar to NGRIP or inferring some kind of causal relationship. As we seek to investigate the effect of bioturbation, no attempt has been made to modulate the input signal's absolute values to mimic expected $\delta^{18}O_c$
- 10 values and this is why each plot of the synthetic down core time series retains the use of V-SMOW, despite carbonates being required to be V-PDB (Coplen 1995). Keeping all things constant, and varying a single parameter between experiments with SEAMUS, the sediment accumulation rate (SAR) was varied to fixed values of either 1, 2, 5 or 10 cm kyr⁻¹ (representative of typical Pacific SAR) and a bioturbation depth (BD) of either 5, 10 or 15cm based upon the global estimate and it's error bounds (Boudreau, 1998). For each experiment, the selected values of SAR and BD were kept constant for the entire
- 15 SEAMIS model run (i.e., the intensity and magnitude of bioturbation was not varied). In reality, SAR and BD may vary temporally depending on local conditions (*e.g.*, food, oxygen). Finally, the FAME results for the three species are overlaid with a water depth mask that highlights whether grid points are above or below 3500 m below sea-level (mbsl), to also show seafloor areas under the CCD depth, where carbonate material is not preserved (Berger, 1967, 1970b). A comparison be we en water depth and time-averaged deep-sea SAR (Olson et al., 2016), adapted by Lougheed *et al.* (2018) is shown in Figure's 7 and 9.

3. Results

The results of the forward model (Figure 2 and 3) are compared with the input values (Figure 4) in order to identify regions in which the values are statistically distinct for different climate states (Figure's 5-7). These results are then shown against the water depth (Figure's 7 to 9) and the SAR (Figure's 9-11) for the region. The results utilise Foraminifera as Modelled

25 Entities (FAME; Figures 2, 3, 5, 6, and 7); the original Ocean Reanalysis data with computed $\delta^{18}O_{eq}$ (Figure 4); the General Bathymetric Chart of the Oceans (GEBCO; Figures 7 to 9); and the single foraminifera sediment accumulation simulator (SEAMUS; Figures 9 to 11).

3.1 FAME Output: Variance

We compute growth-weighted δ¹⁸O_c (Figure 5 and 7) and temperature (Figure 6 and 7) distributions for each grid cell in the
 fifty-eight year simulation using FAME (Roche et al., 2018), constraining the calculation to the Tropical Pacific Ocean (between -20°S to 20°N and 120°E to -70°W). Our model produces 696 individual monthly maps for all three species

(Figure 2). While two of the three species (*G. ruber* and *G. sacculifer*) have similar ecologies, they show differences in their resultant $\delta^{18}O_c$ for the same ocean conditions (Figure 2). A comparison of our computed variance with measured data (Supplementary Table 1) is made, we compare both the value of the nearest grid-cell and because of the size of the grid and drift of foraminifera (van Sebille et al., 2015) an average of a 3 by 3 grid in which the nearest grid-cell to the core location is

- 5 in the center. A comparison is made with both the iCESM model output and the core's that match this output (Zhu et al., 2017a). For the Late Holocene sample (~1.5 ka) MD02-2529 (08°12.33'N 84°07.32'W; 1619 m) in which *N. dutertrei* individual foraminifera were an set of from >250 μ m (Leduc et al., 2009) giving a calculated standard deviation of measured foraminifera of 0.38 ‰. Whereas, the full ~60 year une series (n = 696) of FAME presented here, gives a standard deviation for all species, at depth cut off 60 m between 0.26 and 0.32 ‰; at depth cut off 100 m between 0.20 and 0.29 ‰; at depth cut
- 10 off 200 m between 0.20 and 0.25 ‰; and at depth cut off 400 m between 0.20 and 0.24 ‰ (see Table 1). Although these values vary if the average of the surrounding grid cells is used (see Table 1). In comparison the iCESM results have the following standard deviation values, for a Eulerian (fixed) depth of 50 m: 0.4 ‰; Eulerian 100 m: 0.6 ‰; and Lagrangian value of 0.49 ‰. There are three samples (Koutavas and Joanides, 2012; Sadekov et al., 2013) located south of core site MD02-2529, these are the Late Holocene (~1.6 ka) samples of V21-30 (01°13'S 89°41'W; 617 m) and (~1.1 ka) V21-29
- 15 (01°03'S 89°21'W; 712 m) in which *G. ruber* was measured individually (Sadekov et al., 2013). For these two sites the measured standard deviation is 0.507 ‰ and 0.510 ‰ for V21-30 and V21-29 respectively (Koutavas and Joanides, 2012). The third core site at a similar location is (~1.6ka) CD38-17P (01°36'04 S 90°25'32W; 2580 m) was not analysed individually, instead replicates of pooled samples of 2 or 3 shells of *N. dutertrei* (Sadekov et al., 2013) were made these measured values give a standard deviation of 0.28 ‰. The full ~60 year time series (n = 696) of FAME presented here, gives
- a standard deviation for all species, at depth cut off 60 m between 0.33 and 0.41 ‰; at depth cut off 100 m between 0.27 and 0.40 ‰ at depth cut off 200 m between 0.25 and 0.35 ‰; and at depth cut off 400 m between 0.25 and 0.34 ‰ (see Table 1). Although these values vary if the average of the surrounding grid cells is used (see Table 1). In comparison the iCESM results have the following standard deviation values, for a Eulerian (fixed) depth of 50 m: 0.53 ‰; Eulerian 100 m: 0.75 ‰; and Lagrangian value of 0.35 ‰.
- The study of ENSO nas focused on whether the variability is entirely in response to ENSO or whether it is dominated by interannual variability (Xie, 1994, 1995; Wang et al 1994, 2010), here the interannual (Figure 3C) and total variance (Figure 3A) was computed and a ratio between the two calculated (Figure 3B; see Supplementary Table 1). Like the same analysis of interannual and total variance computed for iCESM and SODA reanalysis (Carton et al., 2000), outlined in Zhu et al. (2017a), there is also high ratio of interannual to total variance in our computed FAME dataset (Figure 3B). Although there
- 30 are regions in the Eastern Equatorial Pacific wherein this ratio reduces. Despite this reduction, the ratio between total and interannual variance is still above > 0.5.

The Monte-Carlo experiments (Figure 3D-I) highlight the variation in picking a subset of the months, here 60, from the full timeseries. The FAME- $\delta^{18}O_{eq}$ *G. sacculifer* with a depth cut-off of 60 m is plotted here, the values for each grid point is the range in standard deviation (*i.e.*, the maximum standard deviation minus the minimum standard deviation) between iterations

of the Monte-Carlo (n=10,000). The range in standard deviations between iterations is plotted instead of the mean of the standard deviations; with increasing *n* the mean converges toward the sample mean, however as the point of the Monte-Carlo is to generate plausible 'samples' it is more important to take into account the range in possible values which would help to establish the potential variability of subsampling. For the most part, regions with high total variance (Figure 3A) also have a

- 5 larger range in standard deviations between the iterations 'picked'. It is interesting to note that by changing from the same months picked for each grid-point (Monte-Carlo I: Figure 3D) to varying the months picked between grid-points (Monte-Carlo II: Figure 3E or Monte-Carlo III: 3F) the range goes from 'smooth' to a more noisy dataset. Whilst the values plotted here are not the absolute values (as they are the range in standard deviation for a given grid point for the entire 10,000 iterations), it can be seen that some of the interesting comparisons could in essence relate to differences in picking, *i.e.*
- 10 different 'months' picked between grid-points may exacerbate or accentuate differences. Likewise, adding random variability, between -0.4 and 0.4 ‰ (Figure 3G and 3I), may also reduce the differences between areas of high Total variance and low Total variance. Though the values associated with machine error (-0.12 to 0.12 ‰) appear to do little to affect the range (Figure 3H and 3I). Whilst again the values plotted are not the absolute values, the variability added in an attempt to mimic biological variation of a given time slice increases the range of possible standard deviations in regions with low Total
- 15 variance (Figure 3G and 3I). Therefore, understanding the biological variability on shorter timescale (e.g., Steinhardt et al., 2015; Mikis et al., 2019) which, maybe here over exaggerated, may be crucial for understanding discrepancies between cores.

