



# Data-Model Comparisons of the Tropical Hydroclimate Response to the 8.2 ka Event with an Isotope Enabled Climate Model

Andrea L. Moore<sup>1</sup>, Alyssa R. Atwood<sup>1</sup>, and Raquel E. Pauly<sup>1</sup>

<sup>1</sup>Department of Earth, Ocean, and Atmospheric Science, Florida State University, Tallahassee, Florida, USA

**Correspondence:** Andrea L. Moore (andilee.moore@gmail.com) and Alyssa R. Atwood (aatwood@fsu.edu)

**Abstract.** The 8.2ka Event, a prominent climate anomaly that occurred approximately 8,200 years before present (8.2ka), has been the subject of extensive research due to its potential implications for understanding the characteristics and mechanisms of abrupt climate change events. We characterize the tropical hydroclimate response to the 8.2ka Event based on a multiproxy compilation of 61 tropical hydroclimate records and assess the consistency between the reconstructed hydroclimate changes and those simulated by a new isotope-enabled climate model simulation of the 8.2ka Event with iCESM. The timing and duration of the hydroclimate anomalies is calculated using two event detection methods, one of which uses a new changepoint detection algorithm to account for age uncertainty. When age uncertainties are explicitly accounted for, significant hydroclimate anomalies associated with the 8.2ka Event are detected in 30% of the records in the compilation, with a mean onset age of  $8.28 \pm 0.12$  ka ( $1\sigma$ ), mean termination age of  $8.11 \pm 0.09$  ka ( $1\sigma$ ), and mean duration of  $152 \pm 70$  years ( $1\sigma$ ; with a range of 50–289 years), comparing well with previous estimates, and lending support to a regionally-variable tropical hydroclimate response to the 8.2ka Event, with events that span decadal to multi-centennial timescales in the proxy record. Notably, the hydroclimate anomalies are not hemispherically uniform, but rather display rich regional structure. Anomalous conditions are characterized by pronounced isotopic enrichment across East Asia, South Asia, and the Arabian Peninsula. In the Americas, drying and isotopic enrichment occurred in southern Central America, contrasting with isotopic depletion in central/eastern Brazil. In contrast, no robust signatures of the 8.2ka Event were found over the Maritime Continent. Many of these regional patterns generally agree with the new set of iCESM simulations of the 8.2ka Event. In iCESM, the North Atlantic meltwater forcing leads to a broad southward shift in tropical rainfall, resulting in a generally drier Northern Hemisphere and wetter Southern Hemisphere, but with large regional variations in precipitation amount and the isotopic composition of precipitation. Over the oceans, the precipitation  $\delta^{18}\text{O}$  anomalies are generally consistent with the "amount effect", wherein areas characterized by drying have more isotopically enriched precipitation and areas of wetting have more isotopically depleted precipitation. However, the precipitation  $\delta^{18}\text{O}$  anomalies are more decoupled from changes in precipitation amount over land. iCESM captures many of the regional hydroclimate responses observed in the reconstructions, including the large-scale isotopic enrichment pattern in precipitation  $\delta^{18}\text{O}$  in South and East Asia and the Arabian Peninsula, drying and isotopic enrichment in precipitation  $\delta^{18}\text{O}$  in southern Central America, isotopic depletion in parts of northeastern South America, and a muted hydroclimate response in the Maritime Continent. Overall, this study provides new insights into the tropical hydroclimate response to the 8.2ka Event, emphasizing the importance of accounting for age uncertainty in the hydroclimate reconstructions and the value of using isotope-enabled model simulations for data-model intercomparison.



## 1 Introduction

The tropics play a fundamental role in Earth's climate variability, acting as a heat source that drives global weather patterns via complex atmospheric teleconnections. A key component of the tropical climate system is the Intertropical Convergence Zone (ITCZ). From a zonal mean perspective, the ITCZ is the ascending branch of the Hadley cell and is characterized by the convergence of low-level trade winds, ascent, and heavy precipitation near the equator. On a regional basis, ITCZs exist over the Atlantic and eastern Pacific Ocean, where strong sea surface temperature (SST) gradients drive convergence, ascent, and narrow, well-defined rainbands. Distinct processes govern the large-scale circulation and precipitation in other regions of the tropics like monsoon systems and the Indian Ocean. Throughout the tropics, rainfall patterns migrate on a seasonal basis, following the warmer hemisphere. The migrations are regionally variable, with the Atlantic and Pacific ITCZs migrating between 9°N and 2°N in boreal summer/fall and winter/spring, respectively, while rainfall over the Indian Ocean and adjacent land masses swings more dramatically between 20°N and 8°S (Schneider et al., 2014). These fluctuations drive distinct wet and dry seasons through many regions of the tropics, providing critical access to water for roughly 40% of Earth's population (Penny, 2021). As the tropics comprise some of the most densely populated areas on Earth, it is critical to understand how tropical precipitation patterns may change in the near future. However, there is currently no agreement across models on how the ITCZ and monsoons will change with continued greenhouse gas forcing (Biasutti et al., 2018; Geen et al., 2020), in part due to persistent biases in the representation of the tropical mean state in global climate models (Li & Xie, 2014). Therefore, improving our understanding of how tropical rainfall patterns respond to external forcing is a critical target in the climate modeling community.

Our ability to make robust predictions about the climate system is also limited by the relatively short instrumental record. Paleoclimate proxy records extend the observational record beyond the instrumental era, helping to illuminate the linkages between climate forcings and the response of the climate system to forcing. Such data are critical for ground-truthing climate models to observations outside of the short period of direct observational data.

Past periods of abrupt climate change are especially important to examine in the context of evaluating future climate change risk, as we have no modern analogue with which to compare these events and we cannot rule out the possibility of such abrupt events in the future. Evidence from paleoclimate records (Arbuszewski et al., 2013; Koutavas & Lynch-Stieglitz, 2004; Rhodes et al., 2015) and model simulations of past climates (Chiang & Bitz, 2005; Roberts and Hopcroft, 2020) suggest that the location of the tropical rain bands may have shifted significantly and abruptly in the past (upwards of 7° latitude in certain regions) associated with changes in ice sheet extent and meltwater forcing (e.g., during Heinrich Events). The most recent such period of rapid, global climate reorganization occurred approximately 8,200 years before present day (the 8.2ka Event; Alley et al., 1997) and is thought to have lasted over a period of 100-200 years based on oxygen isotopic data from Greenland ice cores and tropical speleothems (Morrill et al., 2013). This period occurred during the otherwise stable Holocene epoch (11,700 years ago to present) and was driven by the discharge of around 163,000 km<sup>3</sup> of meltwater from proglacial Lakes Ojibway and Agassiz (remnants of the Laurentide Ice Sheet) into the North Atlantic, which triggered a large-scale salinity anomaly and resultant reduction in the strength of the Atlantic Meridional Overturning Circulation (AMOC; e.g., Barber et al., 1999; Ellison



et al., 2006). Proxy data and dynamical theory (e.g., Kang et al., 2008; Kang et al., 2009; Schneider et al., 2014) link this event to widespread cooling of the Northern Hemisphere (1–6°C; e.g., Ellison et al., 2006; Kobashi et al., 2007) and an associated southward shift of tropical rainfall patterns, with hydroclimate anomalies lasting anywhere from decades to centuries (e.g., Rohling and Palike, 2005; Morrill et al., 2013).

Morrill et al. (2013) published the most recent multiproxy compilation of high-resolution paleoclimate data related to the 8.2ka Event, incorporating 262 paleoclimate records from 114 global sites. Their synthesis demonstrated a regionally-variable hydroclimate response to the 8.2ka Event. Some of the robust regional hydroclimate changes included drying in Greenland, the Mediterranean, the Maritime Continent (Ayliffe et al., 2013; Chawchai et al., 2021), and across Asia (Wang et al., 2005; Dykoski et al., 2005; Cheng et al., 2009; Liu et al., 2013); while wetter conditions prevailed over northern Europe, Madagascar (Voarintsoa et al., 2019), and northeastern South America (Aguiar et al., 2020). Together, these data provide evidence for an anti-phased hemispheric precipitation response, with a strengthening of the South American summer monsoon (SASM) and a weakening of the Asian (AM) and East Asian summer monsoons (EASM).

Building on this work, Parker and Harrison (2022) used a statistical technique called breakpoint analysis to identify the timing, duration, and magnitude of the 8.2ka Event in 73 high-resolution, globally-distributed speleothem  $\delta^{18}\text{O}$  records from the Speleothem Isotope Synthesis and Analysis database (SISALv2; Comas-Bru et al., 2020). They identified significant isotopic excursions near 8.2ka in over 70% of their records and determined a median duration of global hydroclimate anomalies of approximately 159 years. Parker and Harrison (2022) inferred several regionally-coherent tropical hydroclimate anomalies from their synthesis, based on broad patterns of isotopic depletion across South America and southern Africa and isotopic enrichment in Asia, from which they inferred a weakening of Northern Hemisphere monsoons, strengthening of Southern Hemisphere monsoons, and a mean southward shift of the ITCZ as the most plausible mechanism for transmitting the effects of the 8.2ka Event throughout the tropics.

There are several limitations to these studies which are addressed in the updated proxy compilation presented here. Chiefly, Morrill et al. (2013) rely upon an a priori event window in classifying the climate response to the 8.2ka Event, and do not take radiometric age uncertainty of the proxy records into account. While Parker and Harrison (2022) consider the effects of age uncertainties on their compilation, they did not propagate these uncertainties through their breakpoint analyses. Neither study includes comparisons with isotope-enabled climate model simulations, a critical tool for validating proxy record interpretations. Further, tropical records comprise less than half of each compilation and since the publication of those studies, many new records have been generated in data-sparse regions that are key to understanding the complexities of tropical precipitation variability. Finally, recent studies (e.g., Atwood et al., 2020) have demonstrated significant regional variability in the tropical precipitation response to a variety of forcings, including North Atlantic meltwater events, calling into question the usefulness of invoking a southward shift in the zonal mean ITCZ as the primary mechanism driving hydroclimate changes in response to the 8.2ka Event, as invoked in the reconstructions of Morrill et al. (2013) and Parker and Harrison (2022).

This study seeks to provide new insights into the tropical hydroclimate response to the 8.2ka Event, by compiling an updated set of hydroclimate-sensitive proxy records complete with age model uncertainty and integrating them with new statistical tools to quantitatively evaluate how tropical rainfall patterns responded to this period of abrupt global climate change. We further



assess how well the proxy reconstructions compare to a new isotope-enabled model simulation of the 8.2 ka Event. In doing so, we hope to improve our understanding of the tropical hydroclimate response to abrupt AMOC disruptions and provide a critical benchmark for climate models that are used in projections of future climate change.

## 100 2 Methods

### 2.1 Synthesis of Published Datasets

To assess the tropical hydroclimate response to the 8.2ka Event, we developed an updated compilation of published, high-resolution, continuous, well-dated proxy datasets, collating records that span the period 7ka-10ka, cover latitudes from 30°N to 30°S, and which are sensitive to some aspect of hydroclimate variability. Records were identified through an in-depth 105 literature review, searching public data repositories (e.g., the NOAA National Centers for Environmental Information and World Data Center PANGAEA databases), and incorporated from previous compilations (e.g., Morrill et al., 2013). All records were reformatted into the Linked Paleo Data framework (LiPD; McKay and Emile-Geay, 2016) to facilitate analyses of age uncertainty and quantitative event detection.

To constrain the timing and duration of the abrupt hydroclimate anomaly associated with the 8.2ka Event, the datasets in 110 this compilation were screened to meet the following criteria: (i) data resolution of 50 years or better over the period of 7ka-10ka; (ii) based on hydroclimate-sensitive proxy data interpreted by authors as reflecting precipitation amount or intensity, the isotopic compositions of environmental water (including precipitation, lake water, and seawater), effective moisture, lake level, fluvial discharge, or sea surface salinity (SSS); and (iii) contain at least three radiometric dates over the 7ka-10ka interval. Emphasis was placed on collecting water isotope-based records to enable more direct comparison with isotope-enabled climate 115 model simulations.

The compilation was organized into three categories based on the climate interpretation of the various proxy records: proxies which reflect the isotopic composition of precipitation ( $P_{iso}$ ), proxies which reflect effective moisture (EM; P-E), and proxies which reflect precipitation amount and/or intensity ( $P_{amt}$ ). This categorization scheme enables more robust interpretations of the proxy records and facilitates data-model comparison as our understanding of water isotopes and their manifestations in 120 paleoclimate archives continues to advance (Konecky et al., 2020).

### 2.2 Age Model Development

Published radiometric age data were used to develop an ensemble of age-depth models for each dataset using Bayesian methods. Where available (Table A1), we employed the age ensembles developed by the Past Global Changes (PAGES) Speleothem 125 Isotope Synthesis and Analysis (SISAL) working group, using version 2 of their database (Comas-Bru et al., 2020). For records for which these age ensembles were not available due to lack of inclusion in the SISALv2 database or comprising a lacustrine or marine sediment archive, we developed age-depth models using the geoChronR package in R version 4.2.1 (McKay et al., 2021). All radiometric dates were obtained from the original publications and screened for updated age data



where available. For records originating from the Northern Hemisphere tropics, radiometric dates were calibrated using the Northern Hemisphere calibration curve, IntCal20. Dates of records originating from the Southern Hemisphere tropics were 130 calibrated using the Southern Hemisphere calibration curve, SHCal20. For each record, 1000 age-depth model iterations were run to generate a Markov Chain Monte Carlo (MCMC) age ensemble, which produces median age values and quantile age ranges, facilitating the propagation of age-model uncertainties through subsequent analyses.

In order to reduce uncertainty arising from the differences in age modeling algorithms offered through geoChronR, we prioritized the use of BACON (Blaauw and Christen, 2011) across our records, including those in the SISALv2 database, 135 where available. If a BACON age ensemble was not constructed for a SISALv2 dataset, we employed the Bchron (Haslett and Parnell, 2008) or copRA (Breitenbach et al., 2012) ensembles instead.

### 2.3 Detection of the 8.2ka Event

Two event detection methods were used in this study, as detailed below. The start, end, and duration of the hydroclimate anomalies associated with the 8.2 ka Event were calculated for all records in which events of the same sign were detected in 140 both event detection methods. This was done to leverage the strengths of each detection method and provide a more robust reconstruction of the hydroclimate response to the 8.2ka Event.

#### 2.3.1 Modified Morrill Method

For each record's published time series, we apply a modified version of the event detection methods described in Morrill et al. (2013) as a control to compare against our actR results (hereafter referred to as MM). Using the period from 7.4ka to 7.9ka as 145 a reference period, we calculate the mean and the standard deviation over that interval. From there, we define the upper and lower bounds by the two-sigma level. We repeat this process for a second reference period from 8.5ka to 9.0ka. We take the final upper and lower bounds as the most extreme values between the two reference periods. Then we use the 7.9-8.5ka period as the 8.2ka Event detection window.

Over this period, any values which exceed the upper or lower bound are marked as the 8.2ka Event, with the timing of the 150 event defined by the ages of the proxy values that exceed those bounds. In order for an excursion to be considered part of the 8.2ka Event, the excursions must last at least 10 years. If multiple events are detected within the 7.9-8.5ka window, they are combined into a single event if there are no more than three data points or thirty years separating the different excursions. This modification is necessary to account for the varying sampling resolutions present within and between several of the records in our compilation. If multiple events of differing signs are detected within the 8.2ka Event window, the event with the largest 155 z-score is chosen as the representative hydroclimate response. The magnitude of the event is defined by the largest absolute value z-score within the event detection period.



### 2.3.2 actR Method

A second event detection method was used to account for age model uncertainties in the proxy records. Past studies (e.g., Morrill et al., 2013) of the 8.2ka Event employed statistical techniques to detect excursions in the proxy records using an a 160 priori assumption that the North Atlantic meltwater perturbation propagated globally at exactly 8.2ka and lasted no more than 200 years. To better constrain the timing, duration, and magnitude of the 8.2ka Event in this study, we employed an event detection algorithm based on the changepoint package in the newly developed Abrupt Change Toolkit in R (actR; McKay and Emile-Geay, 2022). This algorithm detects abrupt shifts in the mean of a time series based on a prescribed number of age model 165 ensembles (generated in geoChronR), the minimum length of a segment (in years) over which mean shifts in the time series are detected, and a user-defined changepoint detection method and weighting penalty function. A minimum segment length of 50 or 100 years was assigned for each record in the proxy compilation to minimize short-lived transitions in the noisy proxy records, with the assumption that the 8.2ka Event signal in each of the records lasts at least 50 years.

Detected changepoints were summarized over 10-year-long windows. The Pruned Exact Linear Time (PELT; Killick et al., 2012) changepoint detection method was chosen for its computational efficiency and dynamic programming approach to 170 accurately identify the location and number of changepoints in time series data. The Modified Bayesian Information Criterion (MBIC; Zhang and Siegmund, 2007) was chosen as the penalty weighting function to balance the goodness of fit of the model to the data with the complexity of the model and the number of changepoints. These methods effectively minimize the detection of spurious changepoints within each ensemble. Each time series ensemble was tested against a robust null 175 hypothesis using surrogate proxy data generated by an isospectral noise model. By construction, the surrogate data have the same power spectrum as the original data, but phase scrambling destroys any autocorrelation that was present in the original time series. If autocorrelation is detected in a segment of the original time series ensemble, it fails the null hypothesis test, and any changepoint detected within that segment is excluded from the result. This test helps to ensure that the detected 180 changepoints are statistically significant and not just the result of random variation. Both age and proxy data uncertainties are propagated through each ensemble, improving the robustness of the result. For each record, 1000 age model ensembles were generated and tested against 100 surrogate time series.

Two types of events were characterized based on the actR results. ‘Significant events’ are defined by the presence of two consecutive changepoints with  $p < 0.05$  over the 7.9-8.3ka window (“start” and “end”). If more than two consecutive changepoints exist over that window, the two with the lowest p-values and highest probability are used. The difference between “start” and “end” dates is used to calculate event duration, which we assume to be no greater than 300 years. The magnitude of “events” is 185 determined by the greatest absolute value z-score in each record’s median age ensemble time series between the actR-derived “start” and “end” dates, with interpretation based on the sign of the z-score corresponding to the interpretation direction of the original authors. ‘Tentative events’ are defined by the presence of two consecutive changepoints with  $p < 0.1$  over the 7.7-8.5ka window (“start” and “end”). Events lasting more than 300 years are removed from consideration. If more than two events are detected within that window, the event with the start date closer to 8.2ka is chosen as the final 8.2ka event.



## 190 2.4 iCESM Simulations

The National Center for Atmospheric Research's (NCAR) water isotope-enabled Community Earth System Model (iCESM1.2; Brady et al., 2019) is a state-of-the-art, fully-coupled GCM designed to simulate water isotopes across all stages of the global hydroclimate cycle. It employs the CAM5.3 atmospheric model, with a gridded resolution of 1.9° latitude x 2.5° longitude and 29 vertical levels. Land processes are modeled by CLM4, at the same nominal 2° resolution. CLM is coupled to a River 195 Transport Model which routes runoff from the land into oceans and/or marginal seas. Both the POP2 ocean model and the CICE sea ice model have a common grid size of 320 x 384 with a nominal 1° resolution near the equator and in the North Atlantic. While iCESM faithfully captures the broad quantitative and qualitative features of precipitation isotopes, it is known to have a global bias toward depleted precipitation  $\delta^{18}\text{O}$  (median bias of -2.5‰; Brady et al., 2019).

We performed a new 8.2ka Event meltwater-forced ("hosing") simulation and an early Holocene control simulation ("ctrl") 200 using iCESM1.2. iCESM enables explicit tracking of water isotopes throughout the global water cycle, facilitating quantitative comparisons between model output and water isotope-based proxy records. These simulations were based on the freshwater forcing scenario recommended by Otto-Bliesner et al. (2017) for their proposed Paleoclimate Modeling Intercomparison Project 4-Coupled Model Intercomparison Project 6 (PMIP4-CMIP6) 8.2ka simulation. The climate of the early Holocene is characterized by different orbital configurations, including a larger obliquity and slightly higher eccentricity than present day. 205 In addition, precession resulted in increased seasonality of insolation in the Northern Hemisphere, with greatest insolation receipt in boreal summer (Wu et al., 2018; Hu et al., 2019). These factors produced warmer Northern Hemisphere summers, especially in mid to high latitudes, which are thought to have promoted the retreat of the remnant Laurentide Ice Sheet (LIS) (Otto-Bliesner et al., 2017 and references therein).

To simulate early Holocene climate, the model was forced with prescribed greenhouse gas concentrations ( $\text{CH}_4 = 658.5$  210 ppb,  $\text{CO}_2 = 260.2$  ppm, and  $\text{N}_2\text{O} = 255$  ppb), orbital configurations (eccentricity = 0.019524°, obliquity = 24.2030°, and longitude of perihelion = 99.228°), and a reconstruction of the ice sheet extent (Peltier et al., 2015) representative of conditions at 9ka. A control simulation ("ctrl") was initialized from an earlier 400-year-long 9ka simulation and was run for 100 model 215 years using these parameters. The 8.2ka Event simulation ("hose") was branched from year 100 of this control run. Initially, a simulated 2.5Sv meltwater flux (meltwater  $\delta^{18}\text{O} = -30\text{\textperthousand}$ ; Zhu et al., 2017) was applied across the northern North Atlantic Ocean (50–70°N) for 1 year, followed by 0.13Sv flux for 99 years to approximate the abrupt drainage of Lakes Agassiz and Ojibway and eventual collapse of the LIS at Hudson Bay (Otto-Bliesner et al., 2017). Monthly surface air temperature, 220 precipitation amount, and precipitation  $\delta^{18}\text{O}$  variables were extracted from each simulation for analysis. To isolate the global response to the simulated 8.2ka Event in the model, yearly time series of temperature (°C), precipitation amount (mm/day) and precipitation amount-weighted precipitation  $\delta^{18}\text{O}$  (‰) were obtained. Anomalies for each variable were calculated by subtracting the final 50 years of the "ctrl" simulation from the final 50 years of the "hose" simulation.



### 3 Results

#### 3.1 Data Compilation

61 tropical hydroclimate proxy records have been compiled in this study, covering 17 IPCC-designated scientific regions (Fig. A1; Iturbide et al., 2020). Compared to the compilation in Morrill et al. (2013), there is substantial improvement in hydroclimate 225 proxy data coverage across the Caribbean, Central America, South America, South and East Asia, and the Maritime Continent.

The compilation comprises 42 speleothem records (69%), 14 lacustrine records (23%), and 5 marine records (8%). When categorized into the three hydroclimate interpretation groups, the compilation consists of 43  $P_{iso}$  records (70.5%), 11 EM records (18%), and 7  $P_{amt}$  records (11.5%; Fig. 1; Table 2). For the purpose of this study, records which fully meet all inclusion criteria are designated as Tier 1 records ( $n = 50$ , 82%), forming the basis for the data-model intercomparison. Records which 230 fail to meet either the minimum paleodata resolution or radiometric date requirements are classified as Tier 2 records and are included as supporting datasets ( $n = 10$ , 16%). One record (MWS1; Dutt et al., 2015) failed to meet both of these requirements, thus it is designated as a Tier 3 record, and has been excluded from further analysis.

#### 3.2 Timing, Magnitude, and Duration of the 8.2ka Event in the Proxy Compilation

The approximate start, end, and duration of hydroclimate anomalies associated with the 8.2 ka event were calculated for all 235 records in our compilation in which events of the same sign were detected in both our modified MM and actR event detection methods. This was done to provide a more robust reconstruction of the hydroclimate response to the 8.2ka Event than that which either method would achieve alone. This final set of records comprises 30 of the 61 records (49%) in our compilation. The remaining 31 records in our compilation displayed a lack of agreement in the sign or presence of an event and are thus excluded from further analysis.

240 Of the 30 records that exhibit agreement between the two detection methods, significant hydroclimate events were detected in 18 records (34% of all Tier 1 and 10% of all Tier 2 records), with the remaining 12 records showing no event in either detection method (14% of all Tier 1 records and 50% of all Tier 2 records). Globally, the detected hydroclimate anomalies had an average start age of 8.28ka, and average termination age of 8.13ka, and an average duration of 152 years. The longest events occurred in the foraminifera  $\delta^{18}\text{O}$  record from the Gulf of Mexico (LoDico et al., 2006; MD022550; Fig. A39; 289 years) and 245 the speleothem record from Chongqing, China (Yang et al., 2019; HF01; Fig. A23; 259 years). The Chinese lacustrine magnetic susceptibility record of Hillman et al. (2021; F14; Fig. A17) has the earliest event onset age of 8.49ka, with a termination at 8.34ka, for a total duration of 152 years, while the Chinese speleothem record of Dykoski et al. (2005; D4Dykoski; Fig. A15) has the latest event onset age at roughly 8.11ka, terminating near 8.04ka, for an event duration of 62 years. Because the event detection algorithm in actR can be compromised by highly variable sampling resolution, in records with highly variable 250 resolution, we used the MM method to determine the event onset, termination, and duration. This applies to only two records: the speleothem record from Dykoski et al. (2005; D4Dykoski, Fig. A15) and the speleothem record from Neff et al. (2001; H5; Fig. A21).



Drier and/or isotopically enriched events were detected in 13 of the 30 records in the final compilation, including six records from East Asia (e.g. D4Dykoski ( $+2.8\sigma$ ) and F14 ( $+5.8\sigma$ )). Similarly, drying/isotopic enrichment was seen in three speleothem records from the Arabian Peninsula, with the largest shift ( $+3.5\sigma$ ) detected in the record of Cheng et al. (2009; H14; Fig. A22) between 8.08-8.21ka. The two speleothem records of Chawchai et al., 2021 from Klang Cave, Thailand (TK07, Fig. A57; TK20, Fig. A58) showed similarly high levels of isotopic enrichment ( $+3.1\sigma$  and  $+2.5\sigma$ ) between approximately 8.16-8.30ka. Finally, an isotopic enrichment of  $+3.4\sigma$  between 8.05-8.19ka in the Costa Rican speleothem record of Lachniet et al. (2004; V1; Fig. A61) and a negative excursion in titanium content indicative of a drying event ( $-4.0\sigma$ ) in the Guatemalan lake sediment record of Duarte et al. (2021; Core5LI; Fig. A10) from 8.09-8.16ka suggests a hydroclimate response to the 8.2ka Event in southern Central America (south of the Yucatan Peninsula; Fig. 6).