3.2 FAME Output: Anderson-Darling test

- Using a basin-wide statistical test, we examine whether the δ¹⁸O_c values of a given El Niño foraminifera population (*FP_{EN}*)
 and a given non-El Niño ('Neutral conditions') foraminifera population (*FP_{NEU}*) can be expected to be significantly different at any given specific location. Where *FP_{EN}* and *FP_{NEU}* exhibit significantly different distributions, ENSO events can potentially be detected by paleoceanographers. In cases where *FP_{EN}* and *FP_{NEU}* do not exhibit significantly different values, then the chosen species and/or location represent a poor efficie to study ENSO dynamics. Each simulation time step was placed into a climate states: identification of timesteps that represent El Niño (EN), Neutral (NEU), and La Niña climate
 conditions was done using the Oceanic Nino Index (ONI) derivative (Huang et al., 2017) (Figure 1). Comparison, for each species, ME's predicted growth-weighted δ¹⁸O_c and T_c distributions associated with each climate event was done using an Anderson-Darling (AD) test. This statistical test can be used to determine whether or not two distributions can be said to come from the same population. The results of this test are presented in the following way, in which there are four criteria: areas where the population distributions of the two climate states are found to be statistically similar have black grid cells in
- 30 all panels referring to the Anderson-Darling test results (Figure's 4-7); the results in which the areas where the populations distributions of two climate states are found to be statistically distinct are shown with two distinct colour schemes depending on whether a computable error can be included (Grey and Hashed) or not (White).

For FAME- $\delta^{18}O_c$ the results where the populations are dissimilar are either plot as grey and hashed for *G. ruber* and *G. sacculifer* or white for *N. dutertrei* (Figure 5). This is because for these two species (*G. ruber* and *G. sacculifer*) we have the possibility to determine how robust these results are. We use the 1σ values of the observed (FAME) minus expected (MARGO), as computed by Roche et al. (2018) with the MARGO core top $\delta^{18}O_c$ database, as the potential error associated

- 5 with the FAME model. Regions in which the difference between the two populations are larger than the potential error are associated with grey, whereas those less than the potential error as hashed regions (Figure 5A and B), these errors should be seen as a guide rather than a rejection of a site. Because the MARGO database does not contain *N. dutertrei* we have given the panels concerning this species a separate colour scheme, black represents grid-cells for which the two populations cannot be said to be statistically different, white grid-cells are those in which the two populations can be said to be statistically different.
- 10 different (Figure 5C). As we do not have a similar way to calculate the error for T_c , FAME- T_c results are shown (in Figure 6) with this same binary pattern (*i.e.*, white grid-cells are those in which the two populations can be said to be statistically different and black are those in which the two populations can be said to be similar). To reduce the complexity, the overlay of the species Anderson-Darling results (Figure 7) also uses the binary colour scheme (white or black).

Our results show that much of the Pacific Ocean can be considered to have statistically different population between FP_{EN}

- 15 and FP_{NEU} for both δ^{18} O (Figure 5) and T_c (Figure 6). We consider that the likely cause for such a remarkable result is due to FAME computing a weighted average and, therefore, the lack of a signal found exclusively within the regions demarked in Figure 1 as El Niño regions could represent how the temperature signal is integrated via an extension of the growth rate; growing season and depth habitat of distinct foraminiferal populations. Taking into account the FAME- δ^{18} O_c error for *G*. *ruber* and *G. sacculifer*, we have computed regions in which the difference in oxygen isotopes between the two populations
- 20 $(\Delta \delta^{18}O_c)$ compared with the AD-test is smaller than the aforementioned error (Hatching in Figure 5), *i.e.* where the mean difference between FP_{EN} and FP_{NEU} is within the error. The hatched regions in Figure 5 considerably reduce the areal extent of significant difference between FP_{EN} and FP_{NEU} , with the remaining regions aligning with the El Niño 3.4 region (Figure 1). It is important to note that this error relates to the model and in reality, the difference between the climate states could be larger or smaller. No such test was performed on the *N. dutertrei* dataset, because of its absence from the MARGO dataset.
- To further test the model-driven results and to assess if they are still consistent when the depth limitation is varied, the analysis was rerun with depths of 100, 200 and an extreme value of 400 m (Figure 5-7). Whilst it is possible to discern differences between the depths, it is important to note that a large percentage of the tropical Pacific remains accessible to palaeoclimate studies. A shallower depth limitation in the model increases the area for the 'warm' species, suggesting that the influence of a reduced variability in temperature or $\delta^{18}O_{eq}$ and the depth limit causes the differences between FP_{EN}
- 30 and FP_{NEU} to be reduced. Overlaying the results of the Anderson-Darling test for all three species (Figure 7) per depth for 60, 100 and 200 m highlights the areas where multi-species comparisons could be made. To account for potential differences in depth habitat we make a combination of shallower depth for *G. ruber* and deeper depths for *G. sacculifer* and *N. dutertrei* (Pracht et al., 2019) in the final panels (Figures 7D and 7H).

3.3 Test of input parameters (fixed depth: temperature and $\delta^{18}O_{eq}$)

The model-driven results were assessed with the underlying input dataset (temperature and $\delta^{18}O_{eq}$), these underwent the same statistical test (Figure 4), although with fixed depths of 5 m, 149 m and 235 m (see section 2.5). The results for each grid point are presented as either black, grey or hashed. Areas where the population distributions of the two climate states are

- 5 found to be statistically similar have black grid cells. Regions in which the difference between the two populations are larger than the potential error are associated with grey, whereas those less than the potential error as hashed regions. The threshold error (i.e., the difference between the means of each distribution) is for temperature (Figure 4A-C) 0.5 °C and for δ¹⁸O_{eq} (Figure 4D-F) 0.10 ‰, these errors should be seen as a guide rather than a rejection of a site. The results of this fixed depth, non-FAME, test show that the shallowest depths produce populations that are significantly different both in terms of their
- 10 mean values and their distributions. In the upper panel of Figure 4, the canonical El Niño 3.4 region is clearly visible at 5 m depth. Whilst differences exist between Anderson-Darling results for the input data (Figure 4) and the FAME δ^{18} O (Figure 5) and T_c (Figure 6), for instance close to the Panama isthmus, there are significant similarities between the plots. These plots also show that our FAME data (Figure 5-7), in which we allow foraminiferal growth down deeper than the depths in Roche et al. (2018), are a conservative estimate and thus are on the low-end (Figure 4), to account for potential discrepancies with
- 15 depth habitats. In the original paper on depth habitats based upon temperatures derived from $\delta^{18}O_c$, Emiliani (1954) cautioned that the depth habitats obtained would represent a weighted average of the total population, and while foraminiferal depth habitats are likely to vary spatiotemporally, the average depth habitat is skewed toward the dominant signal (Mix, 1987).

3.4 Water depth and SAR

- 20 Our analysis uses reanalysis data for the time period 1958-2015, a hypothetical core that had a comparable resolution would essentially be analogous with a sediment core with a rapid sediment accumulation rate (SAR), representing 50 yr cm⁻¹ (or 20 cm kyr⁻¹). Based on our analysis, such a hypothetical core could allow for the possible disentanglement of El Niño related signals from the climatic signal, but only in a best-case scenario involving minimal bioturbation, which is unlikely in the case of oxygenated waters. Extracting the oxygen saturation (**SO**₂) state, of the Pacific Ocean bottom waters from the Annual
- 25 Climatology WOA13 give values that are predominantly >40 % (Figure 9B). Oxygen saturation is the concentration of Oxygen in a medium against the maximum that can be dissolved in the same medium. Whilst annual variability may exist, it is unlikely that bioturbation would be prevented by low oxygen. Therefore, using a cut off value that has been considered sufficiently high enough to outpace bioturbation (*e.g.*, Koutavas and Lynch-Stieglitz, 2003) of 5 cm kyr⁻¹ (Figure 9A) it can be demonstrated that much of the Pacific has an inferred lower sedimentation rate (< 5 cm kyr⁻¹; Figure 9C) than this cut off
- 30 value. To test the influence of bioturbation, the bioturbation simulator SEAMUS was run using the NGRIP time series. The results of SEAMUS highlight the potential single foraminifera depth displacement that low sedimentation rates can result in (Figure 9). Following the current available geochronological method (i.e., age-depth method) such specimens that are

displaced in depth are assigned the average age of the depth that they were displaced to, which could result in erroneous interpretations of climate variability when analysis such as IFA is applied (Lougheed et al., 2018). The results of SEAMUS are plotted both as time series of the bioturbated 'NGRIP' signal (Figure 10) and as histograms of the probability of finding a particularly pseudo-foraminifera with a given age in the bioturbation depth (Figure 11). As the bioturbation depth varies

5 between 5, 10 and 15 cm for the different simulations of SAR, the histogram in each panel (in Figure 11) represent different thicknesses of sediment, i.e., for Figure 11 panels a, d, g, and j histograms represent data with a BD thickness of 5 cm. Likewise, the timeseries is plotted with the discrete 1 cm depth median age; the median age of the bioturbation depth (Figure 11) is the reason why each timeseries does not 'start' at 0 age (Keigwin and Guilderson, 2009).

The variance within a single depth in a core largely represents the integrated time signal for that depth (Figure 11), as

- 10 opposed to the variance of a climatic signal for an inferred (or measured) average age for the depths in question. The proxy variance will be based both upon a non-uniform distribution in temporal frequency of specimens, i.e., older specimens are few compared to younger specimens. A large proportion of the specimens in the BD come from years that are 'proximal' (i.e., close to the youngest age) this may give undue confidence that the probability of picking a specimen from these years is higher, however the long-tail of the distribution means that there is an equally high chance of picking a specimen that has
- 15 come from several thousand years earlier than the discrete-depth's median age. If we consider for the moment this as picking specimens from a box, there is a high chance of picking from a single box that represents the age you want however there is an equally high chance of picking from numerous boxes with varying age. If the spread in the climatic variable is uniform throughout this time then it can be possible to reproduce a similar signal, although this would not by definition represent the actual spread in the actual climatic variable for a given time, however the spread in the climate variable is unlikely to be
- 20 constant. With a varying spread in the climatic signal bioturbation can introduce the possibility of spurious interpretations, but it is of course more obvious where the measured distributions over-exaggerate the climate signal (e.g., Wit et al., 2013). Furthermore, if we consider that researchers do not pick as randomly as they profess, there is both a size and preservation bias to specimens selected, and size is not constant down-core (e.g., Metcalfe et al., 2015) we can further introduce bias within the dataset. The SEAMUS output that corresponds with our chosen SAR cut-off value of 5 cm kyr⁻¹ (Figure 10 and
- 25 11), the lower limit of our mask (Figure 9), is shown in panels in Figures 10H to 10J and Figures 11G to 11I. It is important to note however, that much of the region for which FAME is calculated upon has inferred sedimentation rates lower than this cut-off value (Figure 9C to 9H).

An additional factor in the post-mortem preservation of the oceanographic signal in foraminiferal shells is whether the shells can be preserved. The GEBCO bathymetry data is binned into 250 m wide bins, and the data normalised to 1.0. As the data

30 contains both bathymetric and topographic (below and above 0 m), the grey area in each histogram represent > 0 m (Figure 8). Whilst, there are differences depending on the cut-off value (Figures 8C to 8E) much of the canonical El Niño 3.4 region (Wang et al., 2017) used in oceanography (Figure 1) is also excluded from these suitable areas. O.e. laying the water depth and the SAR with the Anderson-Darling results (Figure 7) highlights that of the total area where FP_{EN} is significantly different from FP_{NEU} (*i.e.* those areas where planktonic foraminiferal flux is suitable for reconstructing past ENSO

dynamics), only a small proportion corresponds to areas where the sea floor is both above the CCD (< 3500 mbsl) and SAR is at least 5 cm/ka (Figure 9). However, at certain locations, near islands or seamounts, the SAR and water depth may be high enough to allow for a signal to be preserved (Figure 8B) that may not be represented here.

4. Discussion

5 4.1 From Life to Sedimentary Assemblages

Whilst we are principally interested in understanding whether living foraminifera can theoretically reconstruct ENSO, comparison with data requires additional analysis. This is because data-model comparisons are subjective, nominally supposing that the data is the value to be achieved by the model. However, if the foraminifera modulate the original climate signal, then preservation selectively filters which specimens are conserved whereas bioturbation acts to reorder, transposing

- 10 the order in which they are recovered from the depth domain. Once the sediment is recovered, the researcher acts as a final filter, which is in essence a random picking. Although technically most researchers will pick whole shells so alongside size selectivity (*e.g.*, Metcalfe et al., 2015) there is also preservation bias associated with picking of foraminifera (*e.g.* Koutavas and Lynch-Stieglitz, 2003). Whilst the presence of depths in the ocean whereupon calcite is absent from sediments was described in the earliest work (*e.g.*, Murray and Renaud, 1891; Sverdrup, 1942), overlaying maps of measured surface
- 15 sediment carbonate percentage with water depth in the Pacific Ocean led Bramlette (1961) to coin the term 'compensation depth' (Wise, 1978). This work highlighted the 'narrow' depths (4-5000 m) in the Central Pacific of the CCD. Concess ally Berger (1971) placed three levels in the Pacific ocean that were descriptive of the aspects (e.g., chemical, palaeontological and sedimentological) of the calcite budget; the saturation depth, demarking supersaturated from undersaturated; the lyscoline, the depth at which dissolution becomes noticeable (Berger 1968, 1971); and compensation depth (Bramlette,
- 20 1961), in which supply is compensated through dissolution. The aspects of the lysocline was estimated by the faunal assemblages of Parker and Berger (1971, figure's 14 and 15 of that publication), for much of the equatorial Pacific the lysocline is estimated at ~3800 m. As the lysocline is where dissolution becomes apparent, ergo it is a sample already visibly degraded, we therefore set the limit of the water depth mask shallower, at 3500 m bsl. In fact, in regions of high fertility, such as the Eastern Equatorial Pacific, the lysocline was estimated to be present at ~2800 m (Thunell et al., 1981) or ~3000
- m (Berger, 1971; Parker and Berger, 1971). For instance, core V21-28 close to the Galapagos Islands (01°05'N, 87°17'W) has a shallower dissolution with the either of these two values despite being collected from a water depth of 2714 m (Luz, 1973). A comparison between the hydrographic and sedimentary lyscoline, using a mooring in the Panama Basin showed that the sedimentary lyscocline is a product of where the hydrographic lyscocline meets the seafloor (Thunell et al., 1981), therefore, this could lead to dissolution within the water of the settling flux (e.g., Schiebel et al., 2007). In the EEP region the
- 30 shallower lyscoline is accompanied by an equally shallower CCD (located at \sim 3600 m) for which the highly fertile is considered responsible for its shoaling, lowering the pH through increased CO₂ (Berger et al., 1976). The correspondence between lyscoline depth and CCD depth does not hold true for the entirety of the Pacific, plotting a N-S cross-section from