Wetter and/or isotopically depleted events were detected in five of the 30 records in the final compilation. Namely, the Madagascar speleothem records of Voarintsoa et al. (2017; ANJB2; Fig. A4) and Duan et al. (2021; ABC1; Fig. A3) showed negative isotopic excursions of  $-3.0\sigma$  and  $-2.5\sigma$ , respectively, while the two Brazilian speleothem records from Lapa Grande Cave (Strikis et al., 2011; LG11; Fig. A32 and Padre Cave (Cheng et al., 2009; PAD07; Fig. A43) exhibited negative isotopic excursions of  $-2.9\sigma$  and  $-2.7\sigma$ , respectively (Table 5). In addition, a large isotopic depletion event ( $-3.8\sigma$ ) was detected in the foraminifera  $\delta^{18}\text{O}$  record from the Gulf of Mexico (LoDico et al., 2006; MD022550; Fig. A39).

We found no significant hydroclimate response in the remaining 12 records of our compilation, with both the MM and actR event detection methods in agreement that no event occurred. This category included three lake sediment records from the Yucatan Peninsula (Fig. 6; LC1 [Hodell et al., 1995; Fig. A31], Curtis6VII93 [Curtis et al., 1998; Fig. A13], LagoPuertoArturo [Wahl et al., 2014; Fig. A29]), two speleothem records from Southeast Asia/the Maritime Continent (Fig. 7; KMA [Berkelhammer et al., 2012; Fig. A27], SSC01 [Carolin et al., 2016; Fig. A53]), and two speleothem records from Brazil (Fig. 5; RN1 [Cruz et al., 2009; Fig. A49], TM6 [Ward et al., 2019; Fig. A59]).

### 3.3 Regional Coherency of the Reconstructed Hydroclimate Changes

There is substantial regional coherency in the spatial pattern of reconstructed hydroclimate anomalies (Fig. 2), though they do not conform to the anticipated hemispheric dipole pattern typically associated with the 8.2ka Event, i.e. a generally drier/isotopically enriched Northern Hemisphere and a wetter/isotopically depleted Southern Hemisphere. Under both the MM and actR event detection methods, prominent drying/enrichment occurred across East and Southeast Asia, as well as the Arabian Peninsula. These dry conditions are interspersed with areas of no change in parts of the Maritime Continent and eastern India/Tibetan Plateau. No robust signatures of the 8.2ka Event are observed over the Maritime Continent. Central and South America display more of a hemispheric dipole pattern, with dry/enrichment events occurring north of the equator in Costa Rica and Guatemala, contrasting with wet/depletion events south of the equator in central/eastern Brazil. However, there are also regions in northern and central Brazil that exhibit no hydroclimate response. The proxy records thus present a far more complex, regionally-specific hydroclimate response to the 8.2ka Event than a simple hemispheric dipole pattern.



## 285 3.4 Global Signature of the 8.2 ka Event in iCESM

We now compare these reconstructed hydroclimate patterns to those simulated by iCESM under 8.2ka meltwater forcing. The surface temperature response is characteristic of the “bipolar seesaw” pattern (i.e., a colder northern hemisphere and a warmer southern hemisphere that is most pronounced in the Atlantic Ocean), consistent with a reduction in northward heat transport by AMOC (Fig. A2). Anomalously cool surface temperatures, reaching as low as -20°C where freshwater forcing was applied, 290 stretch across the northern North Atlantic Ocean, down the western coasts of Europe and North Africa, and into the tropical Atlantic via the North Atlantic Subtropical Gyre. Surface air temperatures across the Southern Hemisphere show a positive anomaly of up to 3°C, with the largest warming occurring in the South Atlantic. Over the continents, surface air temperatures cool in all regions except localized parts of northern South America, West Africa, and the southernmost regions of South America and Australia.

295 Accompanying these temperature anomalies are notable anomalies in precipitation amount, precipitation  $\delta^{18}\text{O}$ , and effective moisture (Fig. 3). Precipitation decreases while effective moisture increases throughout much of the North Atlantic, with the responses most pronounced in the regions with greatest cooling. The increase in effective moisture in this region indicates that the evaporation reduction outpaces the precipitation reduction (Fig. 3c). In the tropics, the largest precipitation anomalies appear in the tropical Pacific and Atlantic basins, with a southward shift of the Pacific and Atlantic ITCZs occurring in response 300 to the freshwater forcing (Fig. 3b). These shifts are characterized by a weakening of the northern extent of the ITCZs and an enhancement of the southern extent. The most pronounced drying occurs over central America and the eastern tropical Pacific, including Costa Rica and Panama, while the largest wet anomalies occur across the southern tropical Pacific. A notable hemispheric dry/wet dipole pattern is also observed in the tropical Atlantic, extending over northeastern South America. This pattern is less pronounced but still present over the tropical Indian Ocean and Africa. In contrast, no such dipole occurs over 305 the western Pacific or Maritime Continent.

These temperature and precipitation anomalies project strongly onto the amount-weighted precipitation  $\delta^{18}\text{O}$  values (Fig. 3a). The greatest precipitation  $\delta^{18}\text{O}$  anomalies occur in the northern reaches of the North Atlantic Ocean, reaching up to -8‰ in association with the strong regional cooling of the North Atlantic, as well as the addition of highly depleted (-30‰) meltwater to the surface ocean of the “hosing” site, and subsequent evaporation and rainout. In the tropics, precipitation  $\delta^{18}\text{O}$  310 anomalies closely follow the changes in precipitation amount over the tropical Atlantic and central/eastern Pacific Oceans, with negative precipitation  $\delta^{18}\text{O}$  anomalies south of the equator and positive precipitation  $\delta^{18}\text{O}$  anomalies north of the equator. A pronounced dipole pattern is also evident over northern South America, where anomalously increased (decreased) rainfall amounts correspond to negative (positive) precipitation  $\delta^{18}\text{O}$  anomalies in the southeastern (northwestern) region of South America. In the Middle East, India, Tibetan Plateau, and parts of Southeast Asia, modest drying is accompanied by pronounced 315 positive precipitation  $\delta^{18}\text{O}$  anomalies. However, not all regions are well described by the amount effect. In the Caribbean and Central America, there is a positive relationship between precipitation  $\delta^{18}\text{O}$  and precipitation amount, characterized by strong drying and isotopic depletion (associated with the strong upwind cooling and meltwater addition in the North Atlantic). There



also appears to be no clear relationship between precipitation amount and precipitation  $\delta^{18}\text{O}$  anomalies over Africa, East Asia, the Western Pacific, and Maritime Continent.

### 320 3.5 Data-Model Comparisons

The proxy locations span a total of 15 IPCC scientific regions (Fig. A1). The regions with densest coverage of Tier 1 proxy data are Central America, northeastern South America, East Asia, and Southeast Asia/Maritime Continent. These four regions were thus targeted for the data-model comparisons. The proxy records within each region were compared to the model-simulated anomalous annual mean precipitation amount, amount-weighted precipitation  $\delta^{18}\text{O}$ , and effective moisture (P-E) to investigate 325 the data-model agreement in the four target regions.

In East Asia (Fig. 4; Table 3; Table 4), five speleothem records display an isotopic enrichment event that broadly corresponds to the simulated large-scale isotopic enrichment pattern in precipitation  $\delta^{18}\text{O}$  that stretches across South Asia and the Arabian Peninsula in iCESM (Fig. 4a,b). This broad enrichment pattern in the model corresponds well with the similarly broad isotopic enrichment pattern found in the reconstructions, spanning East Asia, the Arabian Peninsula and southern Thailand. In iCESM, 330 the Chinese speleothem records are located near the node of an east-west dipole pattern in precipitation  $\delta^{18}\text{O}$  in eastern China, which is part of a larger pattern of isotopic depletion in South Asia and isotopic enrichment in the subtropics and extratropics of the North Pacific. In addition to these speleothem records, there is a single lake sediment record in southern China that displays a notable drying event, indicative of reduced precipitation, however this record contrasts with the results from iCESM, which show no notable change in precipitation in this region (Fig. 4b,e).

335 Northeaster South America displays only moderate proxy-model agreement (Fig. 5). Two of the four speleothem records there contain large  $\delta^{18}\text{O}$  depletion events, corresponding with the large-scale isotopic depletion signal in precipitation  $\delta^{18}\text{O}$  in iCESM across northeastern South America. However, two other speleothem records in the region—one in the Nordeste region of Brazil and one in central Brazil—show no significant hydroclimate anomalies during the 8.2ka Event, in contrast with the results from iCESM.

340 Subject to the small sample size in Central America (Fig. 6), broad agreement is found between the simulated and reconstructed hydroclimate anomalies, with the dry event detected in the Guatemalan lake sediment record of Core5LI (Duarte et al., 2021) corresponding with the reduced precipitation simulated throughout central America in iCESM. In addition, a positive (enriched) precipitation  $\delta^{18}\text{O}$  event found in the Costa Rican speleothem record (V1) is not inconsistent with the simulated isotopic enrichment in precipitation  $\delta^{18}\text{O}$  in the southernmost extent of Central America in iCESM, forming part of the wide 345 enrichment signal that stretches across the eastern Pacific. However, like the East Asian records, this speleothem record sits at the nodal point of a pronounced east-west dipole pattern in precipitation  $\delta^{18}\text{O}$  in iCESM, characterized by isotopic enrichment in the eastern tropical Pacific and widespread isotopic depletion in the tropical North Atlantic that stretches into the Caribbean and Mesoamerica. Lastly, the lack of a detected event in three lake sediment records from the Yucatan Peninsula (LagoPuertoArturo, Curtis6VII93, LC1) also agrees with the simulated weak EM response in that region in iCESM.

350 Broad data-model agreement is also found in Southeast Asia and the Maritime Continent (Fig. 7), where one speleothem record in the Thailand peninsula contains a notable isotopic enrichment event, in agreement with the simulated large scale



enrichment signal in precipitation  $\delta^{18}\text{O}$  in iCESM in South Asia (Fig. 7a,d). Two other speleothem records in Sumatra and Borneo show no significant hydroclimate anomalies, in general agreement with the weak simulated precipitation  $\delta^{18}\text{O}$  anomalies in iCESM in this region, which reflect the weak response in precipitation  $\delta^{18}\text{O}$  throughout the western Pacific and Maritime

355 Continent (Fig. 3a).

## 4 Discussion

### 4.1 Comparison to Previous Hydroclimate Compilations

The spatial pattern of the hydroclimate responses found in this study broadly agrees with those presented in Morrill et al. (2013) and Parker and Harrison (2022). Both sets of reconstructions, as well as the present study, find large-scale drying across

360 East Asia and the Arabian Peninsula, as well as a robust wet/depletion signal in central eastern Brazil; in Morrill et al. (2013) and Parker and Harrison (2022) this signal coincides with a dry/enrichment event in northern South America in agreement with the simulated hydroclimate response in iCESM (Fig. 3). All three reconstructions also agree on drying/enrichment in southern Central America, while both the present study and Parker and Harrison (2022) find a dipole pattern of wet/isotopically depleted

conditions in the Caribbean/Gulf of Mexico and dry/isotopically enriched conditions in southern Central America (Fig. 3a-c).

365 There is a reasonable level of agreement among all three compilations regarding the timing and duration of the hydroclimate anomalies. The age ensembles produced in the current analysis yield a mean start age of  $8.28 \pm 0.12\text{ka}$  ( $1\sigma$ ) and a termination age of  $8.11 \pm 0.09\text{ka}$  ( $1\sigma$ ) for the tropical hydroclimate response to the event. This timing agrees within age uncertainty to the timing of the event established from northern Greenland ice core data (8.25ka to 8.09ka; Thomas et al., 2007). In a compilation of eight absolutely-dated speleothems from South America and East Asia, Cheng et al. (2009) calculated an event

370 onset at 8.21ka and termination at 8.08ka. Similarly, Parker and Harrison (2022) used a compilation of 275 absolutely-dated speleothems to calculate the start of the global event at  $8.22 \pm 0.01\text{ka}$  and its termination at  $8.06 \pm 0.01\text{ka}$ . Importantly, the present study is the first to adopt a more comprehensive approach to age uncertainty in the proxy records through the generation of age ensembles in our analysis. The larger uncertainty in the timing of the event identified in this study is likely due to the inclusion of the age ensembles, as well as the multiproxy nature of our dataset, which includes lower resolution lake and

375 marine sediment records, in addition to the higher resolution speleothem records.

The duration of the hydroclimate anomalies identified in this study is also consistent with previous estimates. Parker and Harrison (2022) estimate durations ranging from 159 to 166 years, while Cheng et al. (2009) report a duration of 150 years, both of which fall within the range estimated from layer-counted Greenland ice cores ( $160.5 \pm 5.5$  years; Thomas et al., 2007). In their multi-proxy dataset, Morrill et al. (2013) report a larger range in event duration, spanning from 40 to 270 years. In the

380 present compilation, the mean duration of the tropical hydroclimate anomalies is  $152 \pm 70$  years ( $1\sigma$ ), with a range of 50-289 years, lending support to the multi-decadal to multi-centennial range of timescales found in earlier studies. Importantly, this work is the first to explicitly account for age uncertainty through all phases of the event detection analysis.

The multiproxy nature of this compilation, the explicit accounting of age uncertainties, and a geographic focus on the tropics likely contributed to a lower percentage of records which show robust signals of the 8.2ka Event (30%) relative to previous



385 studies (e.g., 70% of the global speleothem records compiled by Parker and Harrison, 2022). Proxy datasets from the tropics are likely to record lower magnitude climate anomalies relative to those observed in the North Atlantic and Europe, which experienced more pronounced climatic impacts from their proximity to the meltwater forcing that caused the 8.2ka Event. Moreover, while certain records may exhibit significant changes between 7.9ka and 8.3ka, the lack of sufficient age constraints in some records preclude the identification of robust events that can be tied to the 8.2ka Event (as these records display wide  
390 ranging age ensembles that fail the null hypothesis test in *actR*).

#### 4.2 Simulation of the 8.2 ka Event in iCESM

Proxy data play a crucial role in reconstructing local climate, but such data are often sparsely distributed, particularly through the tropics and the Southern Hemisphere, prohibiting a complete global picture of the 8.2ka Event. Ocean-atmosphere coupled General Circulation Models (GCMs) provide an invaluable tool for testing the response of the climate system to various forcings, and provide an estimate of the climate response in regions with scarce proxy data. Moreover, climate model simulations  
395 that track water isotopes through the hydrologic cycle facilitate more direct comparisons between proxy and model data.

Two lower-resolution isotope-enabled GCM simulations have previously been conducted to investigate the 8.2ka Event. LeGrande and Schmidt (2008) used the Goddard Institute for Space Studies ModelE-R (GISS ModelE-R) to evaluate the response of global temperatures, precipitation amount, and precipitation  $\delta^{18}\text{O}$  values to a slowdown of the AMOC. GISS  
400 ModelE-R is a fully-coupled GCM from the IPCC AR4 era, featuring a  $4^\circ \times 5^\circ$  horizontal resolution atmosphere model coupled with an ocean model of the same resolution, comprising 20 and 13 vertical layers, respectively. LeGrande and Schmidt (2008) performed a 1,000-year preindustrial control simulation and a suite of twelve meltwater forced experiments, applying a range of forcings (1.25 Sv to 10 Sv) over the Hudson Bay for 0.25 to 2 years. They found that this range of meltwater forcings inhibited North Atlantic Deep Water (NADW) formation and reduced the strength of the AMOC for up to 180 years.

405 In agreement with the results from iCESM, LeGrande and Schmidt (2008) found large precipitation  $\delta^{18}\text{O}$  anomalies over the meltwater source area in the North Atlantic in the decade following the meltwater forcing, which they similarly attributed to the evaporation and rainout of the isotopically depleted meltwater in the region. They observed reasonable agreement between their simulations and proxy records of temperature and hydroclimate, with the simulations containing larger meltwater forcing exhibiting better agreement with the proxies (emphasizing the importance of considering an ensemble of simulations to find the  
410 best fit to proxy reconstructions). Regarding the tropical hydroclimate response, they identified bands of enriched (depleted) precipitation  $\delta^{18}\text{O}$  anomalies in the northern (southern) tropics as a result of a southward shift in tropical rainfall. Notable patterns of  $\delta^{18}\text{O}$  enrichment were identified in Northeastern Africa, through the Middle East, south Asia, and the Thailand peninsula, which they attributed to large-scale changes in the hydrologic cycle, including shifts in moisture source and moisture transport pathways.

415 In a more recent set of simulations, Aguiar et al. (2021) used the University of Victoria Earth System Climate Model version 2.9 (UVic ESCM2.9) with the addition of oxygen isotopes to test proxy-model agreement under a range of empirically derived freshwater forcing scenarios. UVic ESCM2.9 uses the Modular Ocean Model version 2, with a horizontal resolution of  $3.6^\circ \times 1.8^\circ$  and 19 vertical levels. The version of the UVic ESCM2.9 model used in this study possesses a simple two-



dimensional atmospheric energy moisture balance model, which limits its ability to accurately represent precipitation  $\delta^{18}\text{O}$  values. 420 Aguiar et al. (2021) compared the sea surface temperatures and seawater  $\delta^{18}\text{O}$  values from 28 simulations with 35 proxy records to place new constraints on the amount and rate of freshwater forcing in the North Atlantic. Their analysis revealed that a two-stage meltwater experiment with a background flux of 0.066Sv over 1,000 years (9-8ka), followed by an intensification to 0.19Sv over 130 years (8.31-8.18ka), best replicated the anomalies observed in the proxy records.

The iCESM simulation illustrates clear signatures of the global 8.2ka Event that, at the largest scales, are broadly consistent 425 with the GISS and UVic simulations described above, including the hemispheric dipole pattern in temperature and associated southward shift of the tropical rainbands. On regional scales, the tropical rainfall patterns display substantial regional heterogeneity, with a southward shift of the tropical ocean rain bands, drying in the major NH monsoon regions of South Asia and West Africa, and wetting in parts of the South American Summer Monsoon. Tropical precipitation  $\delta^{18}\text{O}$  values display strong signatures of the 8.2ka Event, including opposing patterns of precipitation  $\delta^{18}\text{O}$  values between northern South 430 America and northeast Brazil (e.g., Zhu et al., 2017) and large precipitation  $\delta^{18}\text{O}$  anomalies over the meltwater region (e.g., LeGrande and Schmidt, 2008; Bowen et al., 2019). Dry (wet) anomalies correspond with enriched (depleted) precipitation  $\delta^{18}\text{O}$  values in some tropical regions, implicating the “amount effect” as the driving force behind the isotopic signal, but a decoupling of precipitation amount and precipitation  $\delta^{18}\text{O}$  anomalies occurs over many tropical continental regions, indicating 435 that other processes such as changes in moisture source, moisture transport pathways, water recycling over land, and/or changes in precipitation seasonality, dominate the isotopic signal in those regions. The model simulations lend support to the proxy reconstructions in demonstrating that the tropical hydroclimate response to the 8.2ka Event cannot be described as a simple hemispheric dipole pattern, particularly over continental regions, and that the rich regional structure of the precipitation amount and precipitation  $\delta^{18}\text{O}$  responses must be considered in order to understand the full picture of the tropical hydroclimate response to this event.

#### 440 4.3 Data-Model Comparisons

Subject to the small sample sizes found in the regional data-model comparisons, the results suggest that iCESM captures many 445 of the regional hydroclimate responses observed in the reconstructions, including the large-scale isotopic enrichment pattern in precipitation  $\delta^{18}\text{O}$  in South and East Asia, the muted hydroclimate response in the Maritime Continent, the drying and isotopic enrichment in southern Central America, and the isotopic depletion in precipitation  $\delta^{18}\text{O}$  in parts of northeastern South America. While qualitative, these areas of agreement between the proxies and model demonstrate that the tropical hydroclimate response to North Atlantic meltwater forcing during the 8.2ka Event was not a simple hemispheric dipole pattern, but is instead characterized by rich regional structure.

However, while there is some qualitative agreement between many of the reconstructed and simulated regional hydroclimate anomalies during the 8.2ka Event, our data-model comparisons are subject to a variety of limitations. For starters, our regional 450 analyses are limited by small sample sizes and in some regions like East Asia, the point-to-point agreement between the proxy and model data is low even as the regional hydroclimate patterns offer more nuanced context. In addition, our data-model comparisons are necessarily qualitative as many of the proxy records in our compilation are carbonate  $\delta^{18}\text{O}$  records, which do



not solely reflect changes in precipitation  $\delta^{18}\text{O}$ . Rather, these archives incorporate a combination of the isotopic composition of groundwater (for speleothem  $\delta^{18}\text{O}$  records; Lachniet, 2009) or seawater (for marine  $\delta^{18}\text{O}$  records; Konecky et al., 2020) 455 as well as the environmental temperature, among other factors (LeGrande and Schmidt, 2009; Bowen et al., 2019; Konecky et al., 2019). Thus, future work should integrate proxy system models with water isotope-enabled climate model simulations to develop more quantitative data-model comparisons of the 8.2ka Event. In addition, quantitative metrics like the weighted Cohen's kappa statistic could be used to quantitatively compare the proxy reconstructions to the pseudoproxy data derived from climate models (Cohen, 1960; Cohen, 1968; Landis and Koch, 1977; DiNezio and Tierney, 2013).

460 However, even when attempting to bridge the gap between models and proxy data using proxy system models and quantitative metrics, robust comparisons remain challenging. Characterizing the point-to-point agreement between the observed and simulated climate anomalies fails to address the well-known hydroclimate biases that exist in GCMs, which arise from factors like coarse model resolution, idealized topography, and the unresolved physics of cloud formation and convection. Furthermore, proxy data often captures localized climate signals which may not be representative of regional conditions. In contrast, 465 model data is averaged over the area of a grid cell, which can be large in coarse-resolution models. This can lead to non-trivial biases, particularly in coastal regions and regions of complex topography. Ultimately, the incorporation of additional well-dated proxy records that resolve different aspects of hydroclimate, paired with proxy system models, and ensembles of water isotope enabled climate model simulations of the 8.2ka Event will notably improve our understanding of the characteristics and mechanisms of the tropical hydroclimate response to abrupt climate change events.

## 470 5 Conclusions

This study has investigated the tropical hydroclimate response to the 8.2ka Event in a new multi-proxy data compilation and state-of-the art isotope-enabled model simulation. Two event detection methods were used in this study. The first method relies on the original age model of each record and uses the 7.9-8.5ka period as the detection window. The second method implements a changepoint detection algorithm that explicitly accounts for age uncertainties in each proxy record. In order to leverage the 475 strengths of each method and provide a more robust reconstruction of the hydroclimate response to the 8.2ka Event, only records in which events were detected in both event detection methods were used to characterize the hydroclimate response to the 8.2 ka Event.

Robust hydroclimate anomalies were detected in 30% of the compilation across the 7.9-8.5ka interval, with z-scores ranging between  $+5.8\sigma$  and  $-4.0\sigma$  in East Asia and southern Central America, respectively. 12 records showed no evidence of a 480 hydroclimate anomaly associated with the 8.2ka Event. The remaining records had conflicting results based on the two event detection methods and were excluded from further analysis. Across the records, a mean onset age of  $8.28 \pm 0.12\text{ka}$  ( $1\sigma$ ), mean termination age of  $8.11 \pm 0.09\text{ka}$  ( $1\sigma$ ), and mean duration of  $152 \pm 70$  years ( $1\sigma$ ; with a range of 50-289 years) was found, comparing well with previous estimates, and lending support to a regionally-variable tropical hydroclimate response to the 8.2ka Event, with events that span decadal to multi-centennial timescales. Importantly, this work is the first to explicitly account for 485 age uncertainty through all phases of the event detection analysis.



Broad agreement is observed between the reconstructions and iCESM model simulations and the results demonstrate that the tropical hydroclimate response to the North Atlantic meltwater forcing was not a simple hemispherically uniform dipole pattern, but is better characterized by rich regional structure. Coherent regional hydroclimate changes identified in proxy records include pronounced isotopic enrichment across East Asia, South Asia, and the Arabian Peninsula. In the Americas, 490 drying and isotopic enrichment occurred in southern Central America, contrasting with isotopic depletion in central/eastern Brazil. In contrast, no signatures of the 8.2ka Event were found over the Maritime Continent.

The isotope-enabled model simulation with iCESM illustrates clear signatures of the global 8.2ka Event that are largely consistent with the proxy records. The characteristic north-south dipole pattern in surface temperature is accompanied by an associated southward shift of tropical rainfall. On regional scales, however, the tropical rainfall changes are highly variable. 495 Major features include a southward shift of the tropical ocean rain bands in the tropical Atlantic, Central and Eastern Pacific, and Indian Oceans (characterized by a weakening of the northern extent and enhancement of the southern extent of the rain-bands), as well as drying in Central America and northern South America and wetter conditions in northeastern Brazil. Modest drying also occurs in the Northern Hemisphere monsoon regions of South Asia and West Africa.

The isotopic composition of tropical precipitation also displays strong signatures of the 8.2ka Event. Over land, precipitation 500  $\delta^{18}\text{O}$  displays a pronounced dipole pattern in South America, with isotopic enrichment in northern South America and isotopic depletion in northeast Brazil. Large-scale isotopic depletion also occurs over the Arabian Peninsula and South Asia. Over the tropical oceans (namely the tropical Atlantic, Central and Eastern Pacific, and Indian Oceans), a pronounced north-south dipole pattern occurs in precipitation  $\delta^{18}\text{O}$ , with isotopic enrichment corresponding with drier conditions north of the equator and isotopic depletion corresponding with wetter conditions south of the equator. Precipitation amount and  $\delta^{18}\text{O}$  anomalies are 505 more muted in the Western Pacific, Maritime Continent, and Africa.