50°N to 50°S Berger (1971) noted that in the Central Equatorial Pacific, the high fertility region generates a larger zone of dissolution resistant facies even with a shoaled lysocline. If we factor in the sedimentation rate of the Pacific, which has been estimated to be considerably lower than 1 cm (Blackman and Somayajulu, 1966; Berger, 1969; Menard, 1964), then dissolution may become further exacerbated. The longer a shell remains at the sediment-water interface the greater the

- 5 prospects for it to be dissolved become, therefore low SAR increases the chance of dissolution (Bramlette, 1961). For instance, in 15 equatorial Pacific cores, below 4000 m, the average SAR was presented (Hays et al., 1969; here calculated) at 0.96 cm kyr⁻¹ ($1\sigma \pm 0.43$ cm kyr⁻¹). Although there are regions and/or core locations in which the SAR is higher, for instance eight EEP cores shallower than the lysocline depth (Thunell et al., 1981) of ~2800 m were presented by Koutavas and Lynch-Stieglitz (2003) which have an average SAR, calculated at 7.20 cm kyr⁻¹ ($1\sigma \pm 2.82$ cm kyr⁻¹). The average age for
- 10 these same core's 0 cm core depth is 2184 years ($1\sigma \pm 1521$ yrs), whilst it cannot be assumed that there has been no loss during recovery (i.e., core top is not sediment-water interface), a non-zero core top age is expected for both bioturbation (Keigwin and Guilderson, 2009) and dissolution. Alongside, the potential for dissolution there is the also the mixing of ocean sediments by the benthos (Bramlette and Bradley, 1942). For instance, Arrhenius (1961) noted that ash beds present in cores of the EEP (Worzel, 1959; Ewing et al., 1959) had a 2-3 cm layer above and below what should have originally been a
- 15 sharp boundary in which they estimated that ~50% of the material originated from the other side of the boundary. If one assumes 1 cm kyr⁻¹ sedimentation rates, then the range in age of the obviously 6 cm mixed sediments is minimally ~6000 years per cm, comparison with an analogous SEAMUS simulation (bioturbation depth 5 cm; SAR 1cm) highlights the considerable spread in age, placing the 95.45% range between 110 and 18954 years (Figure 11). Much of this temporal variability (either through bioturbation or dissolution) will be hidden, especially when proxy values correspond with the
- 20 expected values, and more obvious when the values are larger than expected (*e.g.*, Wit et al., 2013). Owing to the lack of absolute variability during the Holocene the apparent confirmation of similarity between we values and modern distributions of the '*to be reconstructed*' variable is not a confirmation of proxy reliability. Especially in the tropics wherein seasonal variability is limited. The effects of both bioturbation and dissolution are further amplified when combined with finite sampling strategies. Therefore, the results of the sedimentological features, presented here, imply that much of the
- 25 Pacific Ocean is not suitable for preserving (Figures 7-9) the ENSO signal, despite the possibility of the species of foraminifera having unique values for different climate states (Figures 4-7). NSO studies using palaeoceanography have exposed shifts, one can, therefore, question what is being reconstructed in such studies.

4.2 Palaeoceanographic Implications

4.2.1 Pacific climate reconstructions

30 One artefact of sampling (Dolman and Laepple, 2018) is the potential occurrence of aliasing (Pisias and Mix, 1988; Wunsch, 2000; Wunsch and Gunn, 2003): a fundamental problem with proxy records is that they can be confounded by local regional climate, and/or ENSO's teleconnections, that mimic ENSO changes albeit at a different temporal frequency. Our own

analysis using our FAME $\delta^{18}O_c$ and T_c output mimics foraminiferal sedimentary archives, pooling several decades worth of data in which the resolution is coarse enough to obscure and prevent individual El Niño events being visible but allowing for some kind of long-term mean state of ENSO activity to be reconstructed (Cole and Tudhope, 2017). The results of our Anderson-Darling test may be unduly influenced by the Pacific decadal variability (PDV), also referred to as the Pacific

- 5 Decadal Oscillation (PDO) (Pena et al., 2008). In much of the tropical Pacific the ratio of decadal to interannual σ SST suggests that they are comparable in magnitude, therefore fluctuations in SST are more obviously apparent outside of the purely canonical regions of ENSO (Wang et al., 2017). It could be that the areas outside of these canonical ENSO regions (Figure 1) reflect the PDO (Pena et al., 2008; Wang et al., 2017). The use of the variance $\sigma^2(\delta^{18}O_c)$, or standard deviation $\sigma(\delta^{18}O_c)$, as an indicator of ENSO is dependent on whether the original climate signal's variance was or was not dominated
- by interannual variance. Zhu *et al.* (2017) computed the total variance change with and without the annual cycle suggesting that, for some cores the increased assumed ENSO variability at the LGM as deduced by proxy records (Koutavas et al., 2006; Koutavas and Joanides, 2012; Koutavas and Lynch-Stieglitz, 2003) may be purely a by-product of the annual cycle or dominated by it. Although the values of El Nino can be considered significantly different from other climate states, our own analysis using the ratio of total context and variance also suggests that much of the variance in the simulated foraminiferal
- 15 signal is dominated by interannual variance. There are differences in the ratio of total to interannual variance between species and in different regions of the tropical Pacific, however, even with a dynamic depth habitat the ratio is still high (Figure 3; Supplementary Table 1).

4.2.2 The use of models in reconsulations

- Reconstructions of the past climate of the Pacific have inferred a relatively weaker Walker circulation, a displaced ITCZ and
 equatorial cooling (Koutavas and Lynch-Stieglitz, 2003); both a reduction (Koutavas and Lynch-Stieglitz, 2003) and intensification (Dubois et al., 2009) in eastern equatorial Pacific upwelling; and both weakened (Leduc et al., 2009) and strengthened ENSO variability (Koutavas and Joanides, 2012; Sadekov et al., 2013). However, a number of the inferences are contentious, for instance the reduction in upwelling in this region (Koutavas and Lynch-Stieglitz, 2003) is contradicted by Dubois et al. (2009), who used alkenones (i.e., U^{KI}₃₇ ratios) to suggest an upwelling intensification (Zhang et al., 2017).
 Whilst the U^{KI}₃₇ proxy has problems within coastal upwelling sites (Kienast et al., 2012) it does not discount their claim, especially considering that δ¹⁸O records can themse¹⁴₁₄₅ be influenced by salinity upon the δ¹⁸O_{sw} component (Rincón-Martínez et al., 2011) and the potential influence of [CO₃²] upon foraminiferal δ¹⁸O_c (de Nooijer et al., 2009; Spero et al., 1997; Spero and Lea, 1996). The discrepancies in reconstructed climate between marine cores' is worth noting, as ultimately it is from proxies that inferences are made about past climate (Trenberth and Otto-Bliesner, 2003; Rosenthal and Broccoli,
- 30 2004). Such inferences have suggest that the past climate of the Pacific region (from the geologically recent too deep time) ^C has been in an: El Niño state (Koutavas et al., 2002; Stott et al., 2002; Koutavas and Lynch-Stieglitz, 2003); permanent El Niño state (Huber and Caballero, 2003) or Super El Niño state (Stott et al., 2002); La Niña state (Andreasen et al., 2001; Beaufort et al., 2001; Martinez et al., 2003); or a different climatic state (Pisias and Mix, 1997; Feldberg and Mix, 2003).

The possibility of a marine sediment archive being able to reconstruct ENSO dynamics comes down to several fundamentals besides whether the signal can or cannot be preserved (*i.e.*, whether the core site has either too low SAR, too high BD or a water depth not conducive to calcite preservation): the time-period captured by the sediment intervals (a combination of SAR and bioturbation); the frequency and intensity of ENSO events; the foraminiferal abundance during ENSO and non-

- 5 ENSO conditions; as well as what the proxy is recording. There is also the presumption that a particular climate event should be recorded, our Anderson-Darling test for instance highlights that there are locations that cannot discern the difference between El Niño and other climate states whilst for the same time period there are locations where the different climate states can be differentiated. Whilst our analysis is a statistical treatment of the data, each species, and different types of phyto- or zooplankton preserved in ocean sediments, are likely to record the same set of environmental conditions differently
- 10 (Mix, 2006). This is, in brief, the rationale for the development of FAME, the same climate signal seen through the view of species-specific proxies will give a fractured view constrained by each species particular ecophysiological constraints (Mix, 1987; Roche et al., 2018). A dynamic depth habitat in which the environmental signal becomes a weighted average of the water column can further confound the original signal (Wilke et al., 2006). What can be seen as contradictory reconstructions can therefore be viewed as the prevailing or dominant conditions at a given location at the time when
- 15 environmental conditions overlap ecological constraints for a given species.

Terrestrial records suggest the number of El Niño events per century in the early Holocene (8-6 ka BP) was minimal (Moy et a., 2002), with between 0 and 10 events occurring per century. This dampened ENSO is observed within lake core colour intensity and records driven primarily by precipitation - although like other proxies this can also be interpreted differently, *i.e.* as a large change in the hydrological cycle shifting precipitation away regionally (Trenberth and Otto-Bliesner, 2003). If

- 20 we assume for now that the number and magnitude of ENSO events was reduced, the relatively low downcore resolution of marine records may not accurately capture the dynamics of such lower amplitude ENSO events using existing methods. The sensitivity and probability of detecting a change in IFA with changes in frequency and amplitude, has been dealt with before (Thirmulai et al., 2013), although without considering bioturbation. The synthesis of pseudo-timeseries to discern the potential distribution for different scenarics; whilst a necessary approximation, is nonetheless one that is free of cause and
- 25 causality. Modulating a timeseries for events with enhanced or weakened amplitude or fewer or greater number of events assumes in essence that there is limited feedback both regionally (between two sites) and internally within the timeseries (i.e., a process that operates on a higher level). Reconstructions of the past can benefit from inclusion within conceptual frameworks that incorporate both data and modelling studies (e.g., Trenberth and Otto-Bliesner, 2003; Rosenthal and Broccoli, 2004; McPhaden et al., 2006). The use of coupled ocean-atmosphere models (e.g., Clement et al., 1999; Zebiak
- 30 and Cane, 1987); isotope enabled Earth system models (e.g., iCESM; Zhu et al., 2017); or multi-model ensembles with prescribed boundary conditions can be used for the generation of timeseries in which the physics of atmospheric and oceanic circulation are constrained and feedbacks between sites can occur. The perceived failure of several climate models to resolve ENSO adequately, resulting in variable ENSO frequency and amplitude between models, could therefore be used to determine the proxy signal from model derived timeseries at different frequencies and intensities of ENSO. Albeit a

timeseries of variable ENSO that is grounded in ocean-atmosphere coupling. Such analysis could also provide information on a secondary assumption, in which time slices from the same core inherently assume that where a particular oceanographic feature exists now is also where it may have existed before. This gives a somewhat binary view, the feature either occurs or does not occur, and if it occurs then it has either enhanced or weakened. Yet this can (though not always) preclude a scenario

- 5 in which the feature has shifted. Analysis of the El Nino patterns suggests that there are two types of El Nino that are spatially delineated: the dateline Central Pacific El Nino and the Eastern Pacific El Nino. The expansion, contraction or shift of certain large scale oceanographic features (e.g., Polar Front, Upwelling) during periods of warmer than average (e.g., the last interglacial) or colder than average temperatures (e.g. the LGM) can complicate the comparison of two down core samples, i.e., a static core continuously recording a particular climate event as opposed to a shifting oceanographic regime
 10 passing over or beyond a core site (Weyl, 1978). Climate models could therefore also be used to determine applicable core
- locations for comparison of proxy values with *'like with like'* oceanographic features (similar to the analysis of Evans et al. (1998) for predicting coral sites), without necessarily the cost of a time-slice project (e.g., CLIMAP, MARGO).

4.2 Limitations of the methods applied and assessment of model uncertainties

- For simplicity we have assumed that our model is 'perfect', of course that is inaccurate, there are four potential sources of 15 error: the input variables (temperature, salinity and their conversion into $\delta^{18}O_{sw}$ and $\delta^{18}O_{eq}$); the model's error with respect to real world values (Roche et al., 2018); the statistical test's errors (associated Type I – in which attribution of significance is given to an insignificant random event, a false 'positive' – and Type II – in which a significant event is attributed to be insignificant, a false 'negative' - errors); and reducing the complexities of foraminiferal biology via parameterization. The input variables can have errors associated with both the absolute values of temperature and salinity used here; and the
- 20 limitation of input values to a single value per month. Whilst it is possible to interpolate to a daily resolution, this is problematic for two reasons: (1) daily temperature records have much more high frequency oscillations than the data here and (2) the lifecycle of a single foraminifera is approximately monthly, therefore by using monthly data it provides an estimate of the average population signal. Conversion of salinity and temperature into $\delta^{18}O_{sw}$ and $\delta^{18}O_{eq}$ uses a quadratic approximation, one source of error is the unknown influence of carbonate ion concentration on both the Kim and O'Neil
- 25 (1997) equation and the foraminiferal microenvironment (de Nooijer et al., 2008, 2009; Spero et al., 1997; Spero and DeNiro, 1987; Spero and Lea, 1996) which has further implications due to the upwelling of cool, low *p*H, waters in the eastern Tropical Pacific (Cole and Tudhope, 2017; Raven et al., 2005). The spatial variability in salinity, particularly within regions underlying the intertropical convergence zone (ITCZ) and the moisture transport from the Caribbean into the eastern Pacific along the top regraphic low that represents Panama Isthmus, the resultant conversion of salinity to $\delta^{18}O_{sw}$ and then
- 30 $\delta^{18}O_{eq}$ may contain further error. If such errors are independent of the absolute value of the variable, *i.e.* the error on cold temperature is the same and not larger than warm temperatures, then the error terms effectively cancel one another out. A point of note, is that the $\delta^{18}O$ to °C conversion of Kim and O'Neil (1997) is considered to be marginally larger at the cold end then at the warm end (0.2 ‰ per 1°C to 0.22 ‰ per 1°C) than that originally discerned (O'Neil et al., 1969).