To quantify data-model agreement, the proxy records were compared to simulated precipitation  $\delta^{18}\text{O}$ , precipitation amount, and effective moisture (P-E) from co-located sites in four regions with the densest coverage of proxy data: Central America, northeastern South America, East Asia, and Southeast Asia. Subject to the small sample sizes found in the regional data-model comparisons, the results suggest that iCESM captures many of the regional hydroclimate responses observed in the 510 reconstructions, including the large-scale isotopic enrichment pattern in precipitation  $\delta^{18}\text{O}$  in South and East Asia and the Arabian Peninsula, the drying and isotopic enrichment in precipitation  $\delta^{18}\text{O}$  in southern Central America, the isotopic depletion in parts of northeastern South America, and the muted hydroclimate response in the Maritime Continent.

These results serve as a first step toward more quantitative data-model comparisons. Recommendations for future studies include adding more well-dated proxy records that resolve different aspects of hydroclimate during the 8.2ka Event, and 515 quantitatively comparing these records with ensembles of water isotope enabled climate model simulations of the 8.2ka Event paired with proxy system models. Future work should also investigate the mechanisms of the observed hydroclimate changes and their isotopic signatures to improve our understanding of the characteristics and mechanisms of the tropical hydroclimate response to abrupt climate change events.



520 *Author contributions.* ARA and ALM conceptualized the study, with ARA securing funding and providing supervision. ALM performed proxy data curation and led the proxy data analysis with contributions from RP. ALM performed the iCESM simulations and led the writing of the manuscript and the design of figures with contributions from ARA.

*Competing interests.* The authors declare that they have no conflict of interest.

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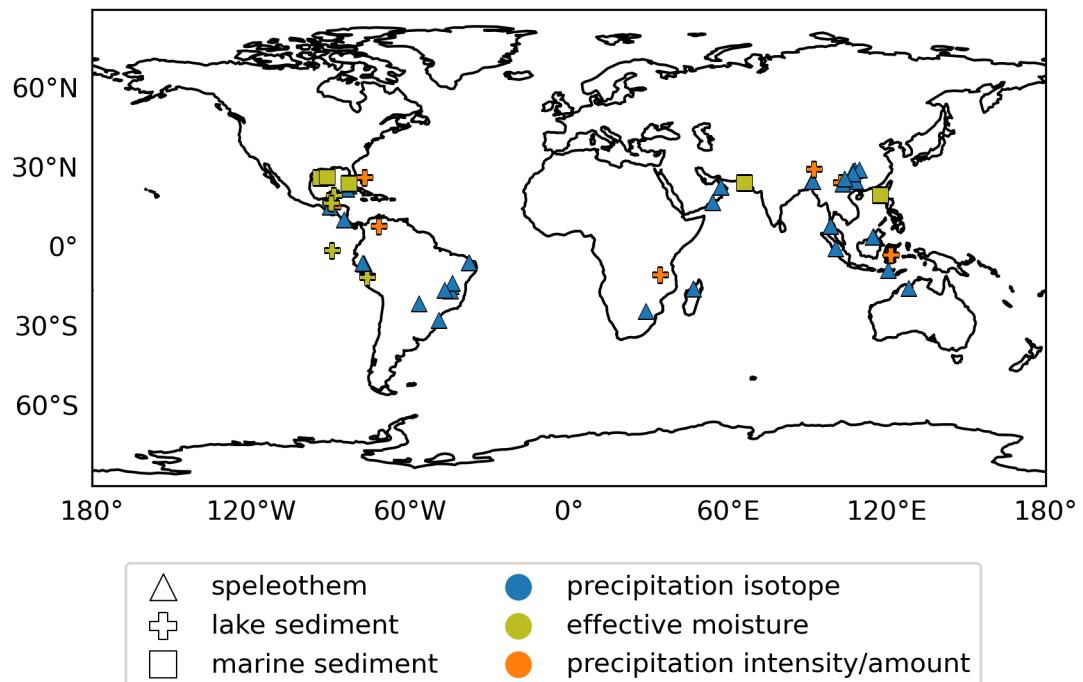
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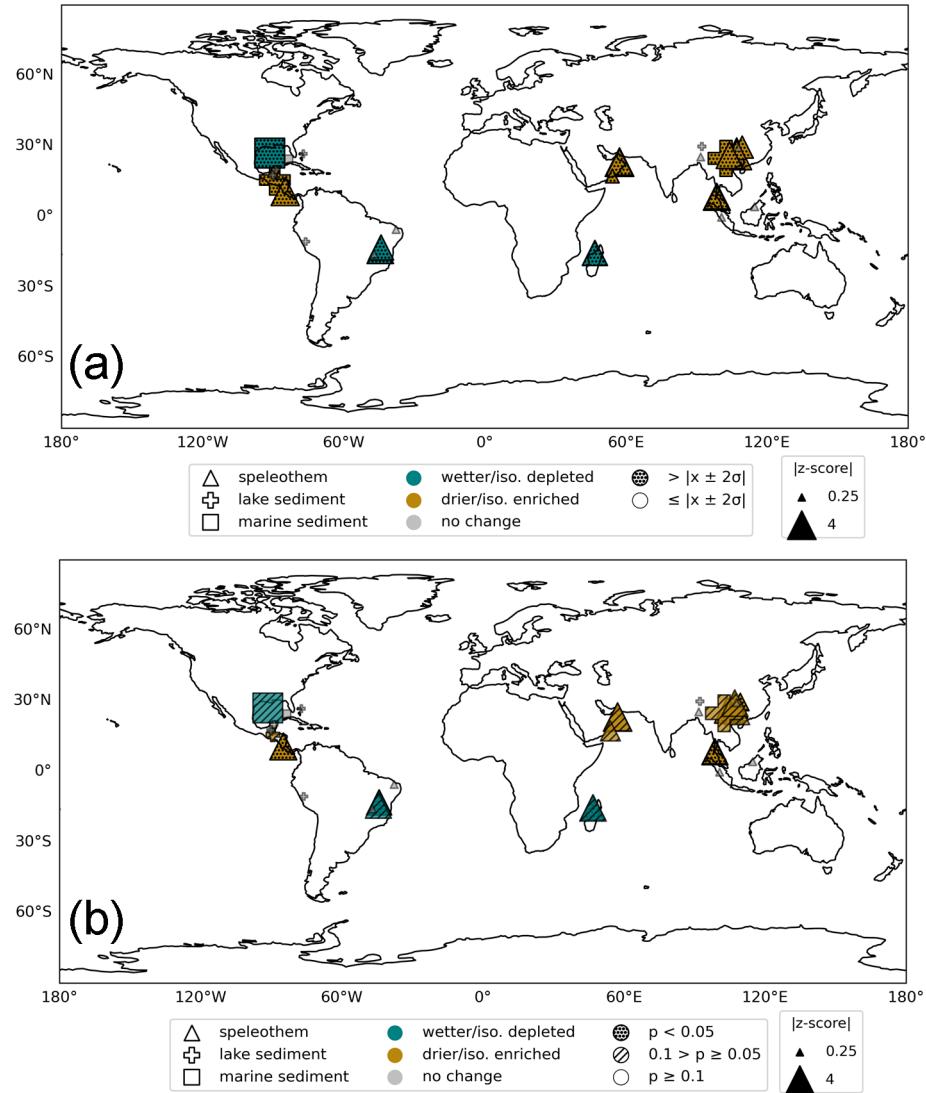
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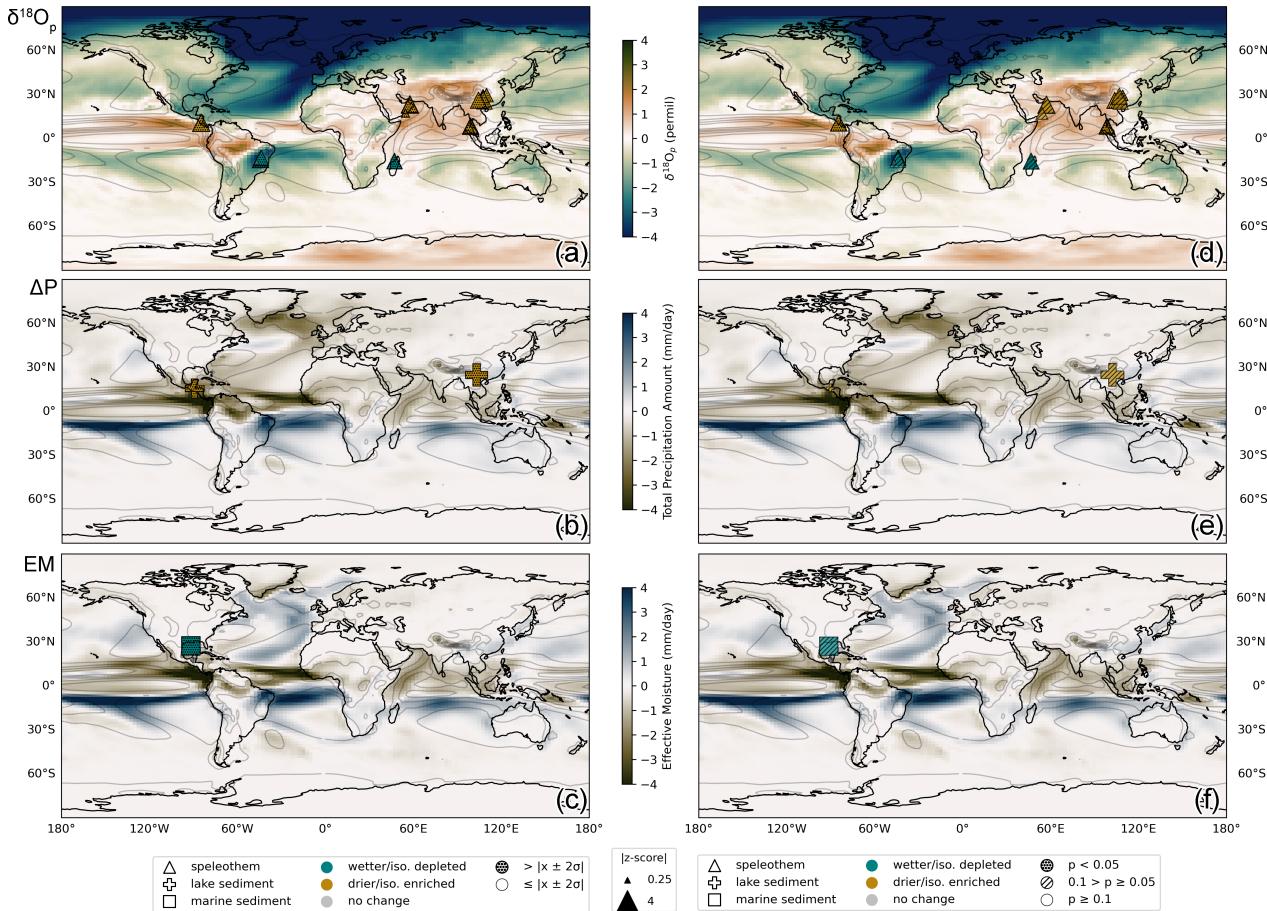
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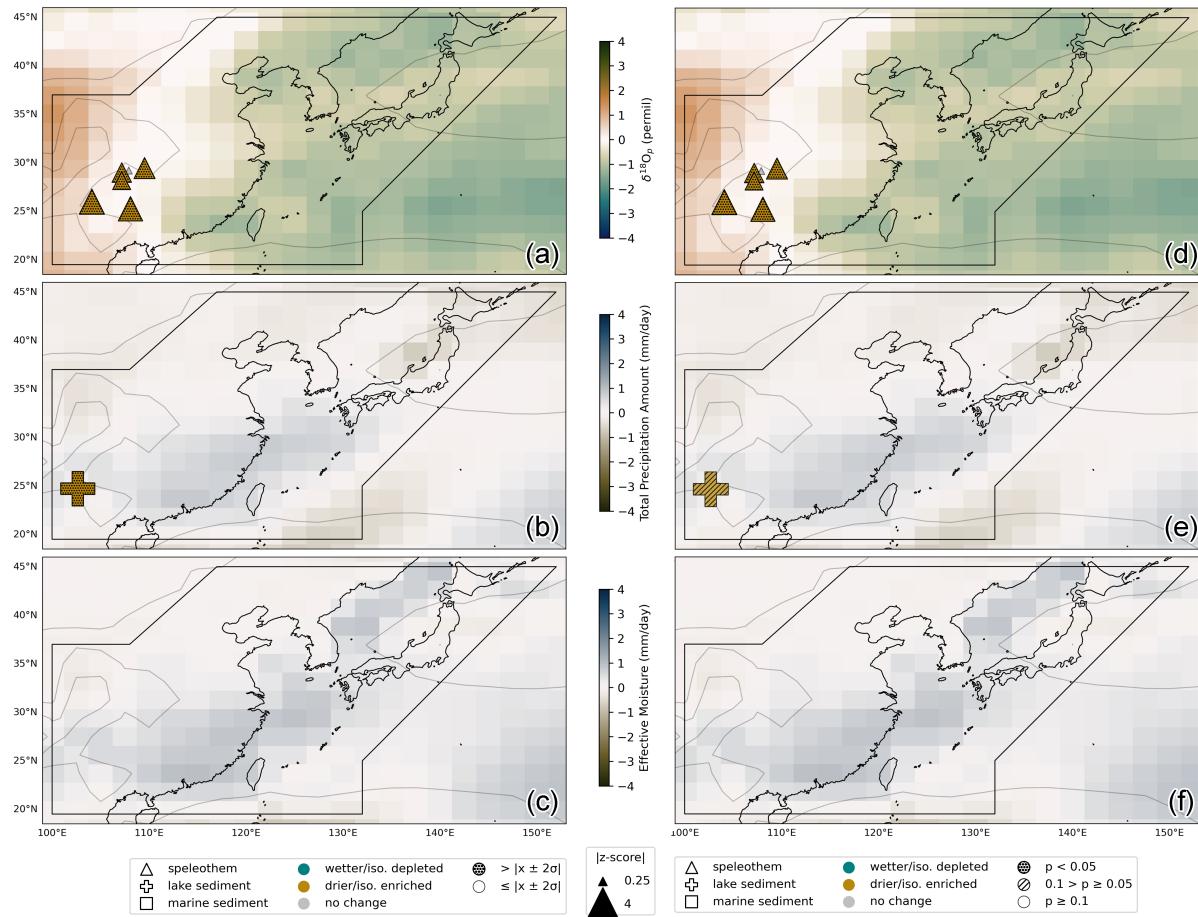
**Figure 1.** The distribution of proxy records comprising each interpretation group included in this study.



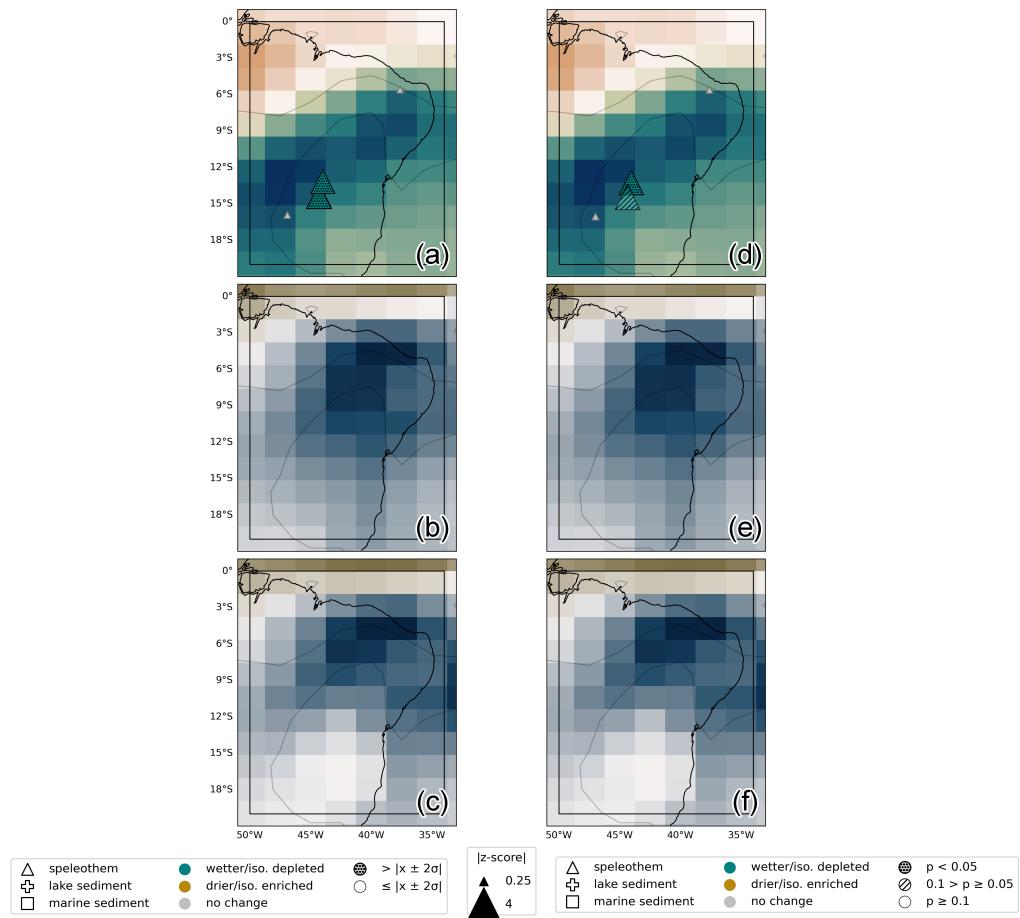
**Figure 2.** (a) Tropical hydroclimate anomalies detected in proxy archives over the 7.9-8.5 ka interval using the method described in Morrill et al. (2013). Blue symbols represent wetter (and/or isotopically depleted) conditions while brown symbols represent drier (and/or isotopically enriched) conditions relative to each record's mean climatology over the 7.4-7.9 ka and 8.5-9.0 ka windows described in the text. Grey symbols indicate the locations of proxy data where no significant 8.2 ka hydroclimate event was detected. Symbol size is scaled by  $100 \times |z\text{-score}|$ , calculated from the per-record mean and standard deviation over the 7 ka-10 ka interval. (b) Tropical hydroclimate anomalies detected in proxy archives over the 7.9-8.3 ka (“significant”) and 7.7-8.5 ka (“tentative”) intervals. Blue symbols represent wetter (and/or isotopically depleted) conditions while brown symbols represent drier (and/or isotopically enriched) conditions relative to each record's mean climatology over 7 ka-10 ka. Grey symbols indicate the locations of proxy data where no significant 8.2 ka hydroclimate event was detected. Symbol size is scaled by  $100 \times |z\text{-score}|$ , calculated from the per-record mean and standard deviation over the 7 ka-10 ka interval.



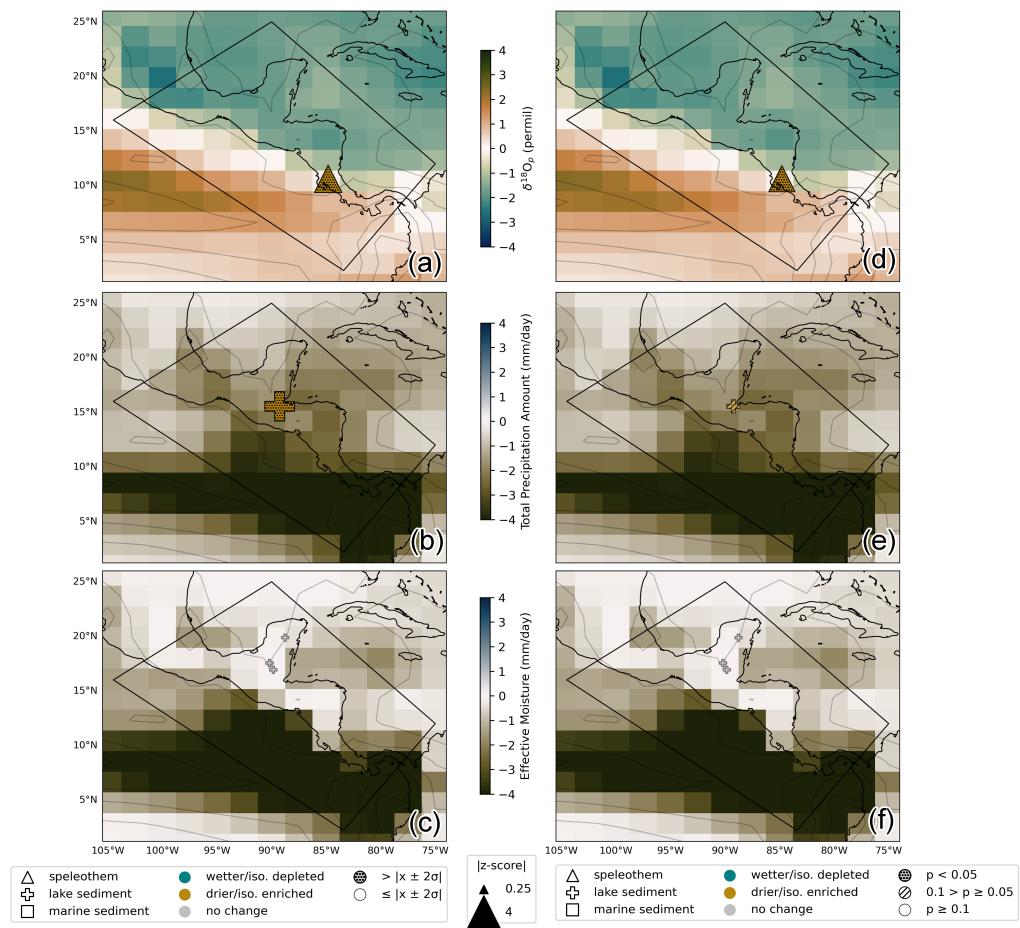
**Figure 3.** Summary of the 8.2 ka events detected using our modified Morrill et al. (2013) method (MM; left) and actR (right) for the paleoclimate records included in this study. Records with drier and/or isotopically enriched events are shown in brown, records with wetter and/or isotopically depleted events are shown in green, and records in which no event was detected are shown in grey. Stippling indicates that a “significant” event was detected in a given record by actR with event “start” and “end” times within the 7.9-8.3ka interval at the  $p < 0.05$  significance level. Slashed hatching indicates the presence of a “tentative” hydroclimate anomaly, with either a “significant” event detected outside of the 7.9-8.3ka window (between 7.7-8.5ka) or an event within that window where  $0.1 > p \geq 0.05$ . The archive type is indicated by the symbol (triangle = speleothem, plus = lake sediment, square = marine sediment). Symbol size is scaled by  $100 \times |z\text{-score}|$ , calculated from the per-record mean and standard deviation over the 7ka-10ka interval. Symbols are shown mapped over the simulated anomalous (a, d) amount-weighted oxygen isotopic composition of precipitation, (b, e) precipitation amount, (c, f) and effective moisture from the last 50 years of the iCESM “hose” and “ctrl” experiments.



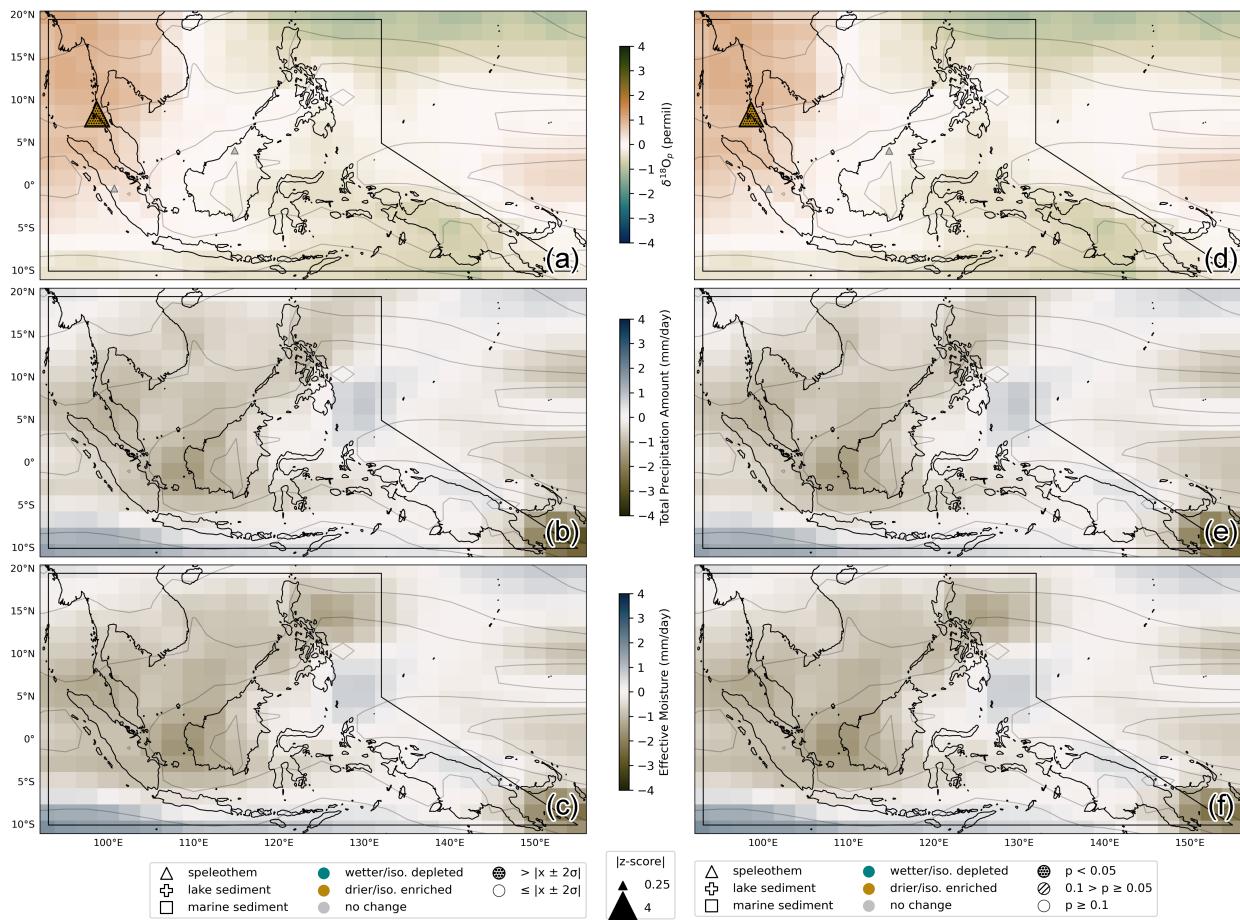
**Figure 4.** Data-model comparison of IPCC region 35: East Asia (box). Model shading represents (a) the precipitation amount-weighted  $\delta^{18}\text{O}$  anomaly, (b) the precipitation amount anomaly, and (c) the effective moisture (precipitation minus evaporation) between the last 50 years of the “hose” and “ctrl” simulations. Symbols represent paleoclimate proxy archives within the region corresponding to each respective climate variable, where the brown shaded triangles indicate speleothem records with recorded dry hydroclimate/enriched isotopic anomalies during the 8.2ka Event and grey symbols indicate records with no hydroclimate anomalies (“no change”) over the 8.3-7.9ka interval. For symbols showing an anomaly associated with the 8.2ka Event, size is scaled by  $200 \times |z\text{-score}|$  relative to each record’s mean and standard deviation.