The comparison of the pseudo-Mg/Ca temperature signal produced here (T_c) to a value corresponding to that reconstructed from measurements of Mg/Ca should be done with caution. Computation of pseudo-foraminiferal δ^{18} O in FAME is aided by the ability to compute an initial δ^{18} O equilibrium value for a given latitude-longitude grid-point and timestep. The weighting of δ^{18} O value used in FAME is an approximation of the foraminiferal shell, chambers are generally homogenous in δ^{18} O

- 5 value excluding either terminal features such as crust or gametogenic calcite which can lead to chamber heterogeneities (e.g., Wycech et al., 2018) although this can be approximated with an additional parameter (Roche et al., 2018). The same cannot be said for Mg/Ca, alongside heterogeneities in the shell which may be the result of diurnal processes (terminal features in the computation of δ^{18} O are technically simpler to model), there are differences in both sample preparation and measurement techniques. Whilst, the change in Mg/Ca with temperature has been validated (*e.g.*, Elderfield and Ganssen, 2000) the
- 10 computation of a pseudo-proxy value for and from model parameters remains enigmatic. Construction of a matrix of equilibrium Mg/Ca would ideally be the most logical step in a second generation of the FAME model. Whilst, simply solving the Mg/Ca palaeotemperature equation for an input of T and an output Mg/Ca is a first approximation, as stated previously several other parameters can alter this technique, this includes abiotic effects such as salinity (Allen et al., 2016; Gray et al., 2018; Groeneveld et al., 2008; Kısakürek et al., 2008) or carbonate ion concentration (Allen et al., 2016; Evans et
- al., 2018; Zeebe and Sanyal, 2002); biotic effects such as diurnal calcification (Eggins et al., 2003; Hori et al., 2018; Sadekov et al., 2008, 2009; Vetter et al., 2013); or additional factors such as sediment (Fallet et al., 2009; Feldmeijer et al., 2013) or specimen (Barker et al., 2003; Greaves et al., 2005) 'cleaning' techniques. Given the role of Mg in inhibiting calcium carbonate formation, the manipulation of seawater similar to the modification of the cell's *p*H (de Nooijer et al., 2008, 2009) may aid calcification and explain the formation of low-Mg by certain foraminifera (Zeebe and Sanyal, 2002).
 Scaling these processes up to a basin-wide model is beyond the remit of this current paper.
- 20 Scaling these processes up to a basin-wide model is beyond the remit of this current paper. Our modelling results also depend upon the species symbiotic nature and potential genotypes. For instance, mixotrophs, those organisms that utilise a mixture of sources for energy and carbon (planktonic foraminifera such as *G. ruber*; and/or *G. sacculifer*) can outcompete heterotrophic (or photoheterotrophic) organisms (planktonic foraminifera such as *Neogloboquadrina pachyderma*; *Neogloboquadrina incompta*) especially in stratified-oligotrophic waters. Whilst FAME
- 25 uses only the temperature component of FORAMCLIM (Roche et al., 2018) and we have only modelled *N. dutertrei*, it is important to note that there are distinctions between the fundamental niche that FAME computes, *i.e.* the conditions that an organism can survive, and the realised niche, *i.e.* what an organism actually occupies given limiting factors within the environment. Likewise, FAME and FORAMCLIM are based upon the original culture experiments that assumed that both species (*G. bulloides* and *N. dutertrei*) are non-symbiotic or have species associations (see Bird et al., 2018, 2017). A species
- 30 that hosts symbionts will likely have a restricted temperature that is associated with the temperature tolerance of their symbionts, given that the next generation of a species of planktonic foraminifera must be re-infected with their symbionts. Likewise, cryptic speciation may lead to foraminiferal genotypes exhibiting distinct environmental preferences (Bird et al., 2018, 2017; Darling et al., 2004, 2000, 1999; Huber et al., 1997; Morard et al., 2013; de Vargas et al., 1999, 2002). Incorporation of both a theoretical genotype abundance (Morard et al., 2013) and ecophysiological tolerances of different

genotypes (Bird et al., 2018) within an ecophysiological model could further reduce error within modelling of planktonic foraminiferal habitats, and thus reduce data-model comparison error. For instance, Morard *et al.* (2013) simulated the impact of genotypes upon palaeoceanographic reconstructions (in particular transfer functions) using a theoretical abundance, calculated with a best-fit gaussian response model, depending upon SST later using a similar approach (Morard et al., 2016) to deduce the impact upon δ^{18} O.

Conclusion

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Concentrating on the period spanning the instrumental record by using the FAME module, we forward modelled the speciesspecific (*i.e.*, *G. ruber*; *G. sacculifer* and *N. dutertrei*) oxygen isotope values ($\delta^{18}O_c$) and pseudo-Temperature (T_c), computed from ocean reanalysis data. The aim of this study was to determine whether the modelled values from different climate states

- 10 are statistically different, our results suggest for large expanses of the Tropical Pacific the climate states do have different values. Whilst, the results show that the values between El Nino states and Neutral climate states are statistically different for a large proportion of the Tropical Pacific, the total variance is dominated by the interannual variance for much of the region. Overlaying our computed foraminiferal distributions with the characteristics of the Pacific Ocean we infer that much of the signal recorded in foraminifera corresponds to areas where several processes will alter the preservation of the foraminiferal
- 15 signal. First, the inferred SAR for much of the region is critically low, and a simulation of bioturbation for different bioturbation depths and SAR, typical for the Pacific indicates that discrete core depths can have a large temporal spread in single foraminifera, possibly precluding the extraction of ENSO-related climate variability. Second, a large proportion of the seafloor lies below the lysocline, the depth at which dissolution of foraminifera becomes apparent. These factors reduce the size of the area available for reconstructions considerably, thus arguably precluding the extraction of a temporally valid
- 20 palaeoclimate signal using long-standing methods. It is our inference that only at exceptional ocean sediment core sites is it possible to determine the variability in ENSO, which makes it difficult to build a Pacific basin-wide understanding of past ENSO dynamics.

Code and data availability

The ocean reanalysis data used in this paper are available from the Universiteit Hamburg. An open source version of the 25 FAME code is available from Roche et al. (2018). Statistical routines are available as part of the Statistical package of MATLAB R2018a; mapping tools (including the topographic colormap) are part of the Mapping Toolbox. The function to **GEBCO** www.gebco.net) retrieve bathymetry (data available at from netcdf format. gebconetcdf(FILE,Wlon,Elon,Slat,Nlat), is available from the MATLAB Central File Exchange (https://mathworks.com/matlabcentral/fileexchange/46669-gebconetcdf-file-wlon-elon-slat-nlat). The single foraminifera

30 sediment accumulation simulator (SEAMUS) is published in Lougheed (2019), available at https://doi.org/10.5194/gmd-

<u>2019-155</u>. A video of the $\delta^{18}O_{\text{shell}}$ output has been archived online (<u>https://doi.org/10.5281/zenodo.2554843</u>, Metcalfe et al., 2019).

Author Contributions

B.M. and D.M.R. designed the study. B.M. analysed the data. B.C.L. processed ocean SAR and depth data, and ran the bioturbation model. B.M. drafted the manuscript with contributions from all authors.

Competing Interests

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The authors declare no competing interests.

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- running the SEAMUS model. We thank the Universiteit Hamburg for their online access server for ocean reanalysis data.

- 5 Figure 1. Oceanic Niño Index and the temperature anomaly for a single El Niño event. (Top) Oceanic Niño Index (ONI), sourced from a 3-month running mean of SST anomalies in ERSST.v5 of the Niño 3.4 region (Huang et al., 2017). Grey vertical bars represent the periods in which El Niño-like conditions exist using a simple one-month threshold. (Bottom) The sea surface temperature difference between week beginning 1st December 1997 minus the long-term climatic mean (1971 2000) for December. The 1997 1998 El Niño represents an EP-ENSO. The long term monthly climatology, the NOAA optimum
- 10 interpolation (OI) SST V2, based upon the methodology of Reynolds and Smith (Reynolds and Smith, 1995) using two distinct climatologies for 1971 2000 and 1982 2000 (Reynolds et al., 2002). Boxes represent the Nino region: Niño 1 and 2 (0° -10°S, 90°W 80°W), Niño 3 (5°N -5°S, 150°W 90°W), Niño 3.4 (5°N -5°S, 170°W 120°W) and Niño 4 (5°N -5°S, 160°E 150°W).