**Figure 5.** As in Fig. 4, but for IPCC region 11: northeastern South America (box).



**Figure 6.** As in Fig. 4, but for IPCC region 7: southern Central America (box).



**Figure 7.** As in Fig. 4, but for IPCC region 38: Southeast Asia (box).



**Table 1.** Location metadata for all paleoclimate proxy datasets in this compilation.

Record ID	Lat	Lon	IPCC Region	Site Name	Reference	
ABC1	-15.54	46.89	Madagascar	Anjohibe Cave, Madagascar	Duan et al., 2021	
ANJB2	-15.54	46.89	Madagascar	Anjohibe Cave, Madagascar	Voarintsoa et al., 2017	
BA03	4.26	114.96	S.E. Asia	Malaysian Borneo	Chen et al., 2016	
BTV21a	-27.22	-49.16	S.E. South America	Botuverá Cave, SE Brazil	Bernal et al., 2016	
C7	26.57	-77.12	E. North America	Great Cistern Sinkhole, Bahamas	Sullivan et al., 2021	
CM2013	22.38	-83.97	Caribbean	Santo Tomas Cave, Cuba	Fensterer et al., 2013	
CM2019	23.38	-82.97	Caribbean	Santo Tomas Cave, Cuba	Warken et al., 2019	
Core17940	20.12	117.38	E. Asia	South China Sea	Wang et al., 1999	
Core5LI	15.53	-89.23	S. Central America	Lake Izabal, Guatemala	Duarte et al., 2021	
CP	22.38	-83.97	Caribbean	Dos Anas Cave, Cuba	Fensterer et al., 2013	
Curtis6VII93	16.92	-89.83	S. Central America	Lake Peten-Itza, Guatemala	Curtis et al., 1998	
D4Cheng	25.28	108.08	E. Asia	Dongge Cave, China	Cheng et al., 2009	
D4Dykoski	25.28	108.08	E. Asia	Dongge Cave, China	Dykoski et al., 2005	
EJConroy	-0.87	-89.45	Equatorial Pacific Ocean	El Junco Lake, Galapagos	Conroy et al., 2008	
F14	24.69	102.67	E. Asia	Dianchi, Yunan, China	Hillman et al., 2021	
FR5	29.23	107.9	E. Asia	Furong Cave, China	Li et al., 2011	
815	GB2GC1	26.67	-93.92	C. North America	Garrison Basin, Gulf of Mexico	Thirumalai et al., 2021
GURM1	15.43	-90.28	S. Central America	Grutas del Rey Marcos, Guatemala	Winter et al., 2020	
H14	23.08	57.35	Arabian Peninsula	Hoti Cave, Oman	Cheng et al., 2009	
H5	23.08	57.35	Arabian Peninsula	Hoti Cave, Oman	Neff et al., 2001	
HF01	29.02	107.18	E. Asia	Chongqing, Southwest China	Yang et al., 2019	
JAR7	-21.08	-56.58	S.E. South America	Jaragua Cave, Brazil	Novello et al., 2017	
JPC51	24.41	-83.22	Caribbean	Florida Straits	Schmidt et al., 2012	
KM1	25.26	91.88	S. Asia	Mawmluh Cave	Huguet et al., 2018	
KMA	25.26	91.88	S. Asia	Mawmluh cave	Berkelhammer et al., 2012	
KN51	-15.18	128.37	N. Australia	Cave KNI-51, Western Australia	Denniston et al., 2013 (a)	
LagoPuertoArturo	17.53	-90.18	S. Central America	Lago Puerto Arturo, Maya Lowlands	Wahl et al., 2014	
LBA99	8.33	-71.78	N. South America	Laguna Blanca, Venezuelan Andes	Polissar et al., 2013	
LC1	19.86	-88.76	S. Central America	Lake Chichancanab, Mexico	Hodell et al., 1995	
LG11	-14.42	-44.37	N.E. South America	Lapa Grande Cave, Brazil	Strikis et al., 2011	
LH2	29.48	109.53	E. Asia	Lianhua Cave, Hunan, China	Zhang et al., 2013	
LP	-10.7	-76.06	N.W. South America	Laguna Pumacocha, Peru	Bird et al., 2011	
LR06_B3_2013	-8.53	120.43	S.E. Asia	Liang Luar cave, western Flores, Indonesia	Ayliffe et al., 2013	
LSF19	-16.15	-44.6	N.E. South America	Lapa Sem Fim Cave, Brazil	Azevedo et al., 2021	



Record ID	Lat	Lon	IPCC Region	Site Name	Reference
M981P	-10.27	34.32	E. Southern Africa	Lake Malawi, Africa	Johnson et al., 2003
MAW6	25.26	91.82	S. Asia	Mawmluh Cave, India	Lechleitner et al., 2017
MD022550	26.95	-91.35	C. North America	Gulf of Mexico	LoDico et al., 2006
MWS1	25.26	91.88	S. Asia	Mawmluh cave	Dutt et al., 2015
NARC	-5.73	-77.5	N.W. South America	Cueva del Diamante, Peru	Cheng et al., 2013
NCB	-5.94	-77.31	N.W. South America	Cueva del Tigre Perdido, Peru	van Breukelen et al., 2008
PAD07	-13.22	-44.05	N.E. South America	Padre Cave, Brazil	Cheng et al., 2009
ParuCo	29.8	92.35	Tibetan Plateau	Paru Co, Tibetan Plateau, China	Bird et al., 2014
PET-PI6	17	-89.78	S. Central America	Lake Petén Itzá, Guatemala	Escobar et al., 2012
PLJJUN15	-11.04	-76.11	N.W. South America	Lake Junín, Peruvian Andes	Woods et al., 2020
Q52007	17.17	54.3	Arabian Peninsula	Qunf Cave, Oman	Fleitmann et al., 2007
Q5Cheng	17.17	54.3	Arabian Peninsula	Qunf Cave, Oman	Cheng et al., 2009
RN1	-5.58	-37.64	N.E. South America	Rainha cave, Brazil	Cruz et al., 2009
RN4	-5.58	-37.64	N.E. South America	Rainha cave, Brazil	Cruz et al., 2009
SG1	28.18	107.17	E. Asia	Shigao Cave, China	Jiang et al., 2012
Sha3	-5.7	-77.9	N.W. South America	Shatuca Cave, Peruvian Andes	Bustamante et al., 2016
SSC01	4.1	114.83	S.E. Asia	Gunung Mulu National Park, Borneo	Carolin et al., 2016
Staubwasser63KA	24.62	65.98	S. Asia	Arabian Sea	Staubwasser et al., 2003
T8	-24.02	29.11	E. Southern Africa	Makapansgat Valley, South Africa	Holmgren et al., 2003
TK07	8.33	98.73	S.E. Asia	Klang Cave, Thailand	Chawchai et al., 2021
TK20	8.33	98.73	S.E. Asia	Klang Cave, Thailand	Chawchai et al., 2021
TM6	-16	-47	N.E. South America	Tamboril Cave, Brazil	Ward et al., 2019
TOW109B	-2.73	121.52	S.E. Asia	Lake Towuti, Indonesia	Russell et al., 2014
V1	10.6	-84.8	S. Central America	Costa Rica	Lachniet et al., 2004
XBL29	24.2	103.36	E. Asia	Xiaobailong cave, China	Cai et al., 2015
ZLP1	26.02	104.1	E. Asia	Zhuliuping Cave, China	Huang et al., 2016



**Table 2.** Archive and interpretation metadata for the paleoclimate proxy datasets used in this study. Tier 1 data meet all strict inclusion criteria, while Tier 2 data are deficient in either dating or data resolution over the 7ka-10ka interval. Tier 3 data meet none of the strict inclusion criteria and are not included in quantitative analyses. All foraminifera used in the compilation are *G. ruber* (white). BSi MAR is the biogenic silica mass accumulation rate, in mg SiO<sup>2</sup>/cm<sup>2</sup>yr.

Record ID	Tier	Archive	Proxy	Interp. Group	Interp. Dir.	Reference
ABC1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Duan et al., 2021
ANJB2	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Voarintsoa et al., 2017
BA03	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Chen et al., 2016
BTV21a	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Bernal et al., 2016
C7	2	lacustrine	grain size	precip. amt.	direct	Sullivan et al., 2021
CM2013	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Fensterer et al., 2013
CM2019	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Warken et al., 2019
Core17940	1	marine	$\delta^{18}\text{O}$	eff. moisture	inverse	Wang et al., 1999
Core5LI	1	lacustrine	Ti	precip. amt.	direct	Duarte et al., 2021
CP	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Fensterer et al., 2013
Curtis6VII93	2	lacustrine	$\delta^{18}\text{O}$ gastropod ( <i>Cochliopina</i> sp.)	eff. moisture	inverse	Curtis et al., 1998
D4Cheng	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cheng et al., 2009
D4Dykoski	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Dykoski et al., 2005
EJConroy	1	lacustrine	clay (%)	eff. moisture	direct	Conroy et al., 2008
F14	2	lacustrine	magnetic susceptibility	precip. amt.	inverse	Hillman et al., 2021
FR5	2	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Li et al., 2011
GB2GC1	1	marine	$\delta^{18}\text{O}$	eff. moisture	inverse	Thirumalai et al., 2021
GURM1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Winter et al., 2020
H14	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cheng et al., 2009
H5	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Neff et al., 2001
HF01	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Yang et al., 2019
JAR7	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Novello et al., 2017
JPC51	1	marine	$\delta^{18}\text{O}$	eff. moisture	inverse	Schmidt et al., 2012
KM1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Huguet et al., 2018
KMA	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Berkelhammer et al., 2012
KN51	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Denniston et al., 2013 (a)
LagoPuertoArturo	1	lacustrine	$\delta^{18}\text{O}$	eff. moisture	inverse	Wahl et al., 2014
LBA99	1	lacustrine	magnetic susceptibility	precip. amt.	direct	Polissar et al., 2013
LC1	1	lacustrine	CaCO <sub>3</sub>	eff. moisture	direct	Hodell et al., 1995
LG11	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Strikis et al., 2011
LH2	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Zhang et al., 2013



Record ID	Tier	Archive	Proxy	Interp. Group	Interp. Dir.	Reference
LP	2	lacustrine	$\delta^{18}\text{O}$	precip. iso.	inverse	Bird et al., 2011
LR06B32013	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Ayliffe et al., 2013
LSF19	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Azevedo et al., 2021
M981P	2	lacustrine	BSi MAR	precip. amt.	direct	Johnson et al., 2003
MAW6	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Lechleitner et al., 2017
MD022550	1	marine	$\delta^{18}\text{O}$	eff. moisture	inverse	LoDico et al., 2006
MWS1	3	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Dutt et al., 2015
NARC	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cheng et al., 2013
NCB	2	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	van Breukelen et al., 2008
PAD07	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cheng et al., 2009
ParuCo	2	lacustrine	Lithics (%)	precip. amt.	direct	Bird et al., 2014
PETPI6	1	lacustrine	magnetic susceptibility	eff. moisture	direct	Escobar et al., 2012
PLJJUN15	1	lacustrine	Ti	eff. moisture	direct	Woods et al., 2020
Q52007	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Fleitmann et al., 2007
Q5Cheng	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cheng et al., 2009
RN1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cruz et al., 2009
RN4	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cruz et al., 2009
SG1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Jiang et al., 2012
Sha3	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Bustamante et al., 2016
SSC01	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Carolin et al., 2016
Staubwasser63KA	1	marine	foraminifera $\delta^{18}\text{O}$	eff. moisture	inverse	Staubwasser et al., 2003
T8	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	direct	Holmgren et al., 2003
TA122	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Wurtzel et al., 2018
TK07	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Chawchai et al., 2021
TK20	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Chawchai et al., 2021
TM6	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Ward et al., 2019
TOW109B	2	lacustrine	Ti (cps)	precip. amt.	direct	Russell et al., 2014
V1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Lachniet et al., 2004
XBL29	2	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Cai et al., 2015
ZLP1	1	speleothem	$\delta^{18}\text{O}$	precip. iso.	inverse	Huang et al., 2016



**Table 3.** Start, end, and duration of 8.2 ka Event calculated from changes in mean detected by actR and our modified Morrill et al. (2013) method for the global compilation and the four regions discussed in this study.

Region	Statistic	Event Start (yr BP)	Event End (yr BP)	Event Duration (yrs)
<i>n</i> = 18				
Global	Average	8282	8130	152
	Median	8283	8105	133
	Min	8106	8029	50
	Max	8489	8337	289
	SD	116	85	70
<i>n</i> = 6				
East Asia	Average	8284	8133	151
	Median	8306	8071	139
	Min	8106	8044	62
	Max	8489	8337	259
	SD	138	117	75
<i>n</i> = 2				
Southeast Asia	Average	8291	8176	116
	Median	8291	8176	116
	Min	8285	8155	101
	Max	8297	8196	130
	SD	8	29	21
<i>n</i> = 2				
Northeast South America	Average	8329	8204	125
	Median	8329	8204	125
	Min	8215	8165	50
	Max	8442	8242	200
	SD	161	54	106
<i>n</i> = 2				
South Central America	Average	8175	8069	106
	Median	8175	8069	106
	Min	8163	8051	77
	Max	8186	8086	135
	SD	16	25	41



**Table 4.** Regional and global summary of 8.2 ka events detected by actR and our modified Morrill et al. (2013) classification method, separated by the sign of the anomaly (“wetter”, “drier”, and “no change”).

IPCC Region	n	% of total	wetter	drier	no change	% of regional records w/ agreed "events"	"significant" actR events	"tentative" actR events	no actR events
Madagascar	2	3.3	2	0	0	100	1	1	0
S.E.Asia	7	11.5	0	2	1	42.9	2	1	4
S.E.South-America	2	3.3	0	0	0	0	0	2	0
E.North-America	1	1.6	0	0	1	100	0	0	1
Caribbean	4	6.6	0	0	1	25	0	3	1
E.Asia	10	16.4	0	6	1	70	4	4	2
S.Central-America	7	11.5	0	2	3	71.4	1	1	5
Equatorial.Pacific-Ocean	1	1.6	0	0	0	0	0	0	1
C.North-America	2	3.3	1	0	0	50	1	1	0
Arabian-Peninsula	4	6.6	0	3	0	75	1	3	0
S.Asia	5	8.2	0	0	1	20	0	4	1
N.Australia	1	1.6	0	0	0	0	1	0	0
N.South-America	1	1.6	0	0	0	0	1	0	0
N.E.South-America	6	9.8	2	0	2	66.7	1	2	3
N.W.South-America	5	8.2	0	0	1	20	1	2	2
E.Southern-Africa	2	3.3	0	0	0	0	2	0	0
Tibetan-Plateau	1	1.6	0	0	1	100	0	0	1
Global	61	100	5	13	12	-	16	24	21



**Table 5.** The timing, duration, magnitude, and interpretation of the 8.2ka Event for records with agreement between MM and actR methods.

IPCC Region	Record ID	Event Start (yr BP)	Event End (yr BP)	Event Duration (yrs)	MM z-score	actR z-score	Interpretation
Madagascar	ABC1	8248	8029	219	-2.5	-2.5	wetter
	ANJB2	8318	8124	194	-2.7	-3.0	wetter
E.Asia	D4Dykoski	8106	8044	62	2.8	2.8	drier
	F14	8489	8337	152	5.5	5.8	drier
	FR5	N/A	N/A	N/A	N/A	N/A	no change
	LH2	8158	8068	90	2.1	1.3	drier
	HF01	8332	8073	259	1.8	3.0	drier
825	ZLP1	8339	8213	126	3.0	2.9	drier
	SG1	8280	8062	218	1.5	2.9	drier
	H14	8208	8080	128	3.5	3.5	drier
	H5	8135	8042	93	2.9	3.2	drier
	Q52007	8407	8199	208	0.8	1.7	drier
Tibetan-Plateau	ParuCo	N/A	N/A	N/A	N/A	N/A	no change
S.Asia	KMA	N/A	N/A	N/A	N/A	N/A	no change
	SSC01	N/A	N/A	N/A	N/A	N/A	no change
S.E.Asia	TK07	8297	8196	101	3.1	2.9	drier
	TK20	8285	8155	130	2.5	2.5	drier
Caribbean	JPC51	N/A	N/A	N/A	N/A	N/A	no change
C.North-America	MD022550	8469	8180	289	-3.8	-3.8	wetter
E.North-America	C7	N/A	N/A	N/A	N/A	N/A	no change
	LC1	N/A	N/A	N/A	N/A	N/A	no change
	V1	8186	8051	135	3.4	3.1	drier
	Core5LI	8163	8086	77	-4.0	-0.8	drier
	Curtis6VII93	N/A	N/A	N/A	N/A	N/A	no change
N.W.South-America	LagoPuertoArturo	N/A	N/A	N/A	N/A	N/A	no change
	LP	N/A	N/A	N/A	N/A	N/A	no change
	PAD07	8215	8165	50	-2.7	-2.7	wetter
	LG11	8442	8242	200	-3.0	-2.9	wetter
	RN1	N/A	N/A	N/A	N/A	N/A	no change
N.E.South-America	TM6	N/A	N/A	N/A	N/A	N/A	no change



## Appendix A



**Table A1.** Age model information.

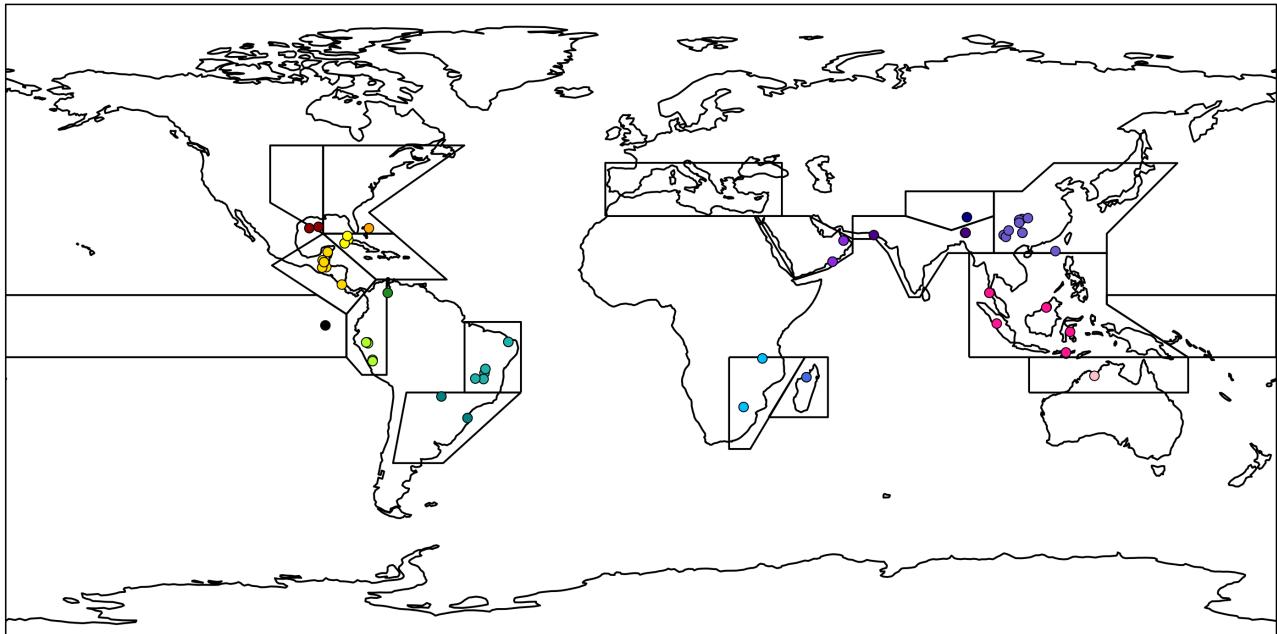
Record ID	Published Age Model Algorithm	Published 14C Cal. Curve	Contains Hiatus?	Contains Reversal?	In SISALv2?	Age Model Chosen
ABC1	MOD-AGE	N/A	N	Y	N	Bacon
ANJB2	StalAge	N/A	Y	Y	Y	SISAL Bacon
BA03	StalAge	N/A	N	N	Y	SISAL Bacon
BTV21a	unknown	N/A	N	N	Y	SISAL Bacon
C7	Bacon	IntCal13	N	N	N/A	Bacon
CM2013	StalAge	N/A	N	N	Y	SISAL copRa
CM2019	StalAge	N/A	N	N	Y	SISAL Bacon
Core17940	CALIB 3.0.3	unknown	N	Y	N/A	Bacon
Core5LI	Bacon	IntCal20	N	Y	N/A	Bacon
CP	StalAge	N/A	N	Y	Y	SISAL Bchron
Curtis6VII93	linear interpolation	unknown	N	N	N/A	Bacon
D4Cheng	unknown	N/A	N	N	Y	Bacon
D4Dykoski	linear interpolation	N/A	N	N	Y	Bacon
EJConroy	CALIB 5.0	unknown	N	N	N/A	Bacon
F14	Bacon	IntCal20	N	Y	N/A	Bacon
FR5	unknown	IntCal09	N	N	Y	SISAL copRa
GB2GC1	Bacon	Marine13	N	N	N/A	Bacon
GURM1	COPRA	N/A	N	N	N	SISAL Bacon
H14	unknown	N/A	N	N	Y	Bacon
H5	unknown	N/A	N	Y	Y	SISAL Bacon
HF01	polynomial fit	N/A	N	N	N	SISAL copRa
JAR7	linear interpolation	N/A	N	N	Y	SISAL Bacon
JPC51	CALIB 6.0	unknown	N	N	N/A	Bacon
KM1	StalAge	N/A	N	N	Y	SISAL Bchron
KMA	StalAge	N/A	N	N	Y	SISAL Bacon
KN51	unknown	N/A	N	Y	Y	SISAL copRa
LagoPuertoArturo	CLAM 2.2	IntCal13	N	N	N/A	Bacon
LBA99	linear interpolation	IntCal04	Y	N	N/A	Bacon
LC1	CALIB	unknown	N	Y	N/A	Bacon
LG11	unknown	N/A	N	N	Y	SISAL Bacon
LH2	linear interpolation	N/A	N	N	Y	SISAL Bacon
LP	CALIB 5.0	unknown	N	N	N/A	Bacon
LR06B32013	linear interpolation	N/A	N	N	Y	SISAL Bchron