Figure 2. A snapshot of the output of FAME. Each panel represents an individual species (Top Panel *G. sacculifer*; Middle Panel *G. ruber* and Bottom Panel *N. dutertrei*) δ^{18} O for a single time step (t = 696). The species δ^{18} O for each grid-point is based upon the integrating the δ^{18} O_{eq} values using a growth-rate based weighting (FAME; Roche et al., 2018). Values are in per mil (‰ V-PDB).

Figure 4. Anderson-Darling Results for Input datasets of Temperature and Equilibrium $\delta^{18}O(\delta^{18}O_{eq})$. Results of the test in which input variables underwent the same statistical procedure (see section 2.0) as the modelled data for (A-C) temperature and (D-F)

- 5 $\delta^{18}O_{eq}$ values. Here, model input data was extracted for three fixed depths ([A & D] 5 m; [B & E] 149 m; [C & F] and 235 m) without any growth weighting applied. Black regions are those grid points in which the null hypothesis (H₀), that the El Niño and Non- El Niño populations are not statistically different (FP_{El Niño} = FP_{Non-El Niño}), cannot be rejected. Gray regions represent grid points where the H₁ hypothesis is accepted, therefore the distributions of the foraminiferal population (FP) for El Niño and Non-El Niño can be said to be unique (FP_{El Niño} \neq FP_{Non-El Niño}). The hatched regions represent areas were the H₁ hypothesis can be
- 10 accepted, therefore the distributions of the foraminiferal population (FP) for El Niño and Non- El Niño can be said to be unique (FP_{El Niño} ≠ FP_{Non-El Niño}), though the difference between the means of tested distribution are less than (A-C) 0.5°C or (D-F) 0.1 ‰. Each panel represents a single depth (5, 149 and 235 m).

Figure 5. Anderson-Darling Results for modelled FAME- $\delta^{18}O_{eq}$: Panels representing locations of where dissimilar and similar values of FAME modelled species $\delta^{18}O$ occur between climate states, for (columns) particular species and (rows) particular model

- 5 depth cut-off limits. Each panel represents the Anderson-Darling test result, which are plotted with ([A] *Globigerinoides sacculifer* and [B] *Globigerinoides ruber*) and without ([C] *N. dutertrei*) model derived error. For all panels black areas reflect latitudinal and longitudinal grid points that failed to reject the null hypothesis (H₀) and therefore the foraminiferal population (FP) of the El Niño is similar to the Non-El Niño (FP_{El Niño} = FP_{Non-El Niño}). The results in which the H₁ hypothesis is accepted, in which the, therefore the distributions can be said to be different (FP_{El Niño} \neq FP_{Non-El Niño}), are plotted as either: (A – G. sacculifer, B – G. ruber) grey and
- 10 hatched or (C N. dutertrei) solely as white regions. For species with calculatable error, grey regions represent values where the difference between the two means of the population is greater than species-specific standard deviation of the FAME model and hatched regions represent those in which the means are less than this standard deviation (Roche et al., 2018). For species without a calculatable error, the regions are plotted in white. The rows represent the model runs with a depth cut-off limit at: (A-C) 60 m; (D) 100 m; (E) 200 m; and (F) 400 m.

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Figure 6. Anderson-Darling Results for modelled FAME-T_c: Panels representing locations of where dissimilar and similar values

- 5 of FAME modelled temperature recorded in the calcite shells (Tc) occur between climate states, for (columns) particular species and (rows) particular model depth cut-off limits. Each panel represents the Anderson-Darling test result, which are plotted with ([A] *Globigerinoides sacculifer* and [B] *Globigerinoides ruber*) and without ([C] *N. dutertrei*) model derived error. For all panels black areas reflect latitudinal and longitudinal grid points that failed to reject the null hypothesis (H₀) and therefore the foraminiferal population (FP) of the El Niño is similar to the Non-El Niño, and therefore the distribution between the neutral
- 10 climate and El Nino cannot be said to be different (FP_{El Niño} = FP_{Non-El Niño}). The results in which the H₁ hypothesis is accepted, in which the distributions can be said to be different (FP_{El Niño} ≠ FP_{Non-El Niño}), are plotted as white regions. The rows represent the model runs with a depth cut-off limit at: (A-C) 60 m; (D) 100 m; (E) 200 m; and (F) 400 m.

Figure 7. Combined A-D plots. As figure 5 and figure 6, in that panels represent locations of where dissimilar and similar values for the two climate states for (a-d) FAME- $\delta^{18}O_{eq}$ modelled oxygen isotope values or (e-h) FAME-T_c modelled temperature recorded in the calcite shells (Tc) occur. Each panel represents the Anderson-Darling test result, the results for *Globigerinoides*

- 5 sacculifer, Globigerinoides ruber and N. dutertrei are overlaid. For all panels black areas reflect latitudinal and longitudinal grid points that failed to reject the null hypothesis (H₀) and therefore the foraminiferal population (FP) of the El Niño is similar to the Non-El Niño, and therefore the distribution between the neutral climate and El Nino cannot be said to be different (FPEI Niño = FP_{Non-El Niño}). The results in which the H₁ hypothesis is accepted, in which the distributions can be said to be different (FPEI Niño ≠ FP_{Non-El Niño}), are plotted as yellow where the depth is deeper than 3500 m bsl or purple where the depth is shallower than 3500 m
- 10 bsl (see Figure 8). Purple locations are where our results suggest that the signal of ENSO has different values and the water depth allows for preservation – although this purple region will be smaller when inferred SAR is taken into account (see Figure 9). The rows represent the model runs with a depth cut-off limit at: (A and E) 60 m; (B and F) 100 m; (C and G) 200 m; and (D and H) where a combination of depths were utilised (Pracht et al., 2019).

Figure 8. Bathymetric map of the Tropical Pacific Ocean highlighting the areas above and below the Lysocline and/or Calcite compensation depth (CCD). (A) GEBCO map of height relative to 0 m; (B) same as (A) with location of seamounts plotted (white stars); (C-E) binary colour map of GEBCO data, yellow is values below cut-off depth value ([C] 3500 m below sea-level (bsl); [D]

5 4000 m bsl; and [E] 4500 m bsl respectively) and purple above the cut-off depth value. The histograms represent the normalised frequency of grid cell height in bins of 250 m wide, yellow is values below cut off value ([C] 3500 m below sea-level (bsl); [D] 4000 m bsl; and [E] 4500 m bsl respectively), purple above cut off value. The grey bins in each histogram are those above 0 m.

Figure 9. Map of the sedimentation rate and oxygen saturation for the Tropical Pacific. (A) Inferred sedimentation rate (Olsen et 2016). White regions represent continental shelf. (B) Oxygen saturation of the bottom grid layer of World Ocean Atlas 2013 (data from: https://www.nodc.noaa.gov/cgi-bin/OC5/woa13/woa13oxnu.pl). (C, E, G) Overlay between water depth and inferred SAR, Red / Pink: Continental shelf sediments that are (Red) shallower or (Pink) deeper than 3500 mbsl; Gray / White: grid point SAR is lower than SAR threshold and the seafloor depth is (grey) shallower or (white) deeper than 3500 mbsl; Light Yellow/Gold: Light yellow represents areas where the SAR is above the threshold but the water depth is deeper than 3500 mbsl in comparison Gold

10 represents areas where the SAR is above the threshold and the water depth is deeper than 3500 mbsl. The ideal locations are therefore plotted as Gold. Cut-off limits for SAR are (C) ≥1 cm kyr⁻¹; (E) ≥2 cm kyr⁻¹ and (G) ≥5 cm kyr⁻¹, (D, F, H) alongside the maps the bioturbation simulations for the minimum SAR threshold is plotted (see Figure 10 and Figure 11 for the output of SEAMUS). Each plot gives the input values of NGRIP (grey) and for each SAR three analysis were performed with different bioturbation depths (BD) these are (Blue) 5 cm; (Green) 10 cm; and (Orange) 15 cm.