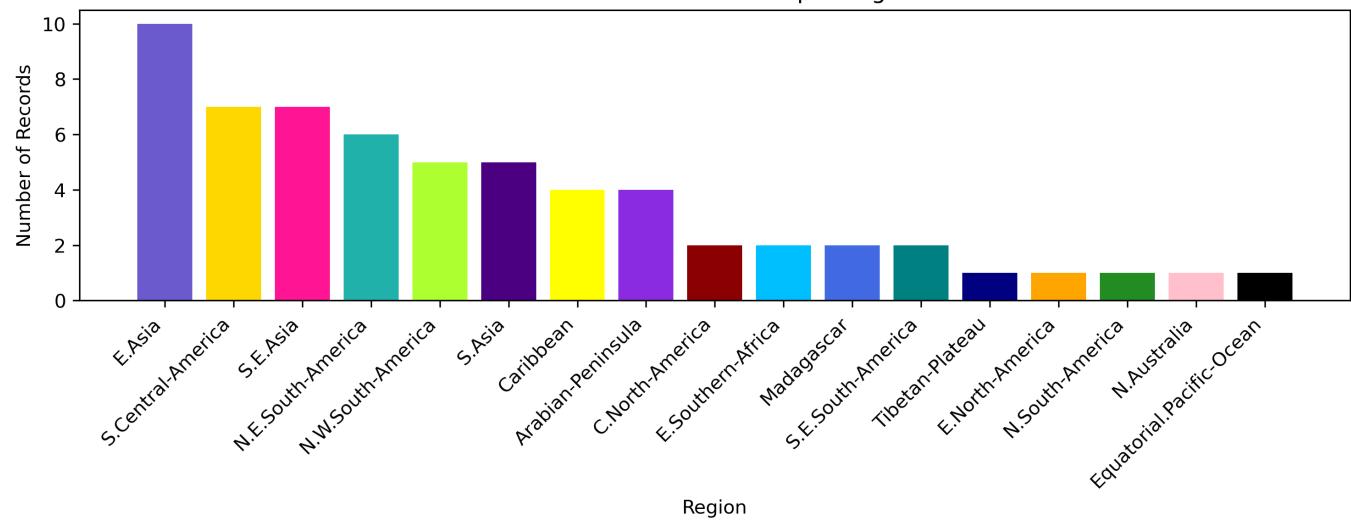
Record ID	Published Age Model Algorithm	Published 14C Cal. Curve	Contains Hiatus?	Contains Reversal?	In SISALv2?	Age Model Chosen
LSF19	unknown	N/A	Y	N	N	SISAL Bacon
M981P	CALIB 4.3	unknown	N	N	N/A	Bacon
MAW6	COPRA	N/A	N	Y	Y	SISAL Bchron
MD02_2550	CALIB 5.0	unknown	N	N	N/A	Bacon
NARC	linear interpolation	N/A	N	N	Y	SISAL copRa
NCB	Isoplot 3	N/A	N	N	Y	SISAL Bacon
PAD07	unknown	N/A	N	N	N	Bacon
ParuCo	CALIB 6.0	IntCal09	N	N	N/A	Bacon
PET-PI6	OxCal	IntCal09	N	N	N/A	Bacon
PLJ-JUN15	Bacon	IntCal13	N	N	N/A	Bacon
Q52007	linear interpolation	N/A	N	N	Y	Bacon
Q5Cheng	unknown	N/A	N	N	Y	Bacon
RN1	unknown	N/A	N	N	Y	SISAL Bchron
RN4	unknown	N/A	N	N	Y	SISAL Bchron
SG1	linear interpolation	N/A	N	N	Y	SISAL Bacon
Sha3	COPRA	N/A	N	N	Y	SISAL Bchron
SSC01	StalAge	N/A	N	N	Y	Bacon
Staubwasser63KA	least-squares	IntCal98	N	N	N/A	Bacon
T8	linear interpolation	N/A	N	N	Y	Bacon
TA122	Bacon	N/A	N	N	Y	SISAL copRa
TK07	Bacon	N/A	N	N	N	SISAL Bacon
TK20	Bacon	N/A	N	N	N	SISAL Bacon
TM6	COPRA	N/A	N	N	Y	SISAL Bacon
TOW109B	CALIB 6.0	unknown	N	N	N/A	Bacon
V1	fifth-order polynomial best-fit age model	N/A	N	Y	Y	SISAL Bacon
XBL29	linear interpolation	N/A	N	N	Y	SISAL Bchron
ZLP1	linear interpolation	N/A	N	N	Y	SISAL Bacon



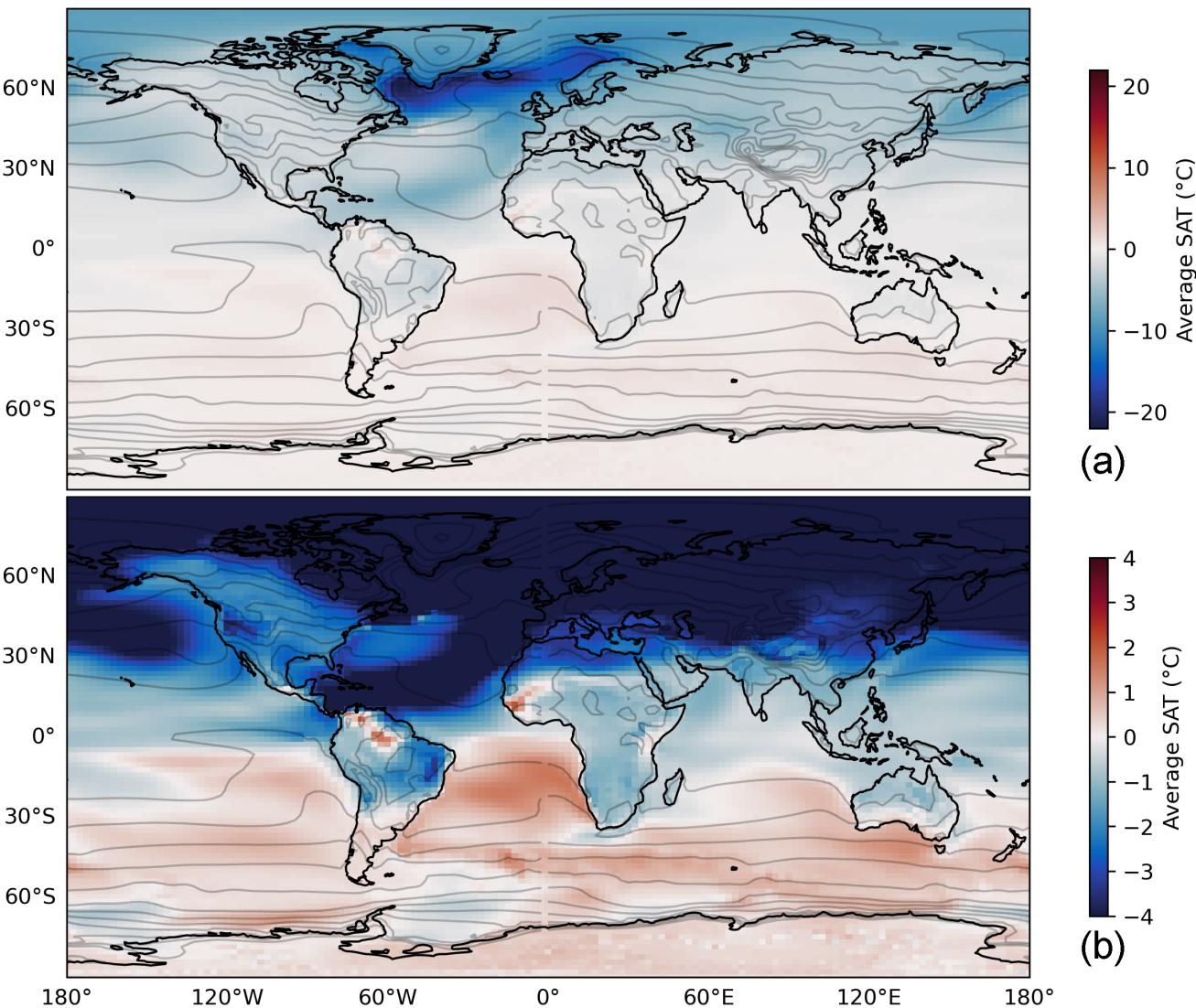
Global Map of Records within IPCC Regions



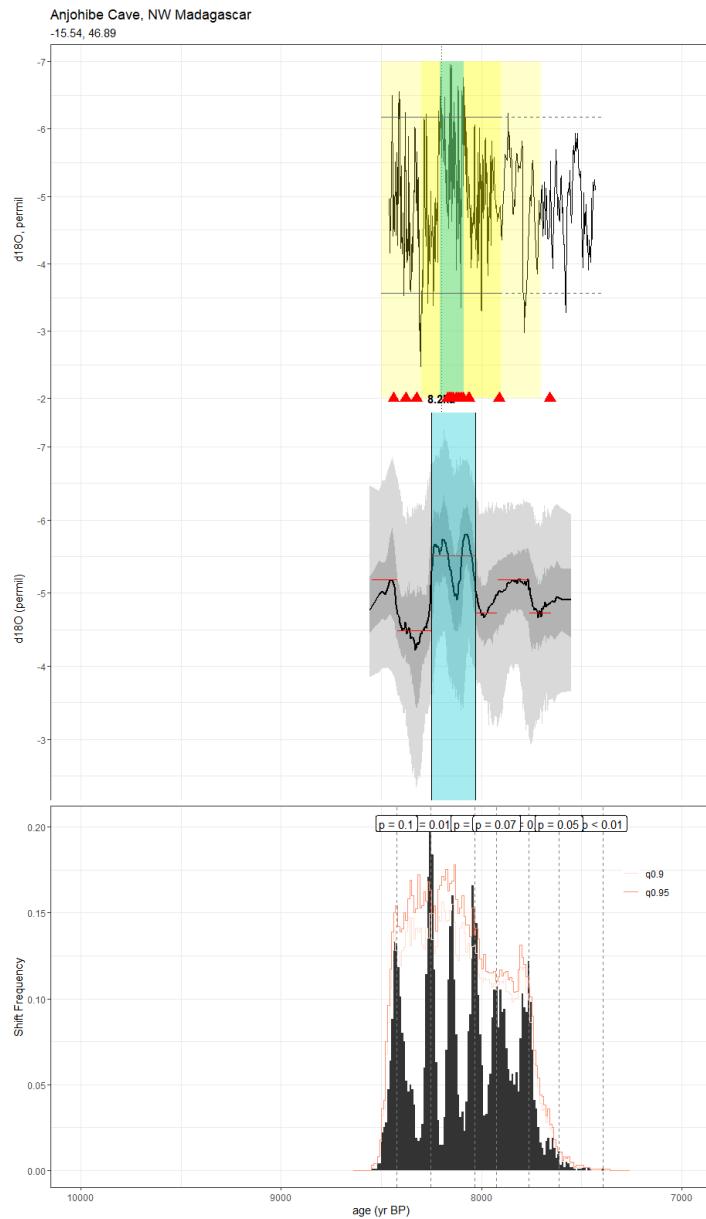
Total Number of Records per Region



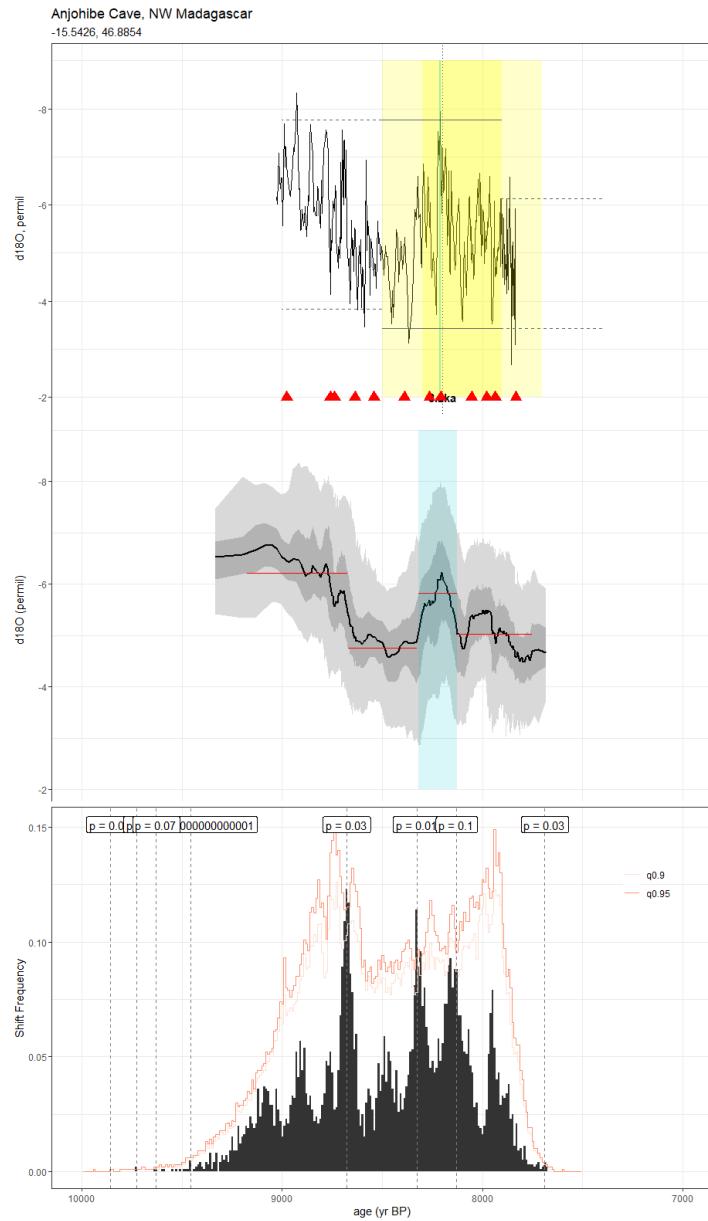
**Figure A1.** Locations of proxy records within climate reference regions defined in Iturbide et al. (2020).



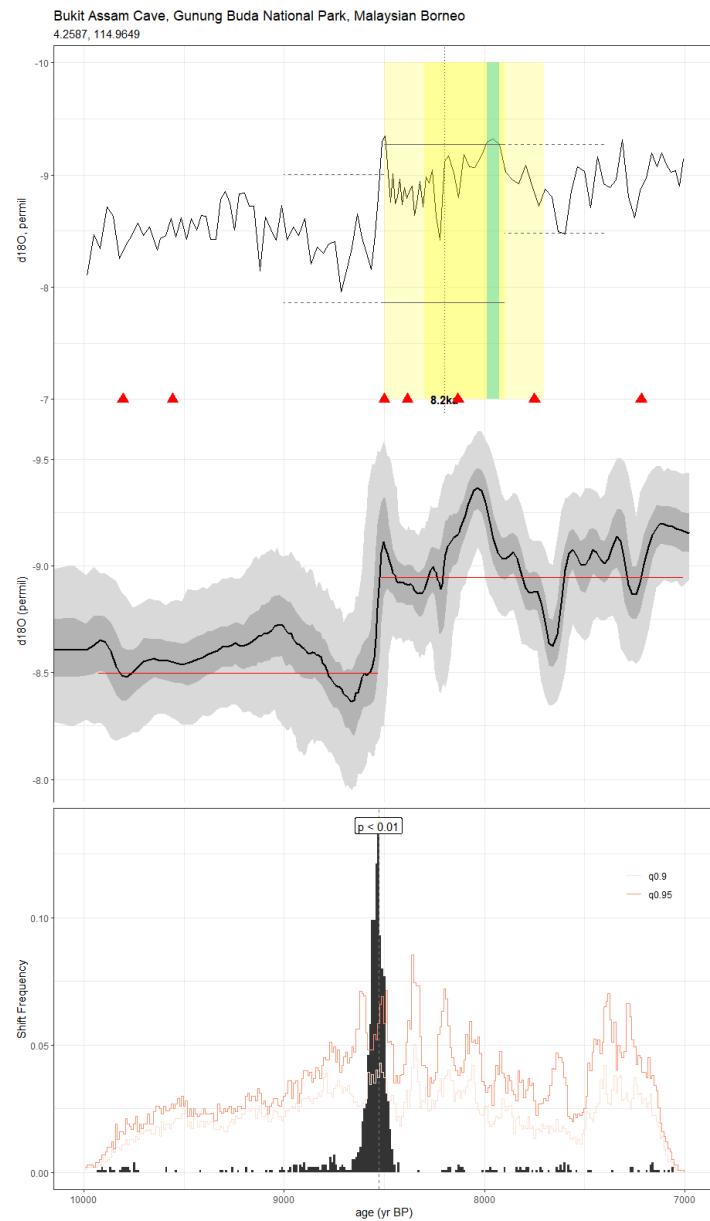
**Figure A2.** The difference in surface air temperatures between the last 50 years of the “hose” and “ctrl” simulations, overlaid by contour intervals indicating the range of temperatures in the “ctrl” simulation over the full 100 years. Blue shaded areas represent anomalously cold regions, while anomalously warm regions are shaded in red on a global (a) and tropical (b) level.



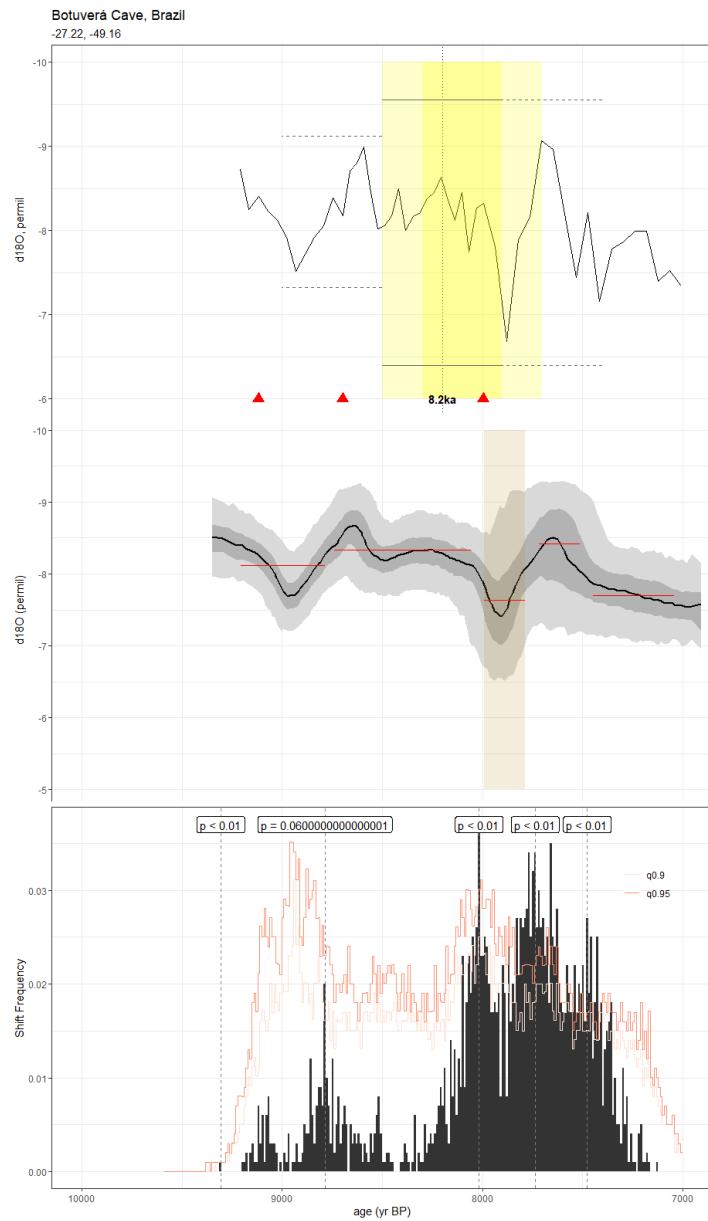
**Figure A.3.** A stackplot from the speleothem record of Duan et al., 2021 (ABC1). The top panel shows the raw oxygen isotope time series with 8.3-7.9ka highlighted in darker yellow and 8.5-7.7ka highlighted in lighter yellow. The middle panel shows the same time series with age and paleodata ensemble uncertainty ribbons. The horizontal red lines represent mean values assigned to the data by actR, with discontinuities indicating significant changepoints. The lower panel depicts the frequency of shifts detected in the ensemble dataset (black) relative to 100 null hypothesis surrogate datasets (orange). The age model in the original publication was based on the MOD-AGE algorithm, while the age model used in this synthesis was constructed using the geoChronR package and BACON algorithm.



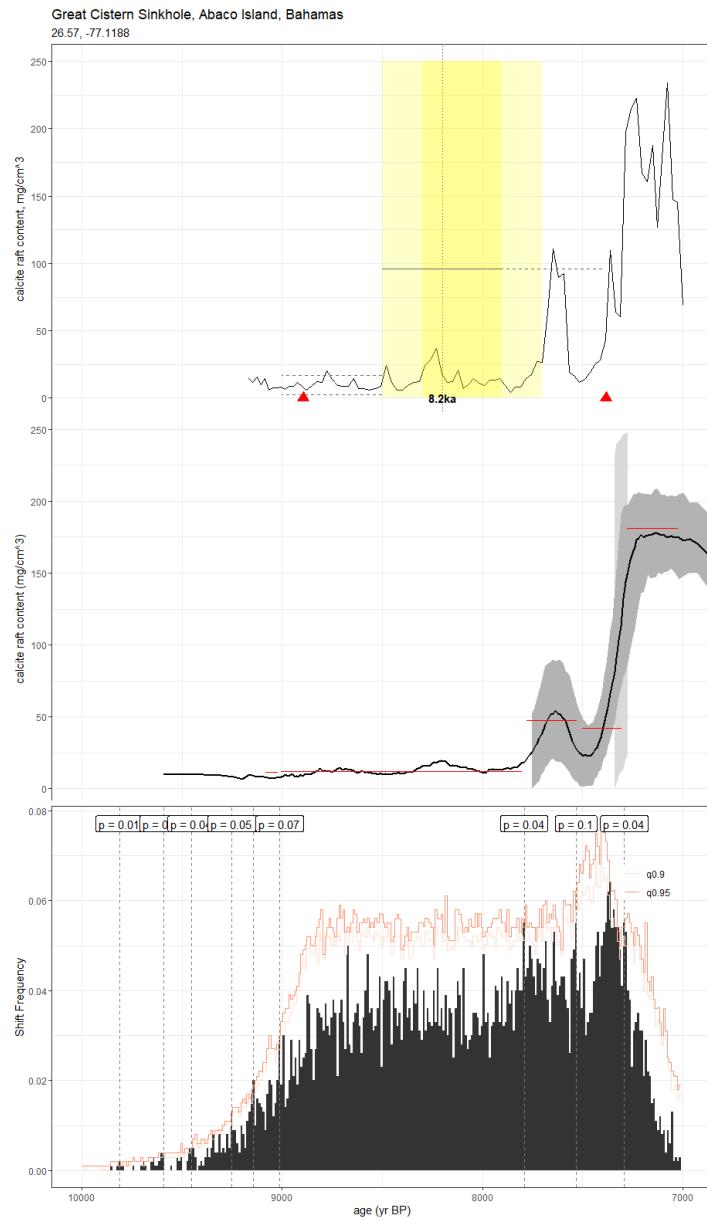
**Figure A4.** As in Fig. A3, but for the speleothem record of Voarintsoa et al., 2017 (ANJB2). The age model of the original publication was constructed using StalAge. Here, we used the BACON age ensemble from SISALv2.



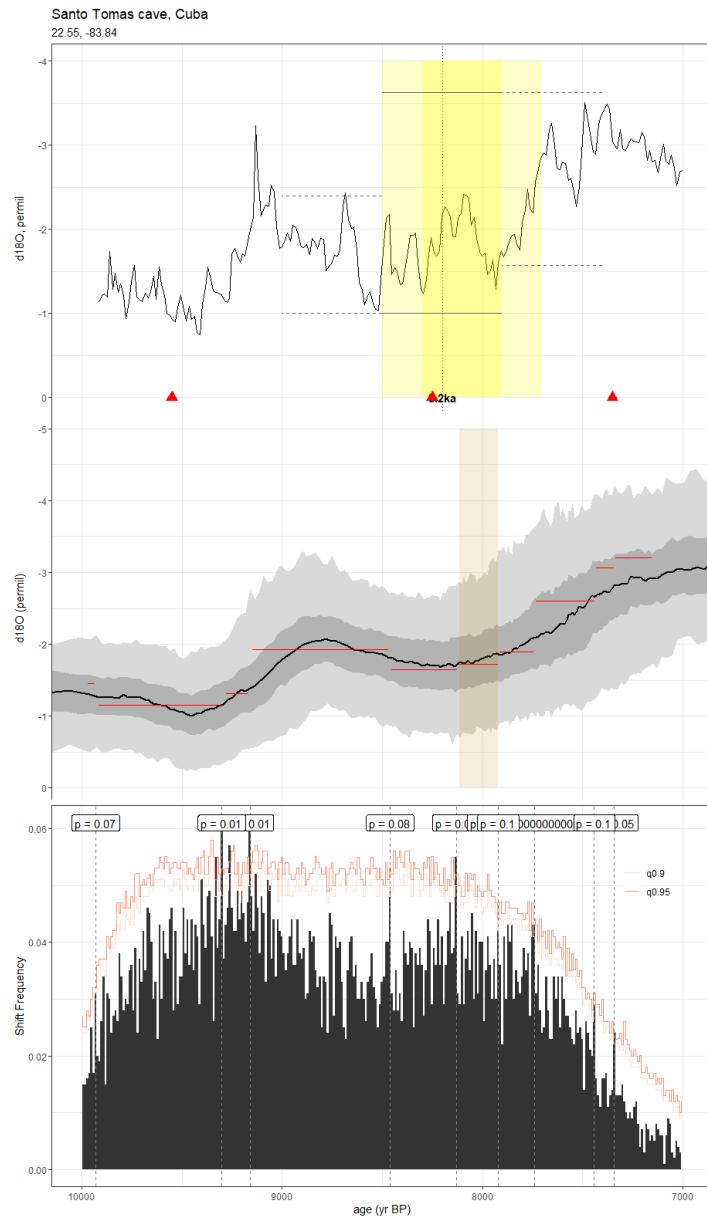
**Figure A5.** As in Fig. A3, but for the speleothem record of Chen et al., 2016 (BA03). The published age model was based on the StalAge algorithm, but here, we use the BACON ensemble from SISALv2.



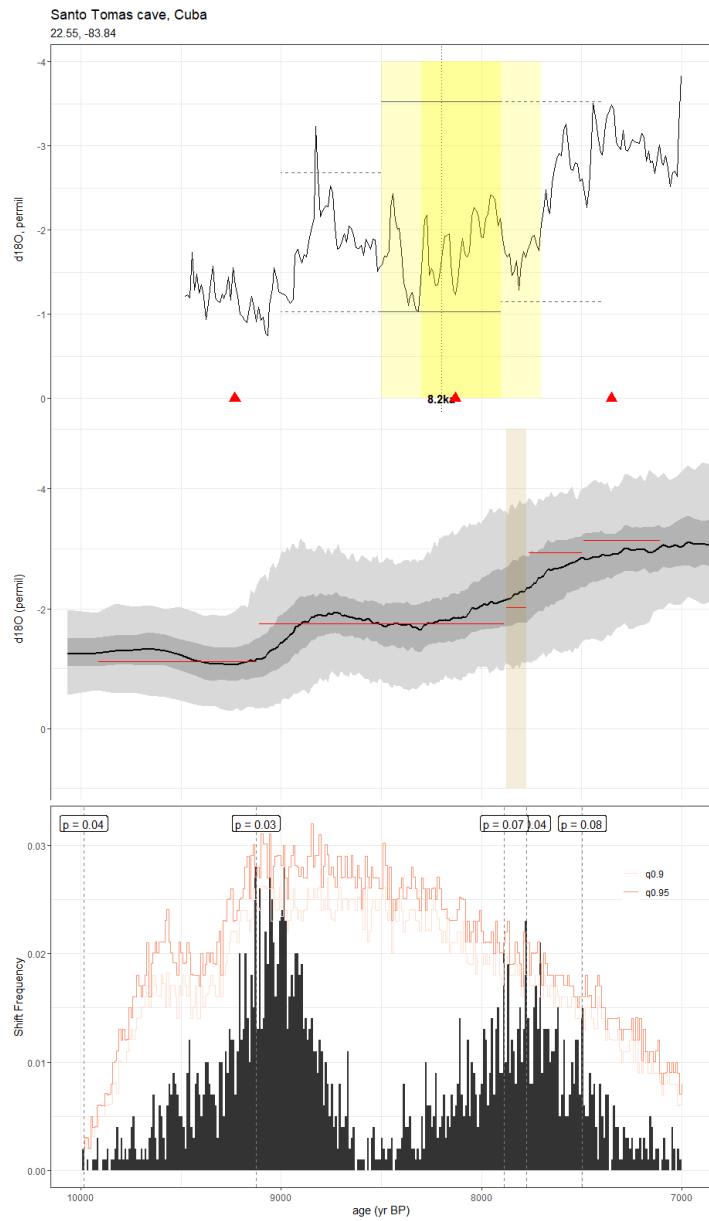
**Figure A6.** As in Fig. A3, but for the speleothem record of Bernal et al., 2016 (BTV21a). Information about the published age model was unreported. Here, we use the SISALv2 BACON age ensemble.



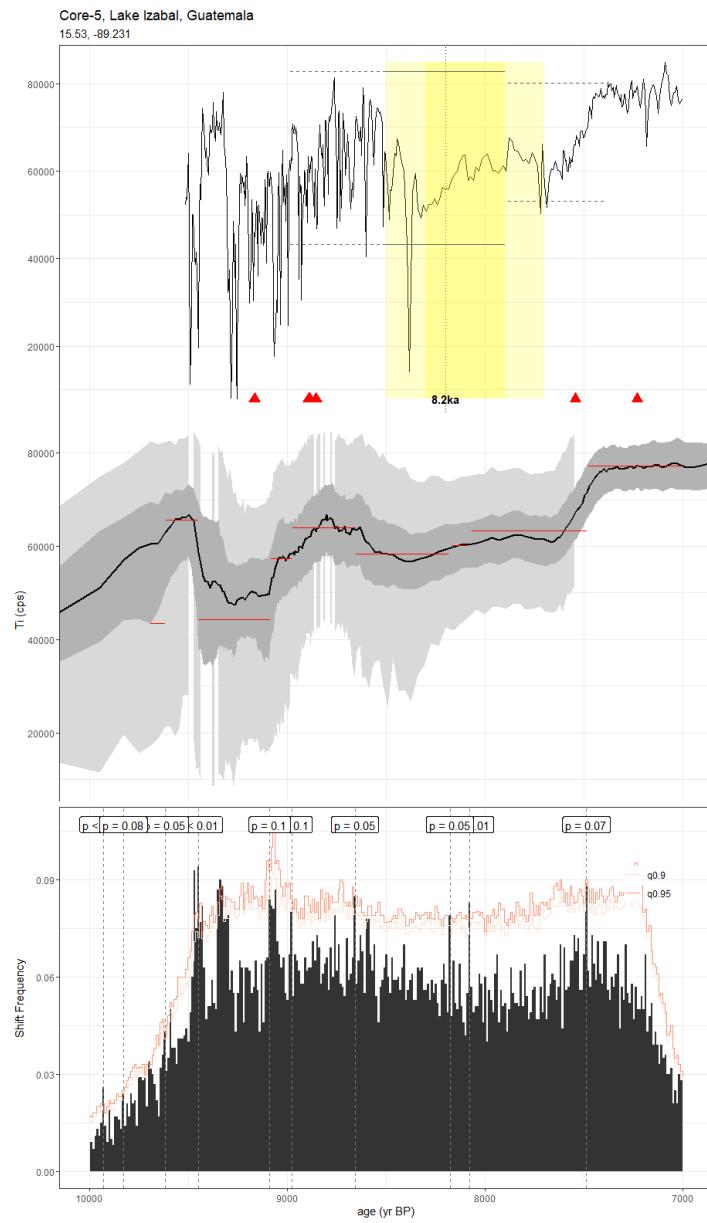
**Figure A7.** As in Fig. A3, but for the lacustrine calcite raft record of Sullivan et al., 2021 (C7). The published age model was constructed using the BACON algorithm and the IntCal13 calibration curve. Here, we reconstruct the BACON age ensemble using geoChronR and the IntCal20 calibration curve.



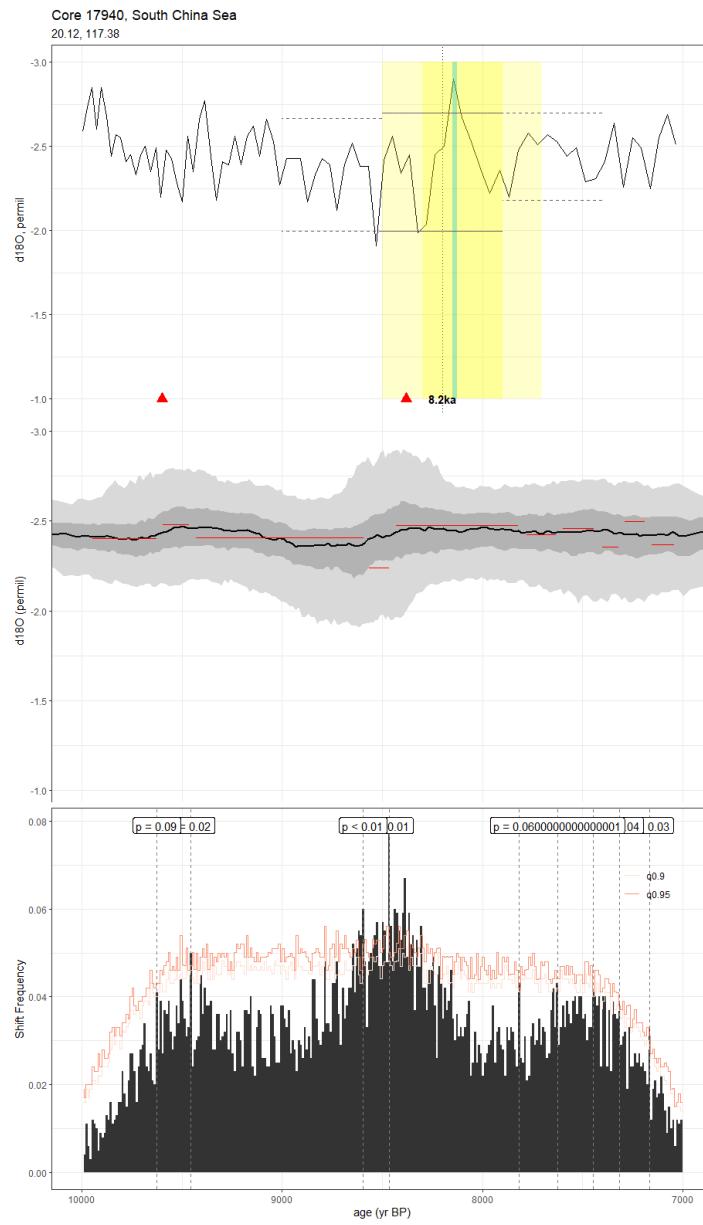
**Figure A8.** As in Fig. A3, but for the speleothem record of Fensterer et al., 2013 (CM2013). The published age model was constructed using the StalAge algorithm. Here, we use the SISALv2 copRa age ensemble.



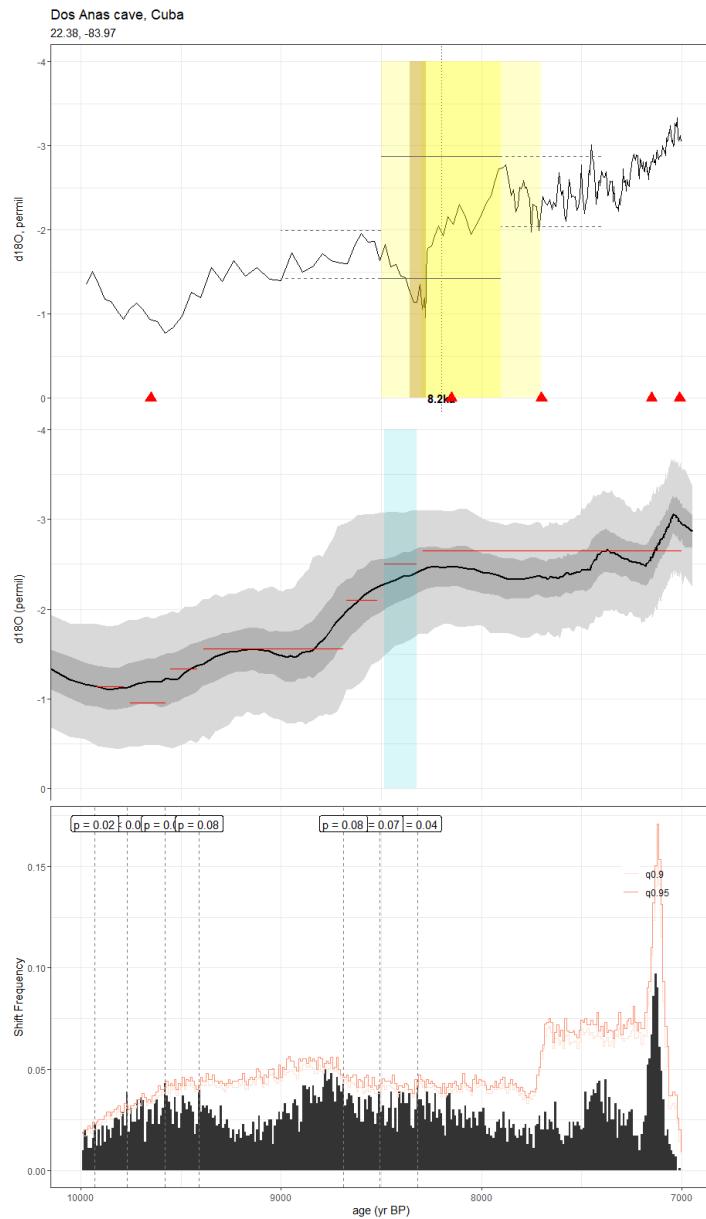
**Figure A9.** As in Fig. A3, but for the speleothem record of Warken et al., 2019 (CM2019). The published age model was constructed using the StalAge algorithm. Here, we use the SISALv2 BACON age ensemble.



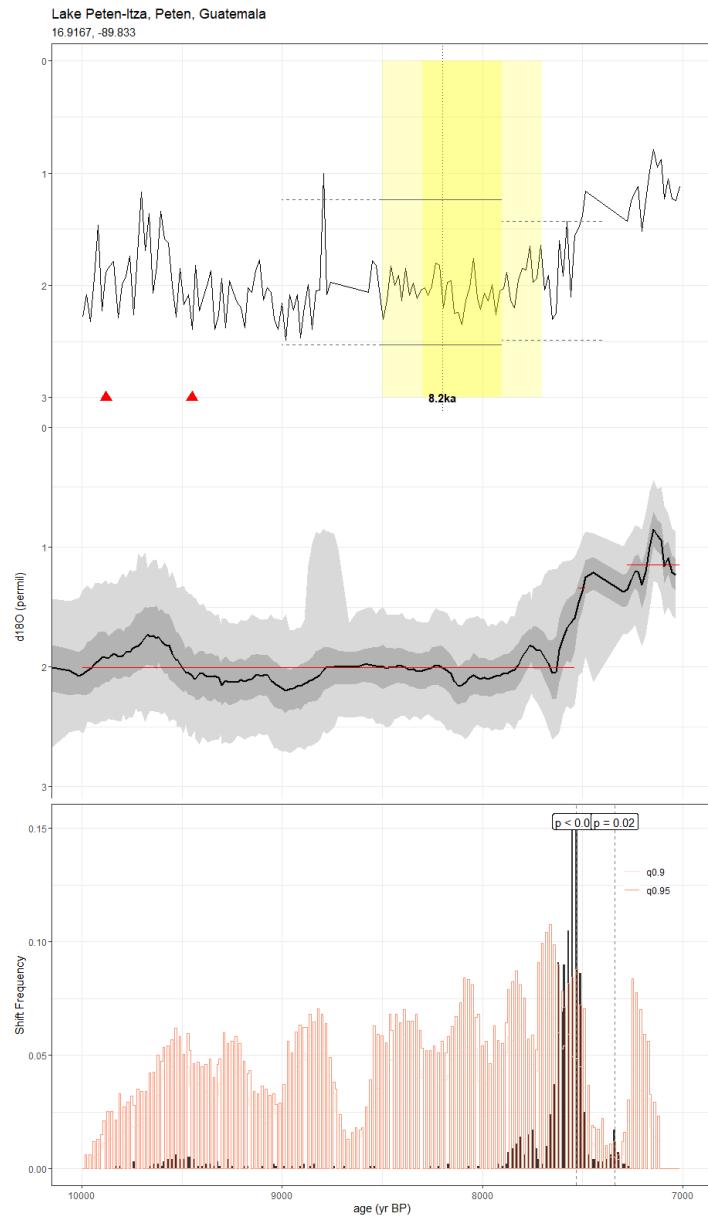
**Figure A10.** As in Fig. A3, but for the lacustrine titanium content record of Duarte et al., 2021 (Core5LI). The published age model was constructed using BACON using the IntCal20 calibration curve, and here, we construct our age ensemble using the BACON algorithm included with geoChronR.



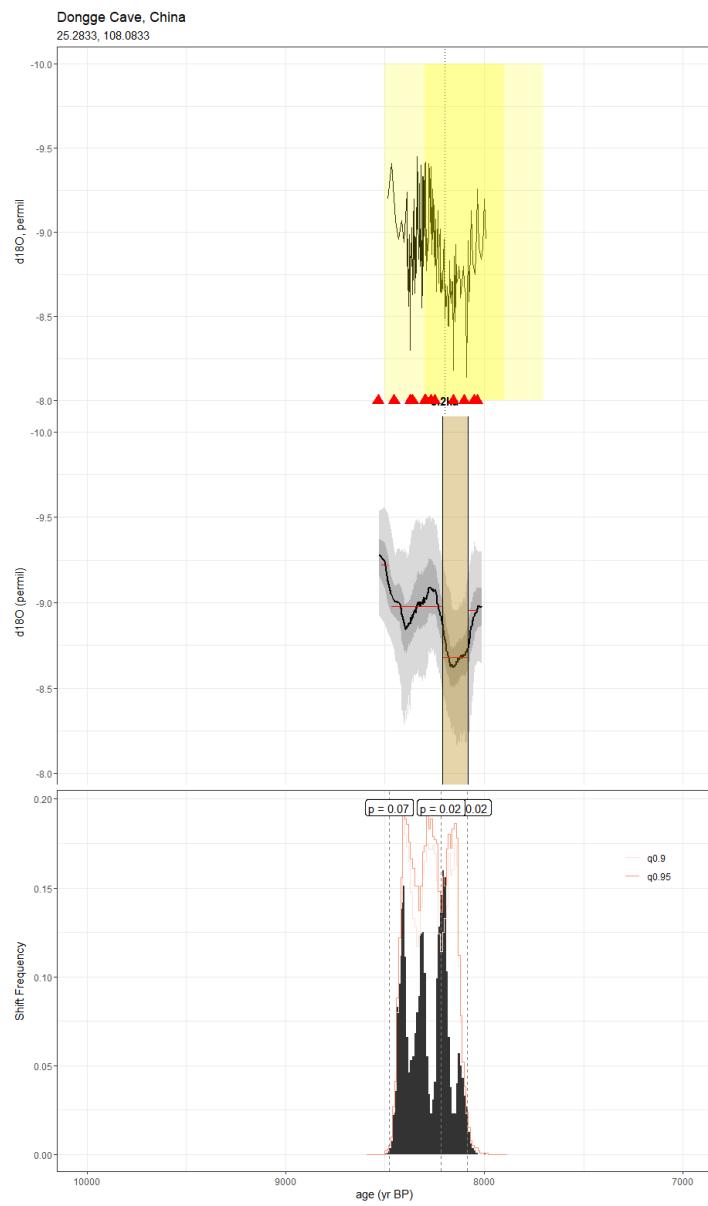
**Figure A11.** As in Fig. A3, but for the foraminifera record of Wang et al., 1999 (Core17940). The published age model was constructed using CALIB 3.0.3, corrected for a 400-year reservoir age and unspecified calibration curve. We constructed our age ensemble using the BACON algorithm included in geoChronR using the IntCal20 calibration curve.



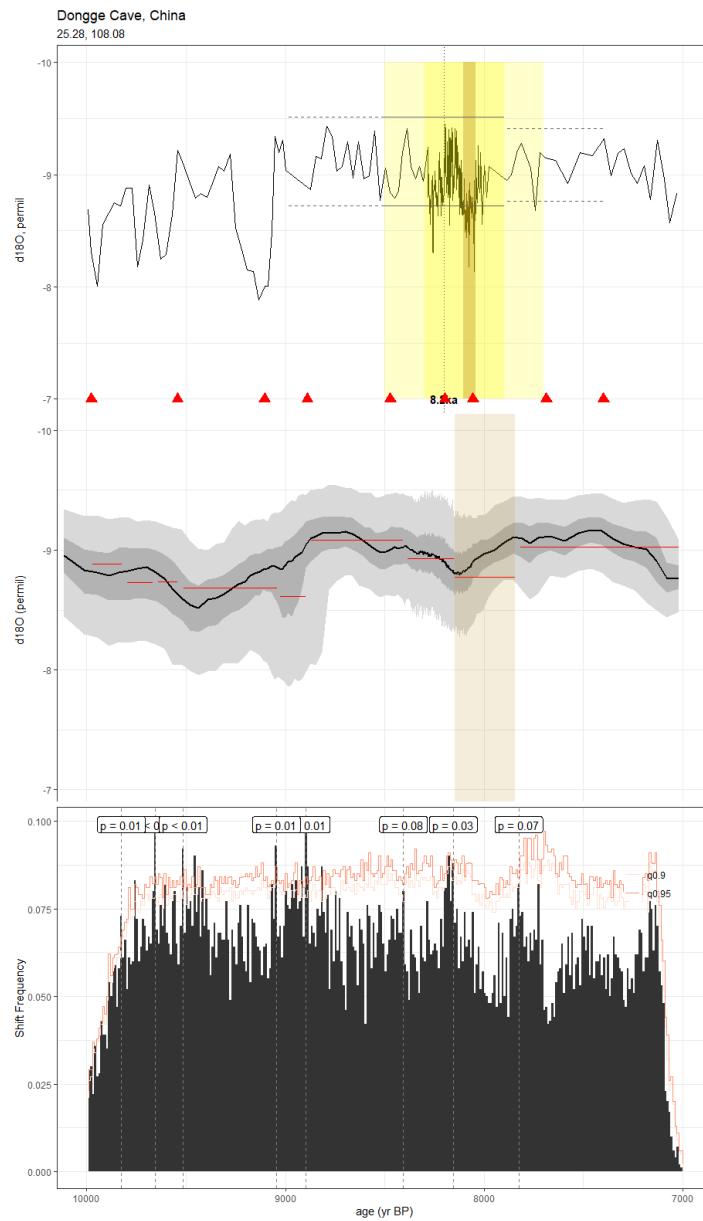
**Figure A12.** As in Fig. A3, but for the speleothem record of Fensterer et al., 2013 (CP). The published age model was constructed using the StalAge algorithm. We use the Bchron age ensemble constructed in SISALv2 here.



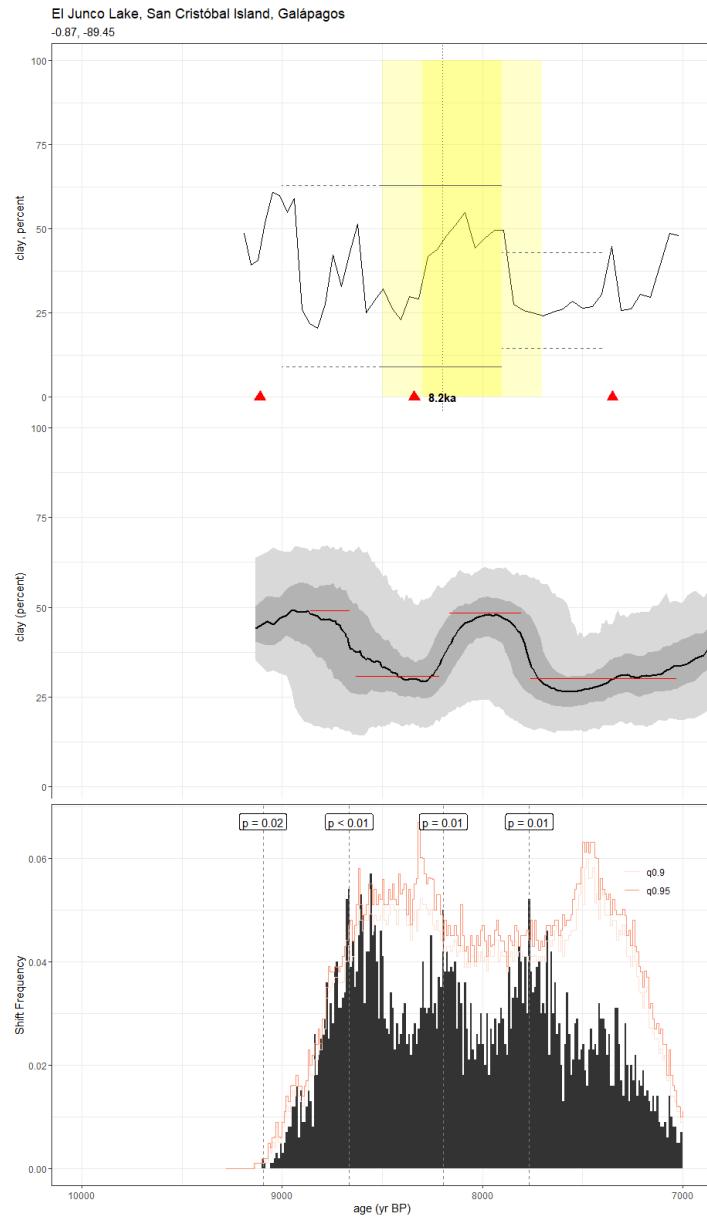
**Figure A13.** As in Fig. A3, but for the lacustrine gastropod  $\delta^{18}\text{O}$  record of Curtis et al., 1998 (Curtis6VII93). The published age model was constructed by linearly interpolating between  $^{14}\text{C}$  dates derived from terrestrial wood and charcoal samples. Here, we construct the age ensemble using the BACON algorithm included in geoChronR.



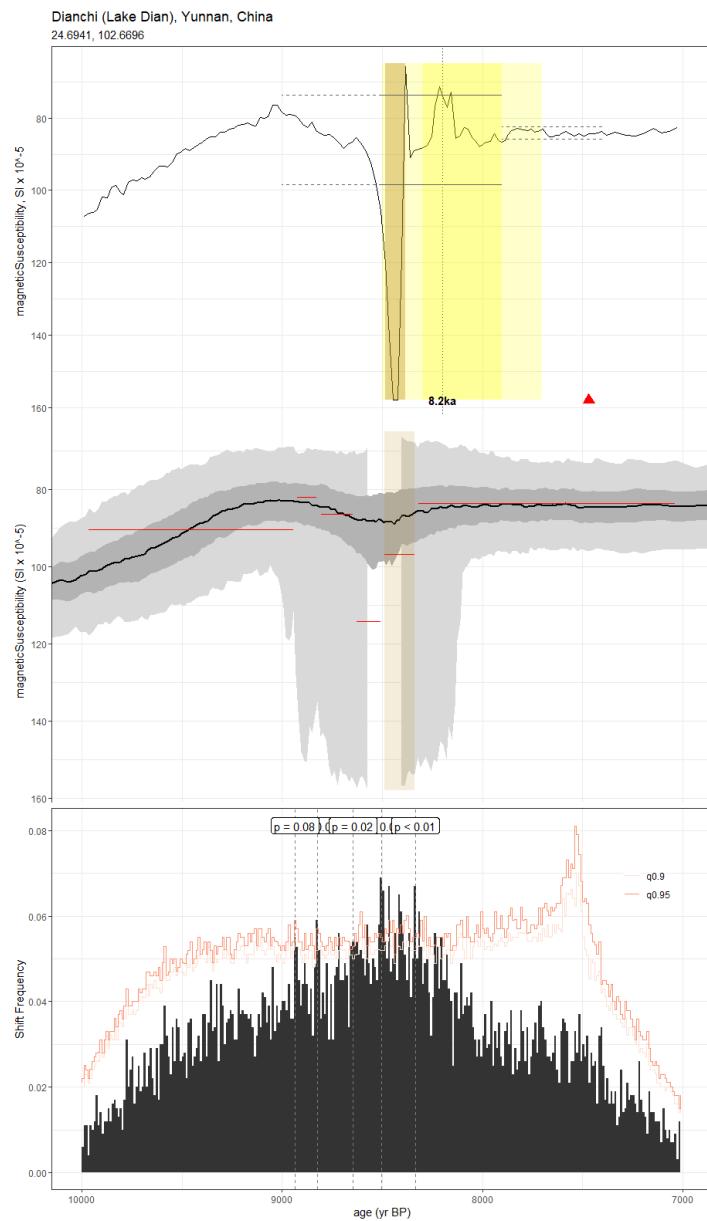
**Figure A14.** As in Fig. A3, but for the speleothem record of Cheng et al., 2009 (D4Cheng). The age model used in the original publication was unreported. Here, we use the BACON algorithm included in geoChronR to produce our age ensemble.