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- 5 Figure 10. Output of the bioturbation model SEAMUS. (A) The unbioturbated input signal, NGRIP (North Greenland Ice Core Project Members, 2004; Rasmussen et al., 2014; Seierstad et al., 2014), used in our simulation of bioturbation for different SAR with SEAMUS (Lougheed, 2019). Sediment mixed layer referred to here as bioturbation depth (BD) is fixed at (B, E, H, K) 5 cm, (C, F, I, L) 10 cm and (D, G, J, M) 15 cm for sedimentation accumulation rates (SAR) of (B-D) 1 cm kyr⁻¹; (E-G) 2 cm kyr⁻¹; (H-J) 5 cm kyr⁻¹ and (K-M) 10 cm kyr⁻¹. The output is plotted as the discrete 1 cm depth median age. In (B-M) grey values represent the
- 10 unbioturbated input signal, NGRIP. Note, we retain the original units (V-SMOW) of the original timeseries used, no inference between Pacific climate and Greenland is intended by the use of NGRIP (see section 2.7).

5 Figure 11. Histograms of simulated specimen age within the bioturbation depth. The simulated age distribution present within the sediment mixed layer, referred to here as bioturbation depth (BD). BD is fixed at (A, D, G, J) 5 cm, (B, E, H, K) 10 cm and (C, F, I, L) 15 cm for sedimentation accumulation rates (SAR) of (A-C) 1 cm kyr⁻¹; (D-F) 2 cm kyr⁻¹; (G-I) 5 cm kyr⁻¹ and (J-L) 10 cm kyr⁻¹. The output is plotted as the discrete 1 cm depth median age. Note the size of the BD varies, therefore the simulated age distribution comes from a varying 'core depth'.

FAME genus	FAME species	Model iteration cut- off depth (in m)	Total Sta Closest Grid Cell Value		viation in Range of 3*3 grid	Interannu Closest Grid Cell Value	in ‰ Mean of	rd deviation Range of 3*3 grid		d Interant Mean of	veen Total nual Range of 3*3 grid	Data reference	Data	Data (age in kyr)	iCESM model Eulerian View (50 m); Standard deviation in ‰	iCESM model Eulerian View (100 m); Standard deviation in ‰	iCESM Lagrangian View; Standard deviation in ‰
	110	(0)	0.00	0.00	0.02	0.00	0.00	0.05	0.00	0.00	0.10				0.40	0.60	0.40
Globigerinodies Globigerinodies	sacculifer sacculifer	60 60	0.32 0.38	0.33 0.42	0.03 0.16	0.29 0.31	0.29 0.33	0.05 0.09	0.90 0.82	0.89 0.79	0.10 0.17				0.40 0.53	0.60 0.75	0.49 0.35
Globigerinodies	ruber	60	0.26	0.27	0.01	0.23	0.23	0.03	0.88	0.88	0.13				0.40	0.60	0.49
Globigerinodies	ruber	60	0.33	0.37	0.13	0.26	0.27	0.08	0.79	0.75	0.20	В	0.51	1.1 or 1.6	0.53	0.75	0.35
Neogloboquadrina	dutertrei	60	0.32	0.32	0.06	0.26 0.33	0.26	0.02	0.79	0.82	0.17	A C	0.38 0.28	1.5	0.40	0.60	0.49
Neogloboquadrina	dutertrei	60	0.41	0.45	0.18	0.33	0.35	0.11	0.81	0.78	0.16	C	0.28	1.6	0.53	0.75	0.35
Globigerinodies Globigerinodies	sacculifer sacculifer	100 100	0.25 0.33	0.26 0.36	0.03 0.14	0.22 0.28	0.22 0.29	0.04 0.08	0.88 0.84	0.87 0.81	0.12 0.16				0.40 0.53	0.60 0.75	0.49 0.35
Globigerinodies	ruber	100	0.20	0.21	0.07	0.16	0.16	0.02	0.80	0.80	0.23				0.40	0.60	0.49
Globigerinodies	ruber	100	0.27	0.31	0.11	0.22	0.22	0.08	0.79	0.73	0.23	В	0.51	1.1 or 1.6	0.53	0.75	0.35
Neogloboquadrina	dutertrei	100	0.29	0.29	0.05	0.23	0.23	0.02	0.79	0.82	0.17	A C	0.38 0.28	1.5	0.40 0.53	0.60 0.75	0.49 0.35
Neogloboquadrina	dutertrei	100	0.40	0.43	0.15	0.33	0.34	0.09	0.83	0.81	0.11	C	0.28	1.6	0.55	0.75	0.55
Globigerinodies Globigerinodies	sacculifer sacculifer	200 200	0.21 0.28	0.22 0.31	0.04 0.11	0.17 0.23	0.18 0.25	0.02 0.09	0.83 0.83	0.81 0.78	0.20 0.21				0.40 0.53	0.60 0.75	0.49 0.35
C																	
Globigerinodies Globigerinodies	ruber ruber	200 200	0.20 0.25	0.20 0.29	0.09 0.10	0.16 0.18	0.16 0.20	0.03 0.07	0.78 0.74	0.79 0.69	0.24 0.27	В	0.51	1.1 or 1.6	0.40 0.53	0.60 0.75	0.49 0.35
Neogloboquadrina Neogloboquadrina	dutertrei dutertrei	200 200	0.25 0.35	0.25 0.37	0.05 0.11	0.20 0.30	0.20 0.31	0.02 0.08	0.78 0.85	0.81 0.83	0.16 0.10	A C	0.38 0.28	1.5 1.6	0.40 0.53	0.60 0.75	0.49 0.35
Globigerinodies	sacculifer	400	0.21	0.22	0.04	0.17	0.18	0.02	0.83	0.81	0.20				0.40	0.60	0.49
Globigerinodies	sacculifer	400	0.28	0.31	0.11	0.23	0.24	0.09	0.83	0.78	0.21				0.53	0.75	0.35
Globigerinodies	ruber	400 400	0.20 0.25	0.20 0.29	0.09 0.10	0.16 0.18	0.16 0.20	0.03 0.07	0.78 0.74	0.79 0.69	0.24 0.27	В	0.51	110116	0.40 0.53	0.60 0.75	0.49 0.35
Globigerinodies	ruber	+00	0.25	0.29	0.10	0.18	0.20	0.07	0.74	0.09	0.27	В	0.31	1.1 or 1.6	0.55	0.75	0.55
Neogloboquadrina Neogloboquadrina	dutertrei dutertrei	400 400	0.24 0.34	0.23 0.36	0.05 0.11	0.18 0.29	0.19 0.30	0.02 0.07	0.77 0.85	0.80 0.83	0.17 0.10	A C	0.38 0.28	1.5 1.6	0.40 0.53	0.60 0.75	0.49 0.35

Data from: A Leduc et al., 2009;

B Koutavas and Joanides, 2012;

C Sadekov et al., 2013

Model (iCESM) values from supplement of Zhu et al., 2017 (converted from variance)

Table 1. Data-model comparison.

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