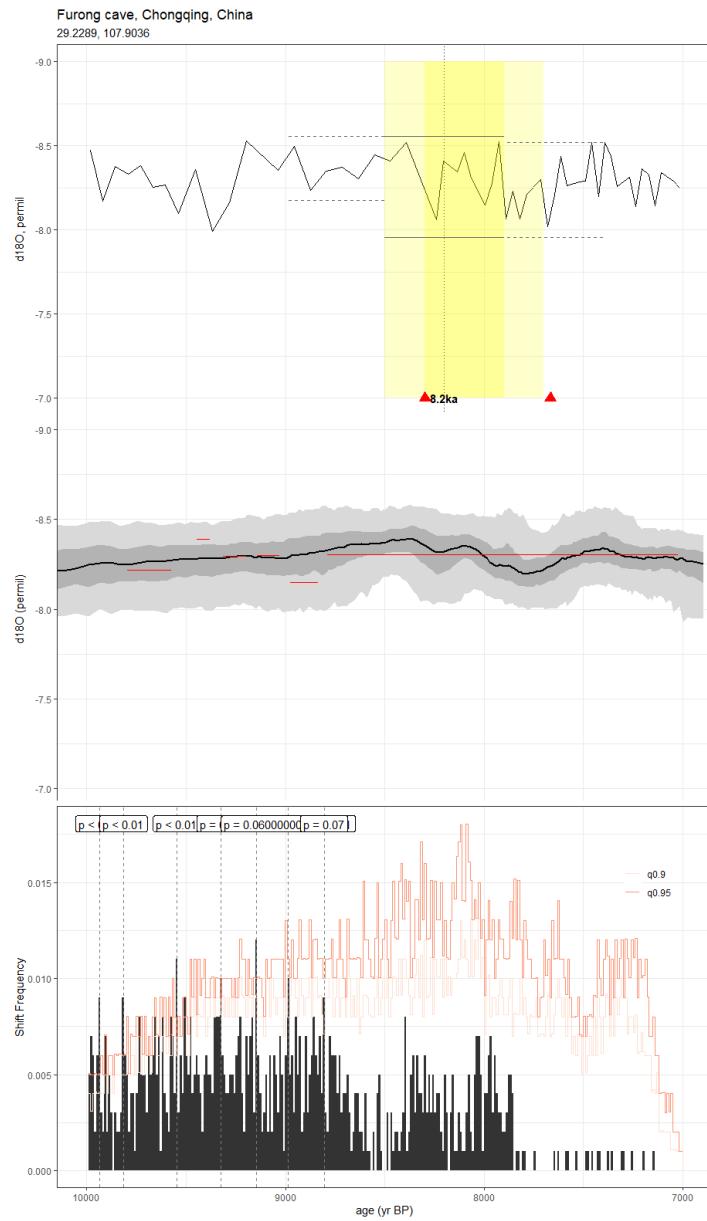
**Figure A15.** As in Fig. A3, but for the speleothem record of Dykoski et al., 2005 (D4Dykoski). The published age model was constructed by linearly interpolating between U/Th dates. Here, we reconstruct the age model using the BACON algorithm in geoChronR.



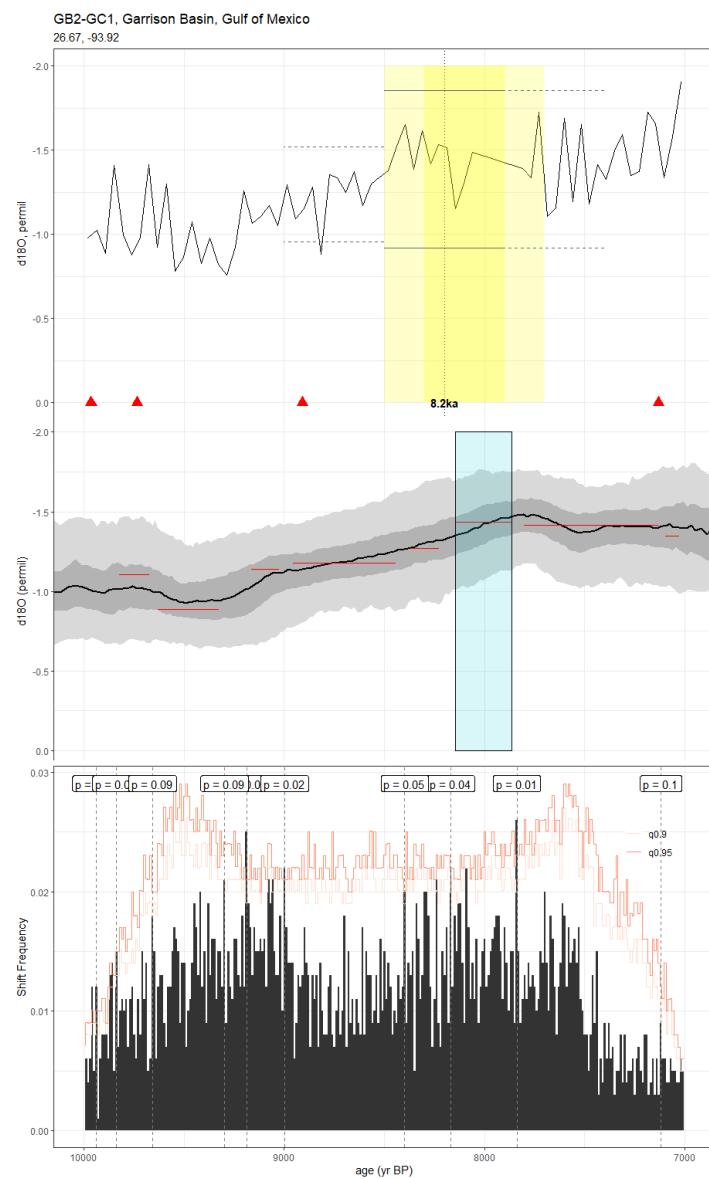
**Figure A16.** As in Fig. A3, but for the lacustrine clay content record of Conroy et al., 2008 (EJConroy). The published age model was constructed using CALIB 5.0 with the Southern Hemisphere dataset. The age ensemble presented here was created using the BACON algorithm with the SHCal20 calibration curve in geoChronR.



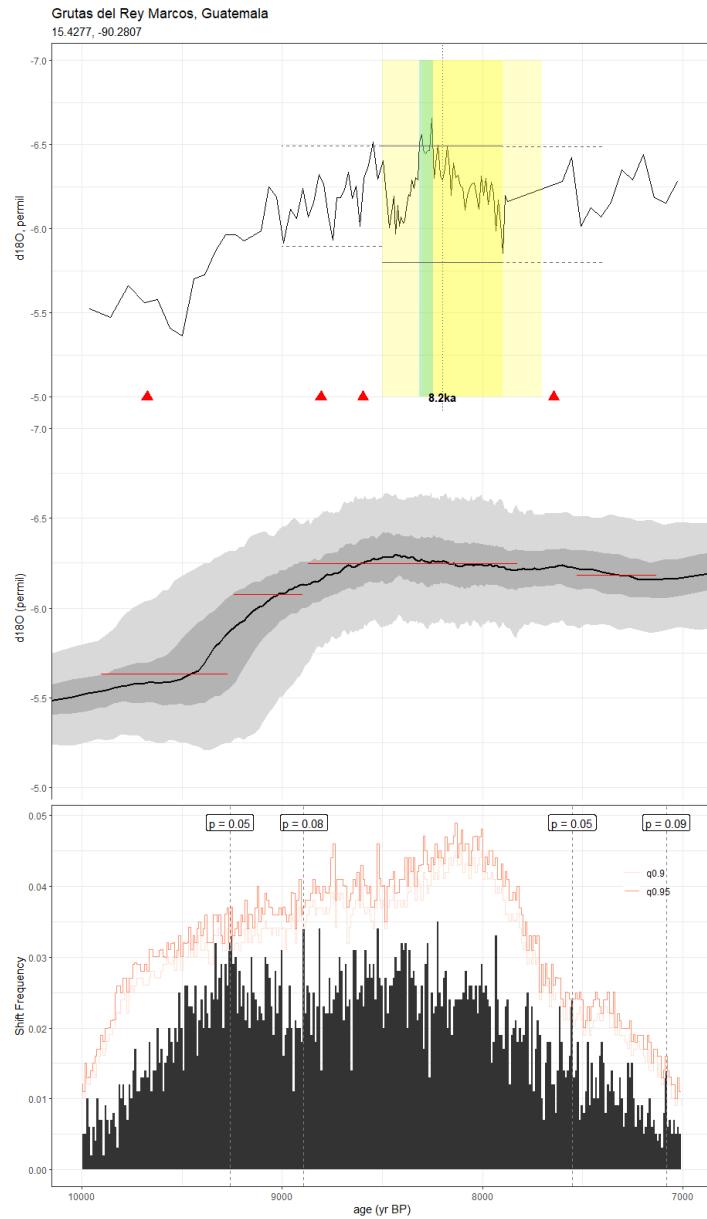
**Figure A17.** As in Fig. A3, but for the lake sediment magnetic susceptibility record of Hillman et al., 2021 (F14). The original age model was constructed using BACON with the IntCal20 calibration curve. Here, we have reconstructed it using the BACON algorithm in geoChronR.



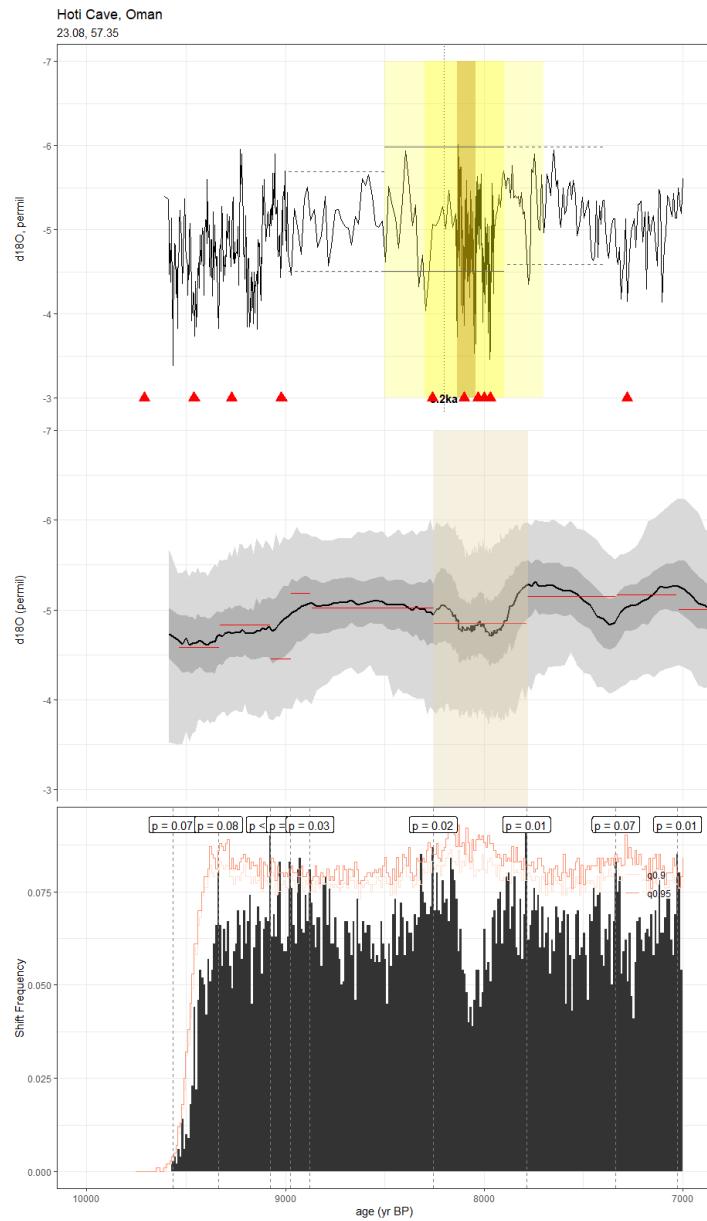
**Figure A18.** As in Fig. A3, but for the speleothem record of Li et al., 2011 (FR5). The age modeling algorithm used to construct the original age model was unreported, but leveraged the IntCal09 calibration curve. Here, we use the copRa age ensemble included in SISALv2.



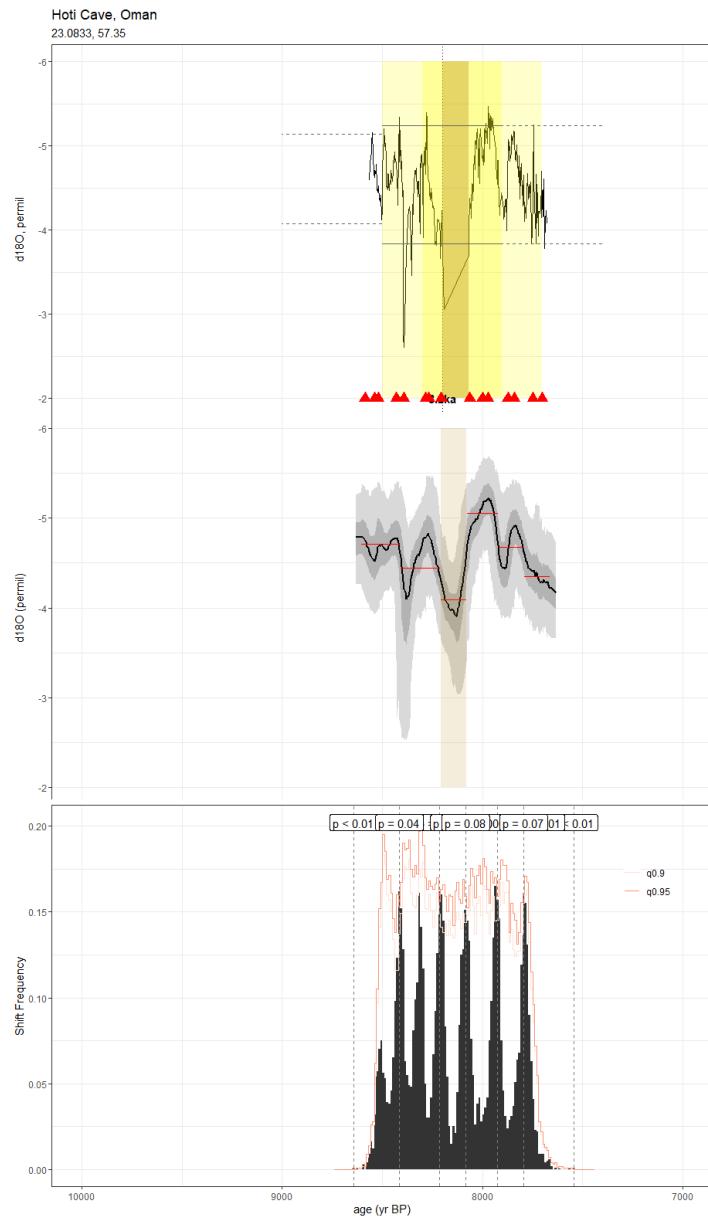
**Figure A19.** As in Fig. A3, but for the foraminifera record of Thirumulai et al., 2021 (GB2GC1). The published age model was developed using the BACON algorithm and Marine13 calibration curve, which we reconstructed using geoChronR.



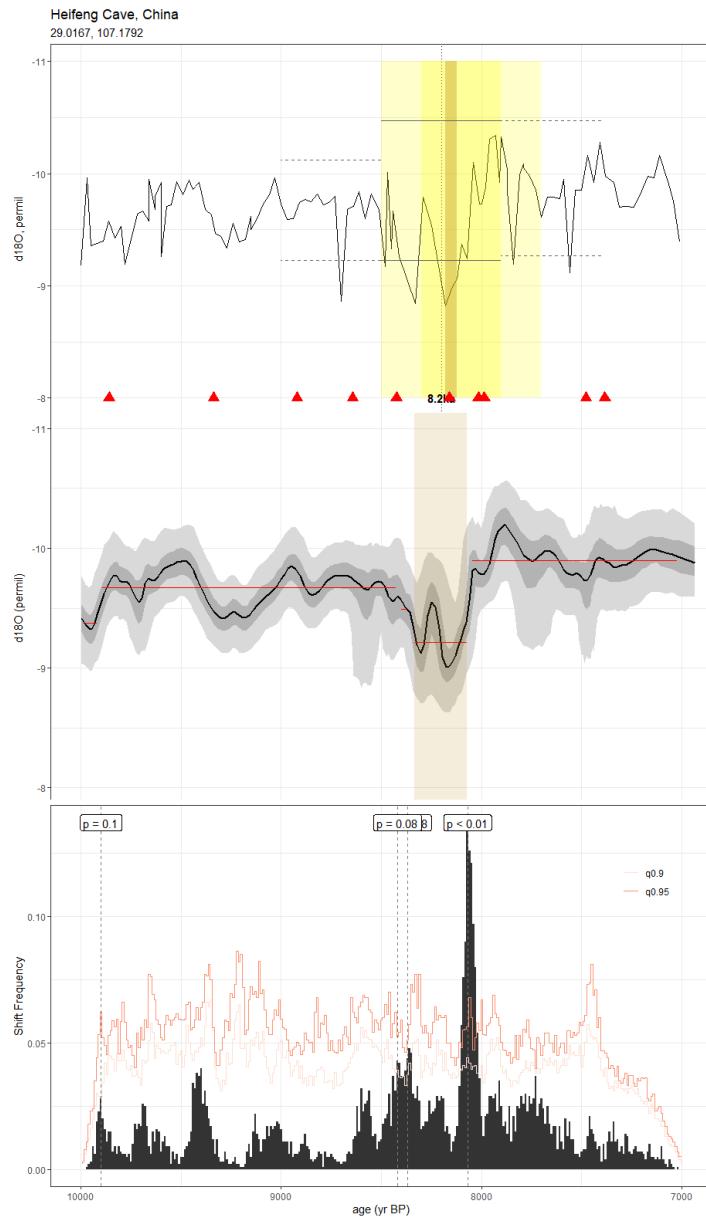
**Figure A20.** As in Fig. A3, but for the speleothem record of Winter et al., 2020 (GURM1). The published age model was constructed using the copRa algorithm, while we use the BACON ensemble produced for SISALv2 here.



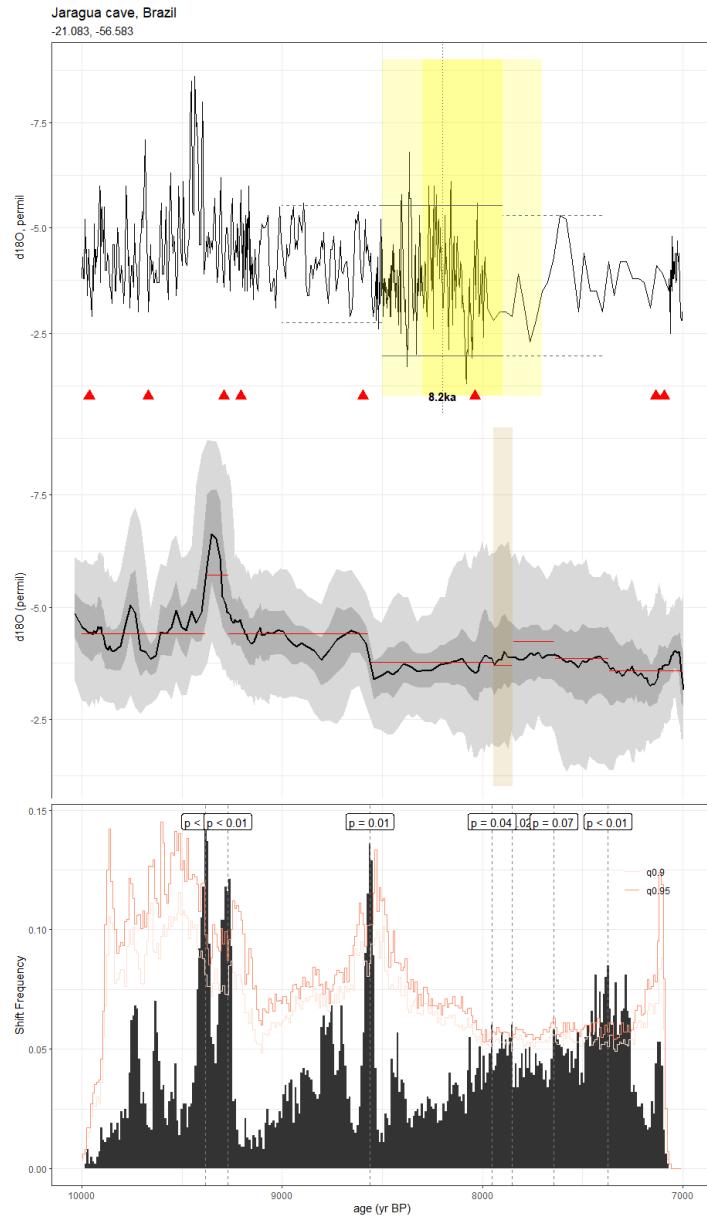
**Figure A21.** As in Fig. A3, but for the speleothem record of Neff et al., 2001 (H5). While the method used in the construction of the published time series was unreported, we leveraged the SISALv2 BACON ensemble for our analyses.



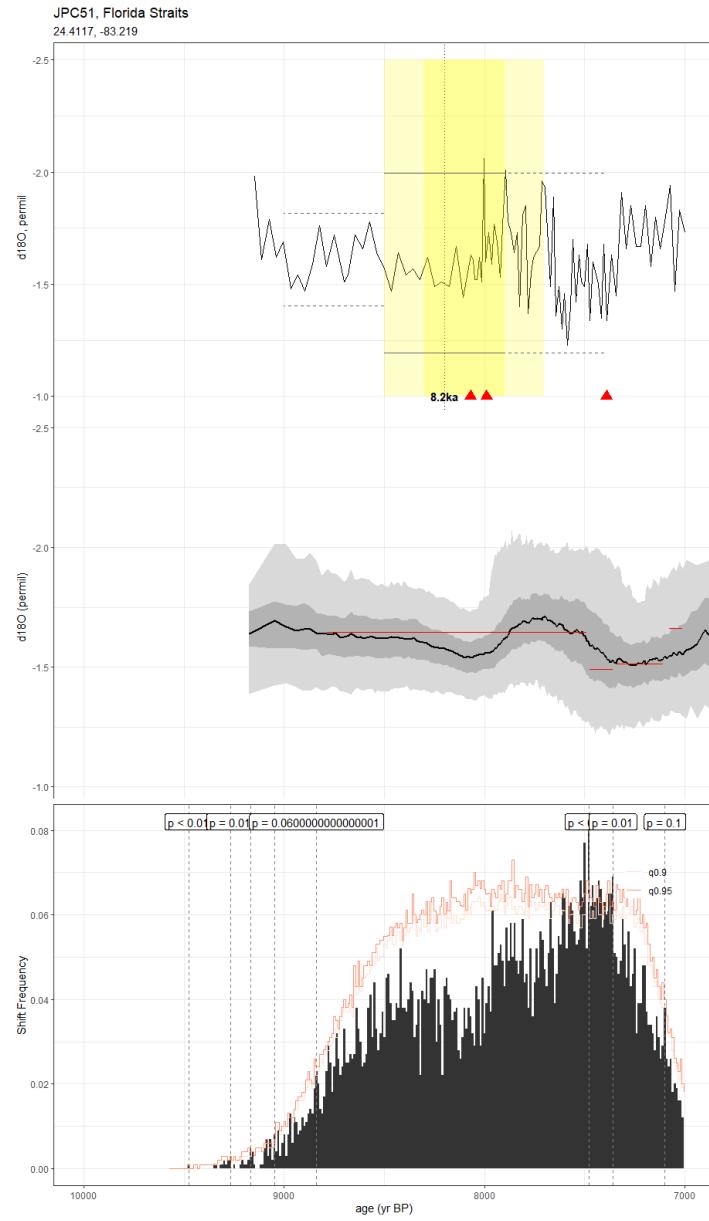
**Figure A22.** As in Fig. A3, but for the speleothem record of Cheng et al., 2009 (H14). The age modeling algorithm used to construct the original age model was unreported. Here, we constructed our age ensemble using BACON in geoChronR.



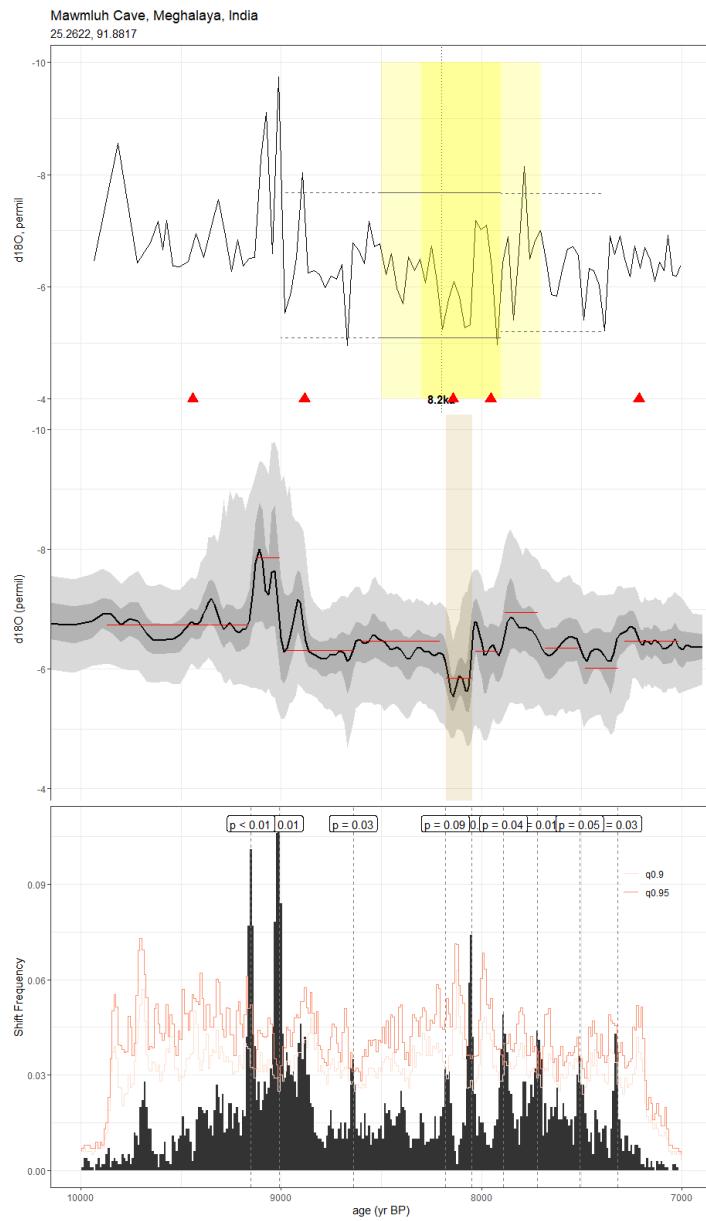
**Figure A23.** As in Fig. A3, but for the speleothem record of Yang et al., 2019 (HF01). The published age model was constructed via polynomial regression between radiometric dates. For our analyses, we leveraged the copRa age ensemble included in version 2 of the SISAL database.



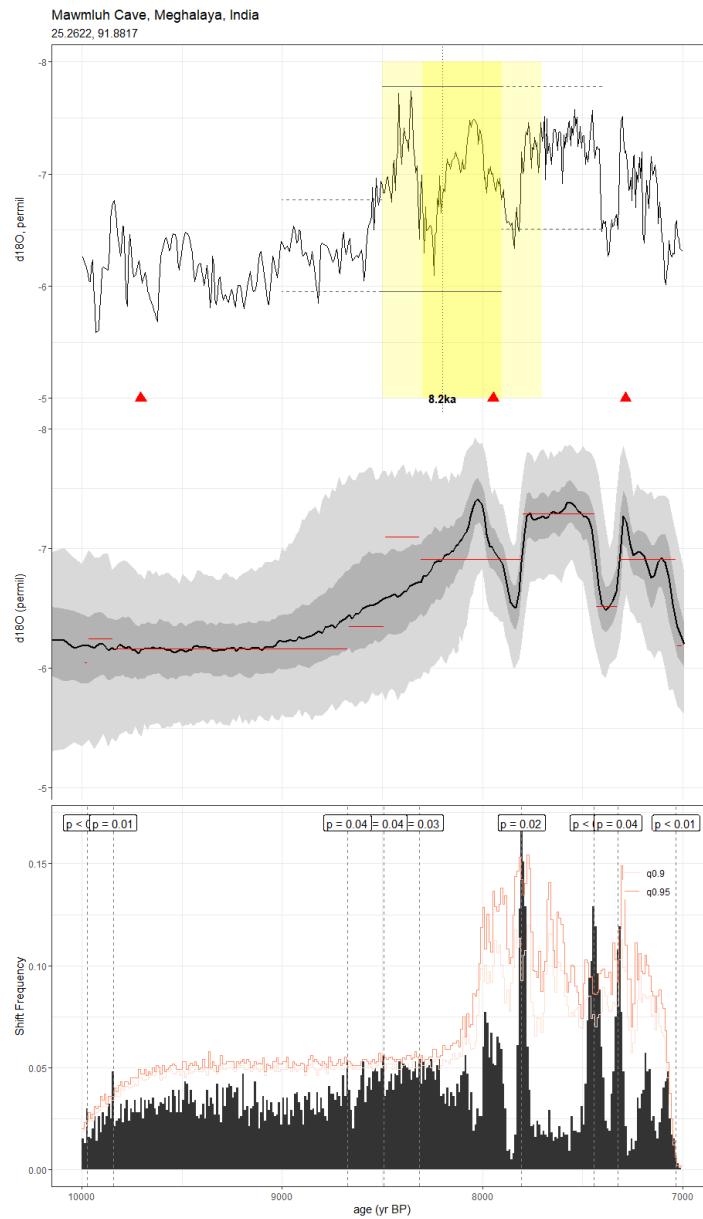
**Figure A24.** As in Fig. A3, but for the speleothem record of Novello et al., 2017 (JAR7). The published age model was constructed via linear interpolation between radiometric dates. For our analyses, we leveraged the BACON age ensemble produced for SISALv2.



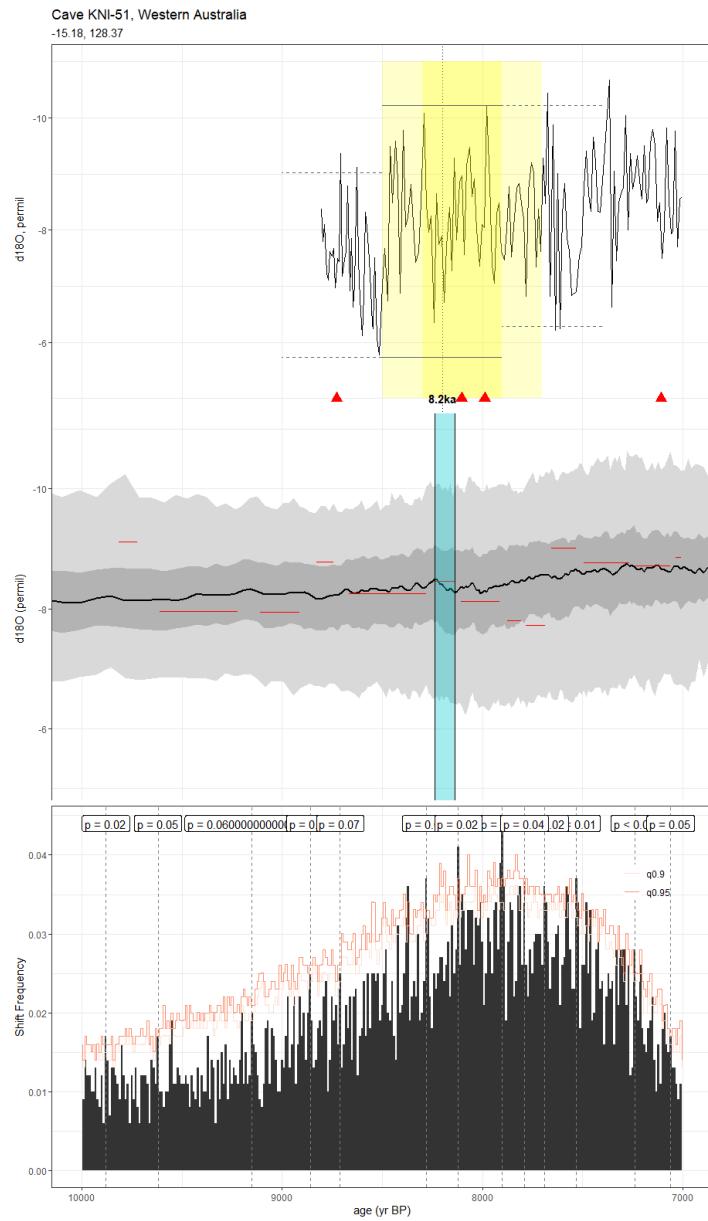
**Figure A25.** As in Fig. A3, but for the foraminifera record of Schmidt et al., 2012 (JPC51). The published age model was created using CALIB 6.0, with a standard -400 year reservoir age correction for surface waters. Here, we use the BACON algorithm included in geoChronR to produce our age ensemble.



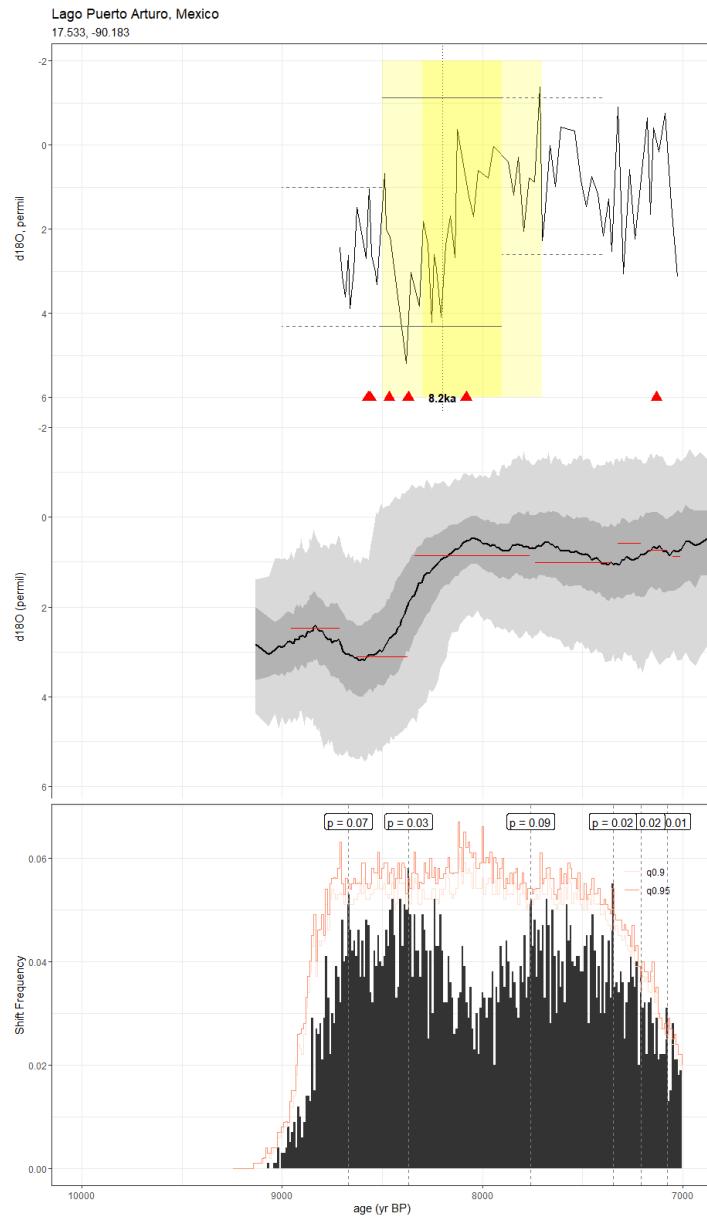
**Figure A26.** As in Fig. A3, but for the speleothem record of Huguet et al., 2018 (KM1). While the published age model was constructed using the StalAge algorithm, we leverage the Bchron age ensemble included in SISALv2 here.



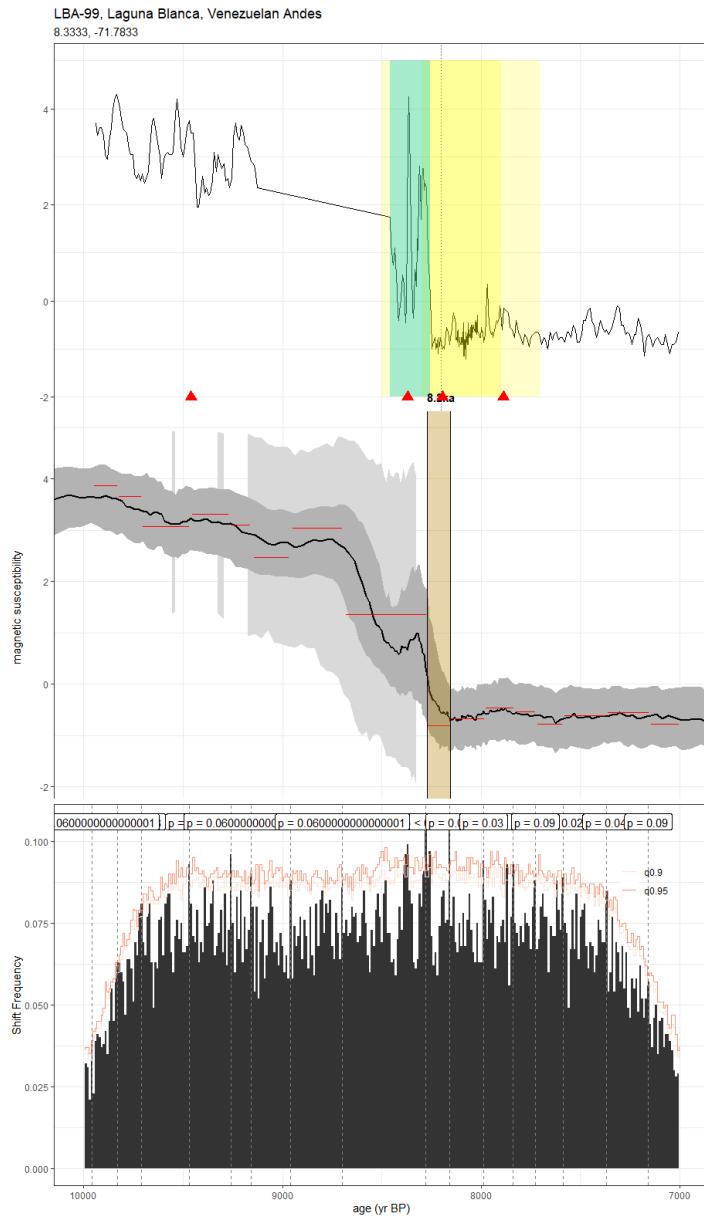
**Figure A27.** As in Fig. A3, but for the speleothem record of Berkelhammer et al., 2012 (KMA). The published age model was created using the StalAge algorithm. Here, we used the BACON age ensemble included in SISALv2.



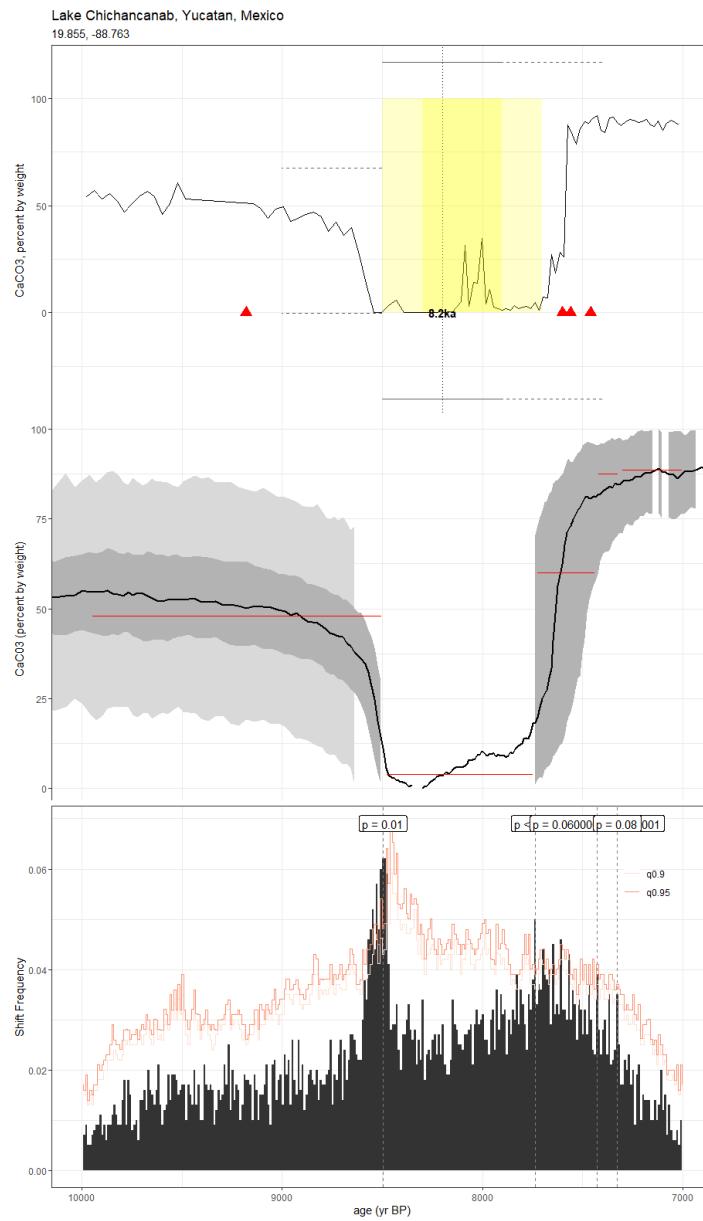
**Figure A28.** As in Fig. A3, but for the speleothem record of Denniston et al., 2013 (KN51). The method used in the construction of the published age model is unknown, but we use the copRa ensemble generated for SISALv2 here.



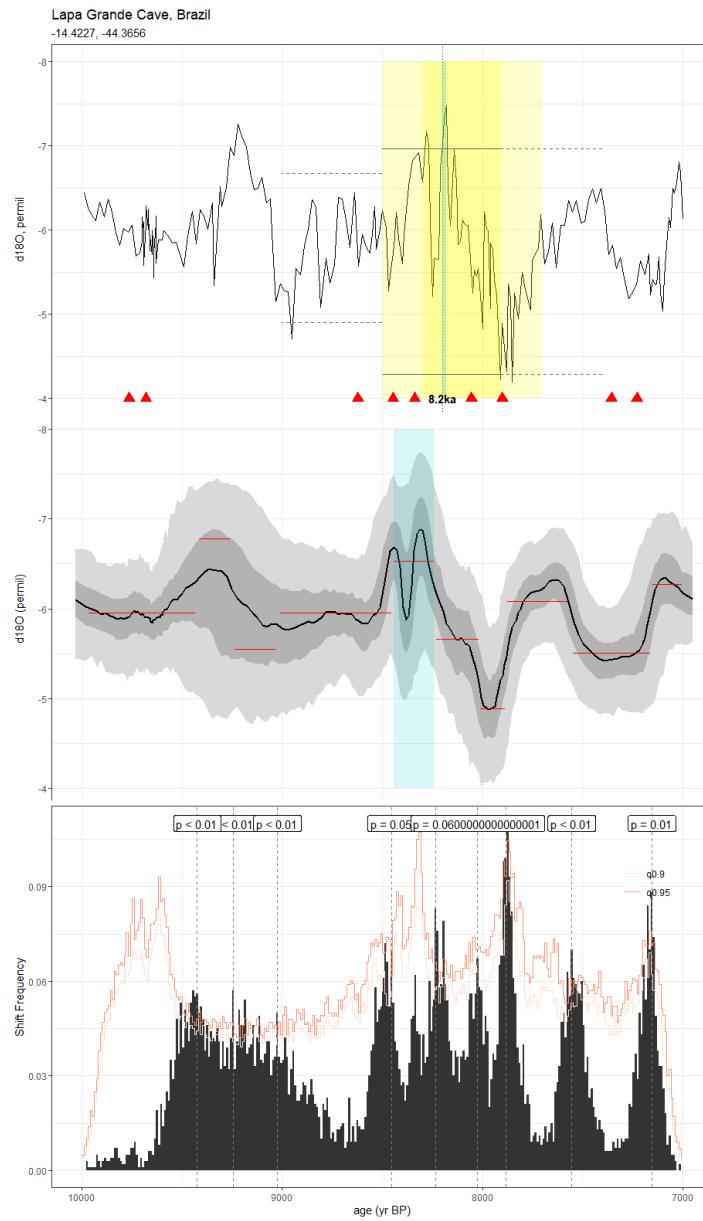
**Figure A29.** As in Fig. A3, but for the lake sediment  $\delta^{18}\text{O}$  record of Wahl et al., 2014 (LagoPuertoArturo). The published age model was constructed using CLAM 2.2 and the IntCal13 calibration curve. For our analyses, we reconstructed the age model using BACON and IntCal20 in geoChronR.



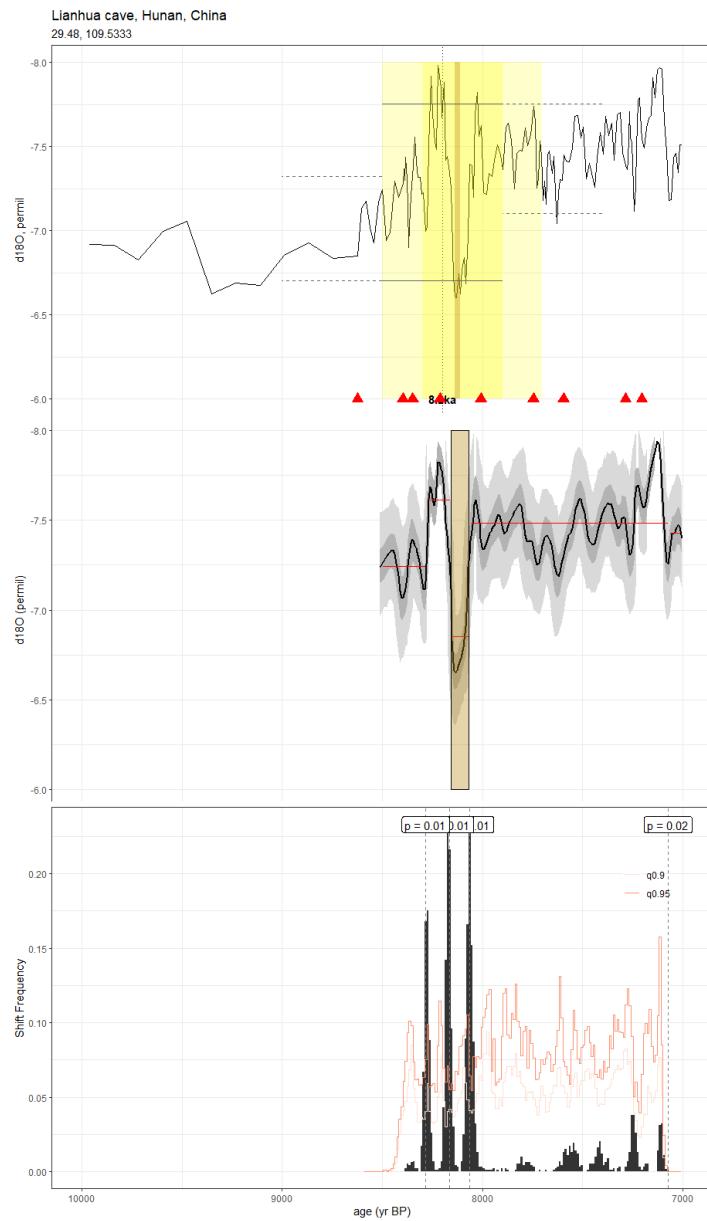
**Figure A30.** As in Fig. A3, but for the lake sediment magnetic susceptibility record of Pollisar et al., 2013 (LBA99). The published age model was constructed by linearly interpolating between radiometric dates with the IntCal04 calibration curve. Here, we constructed our ensemble using the BACON algorithm and IntCal20 curve included in geoChronR.



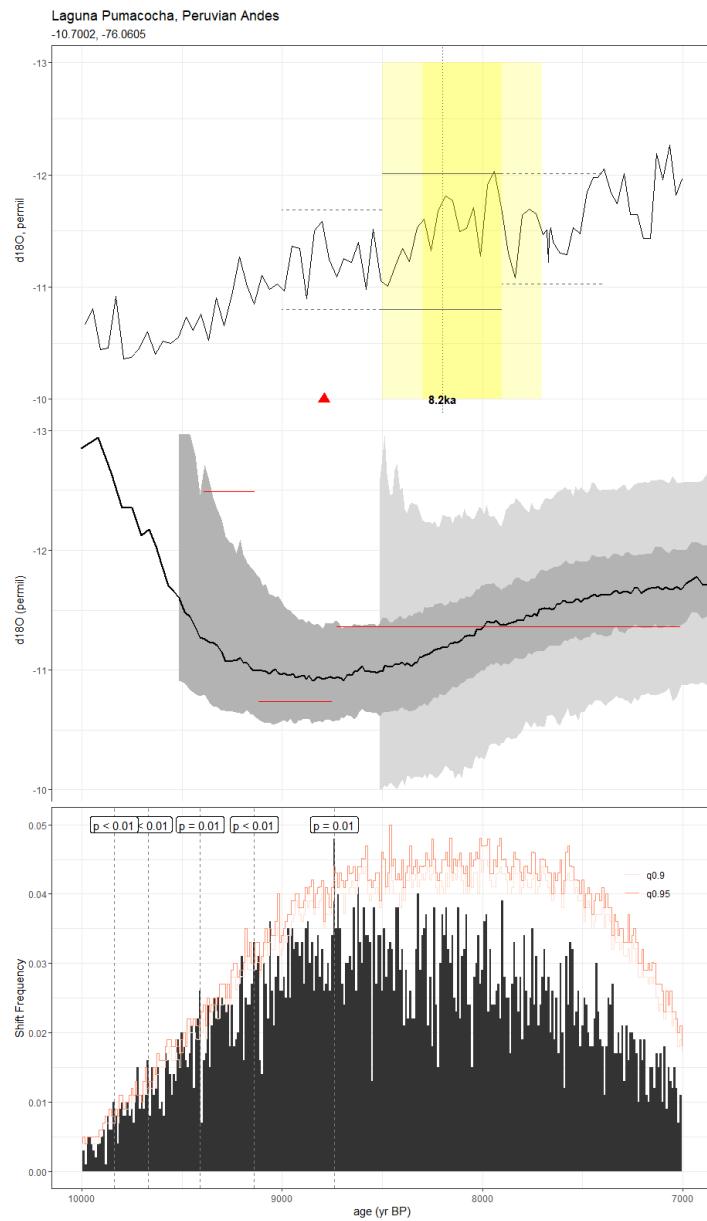
**Figure A31.** As in Fig. A3, but for the lake sediment record of Hodell et al., 1995 (LC1). The published age model was created using the decadal tree ring dataset in CALIB. Here, we use BACON with the IntCal20 calibration curve supplied by geoChronR.



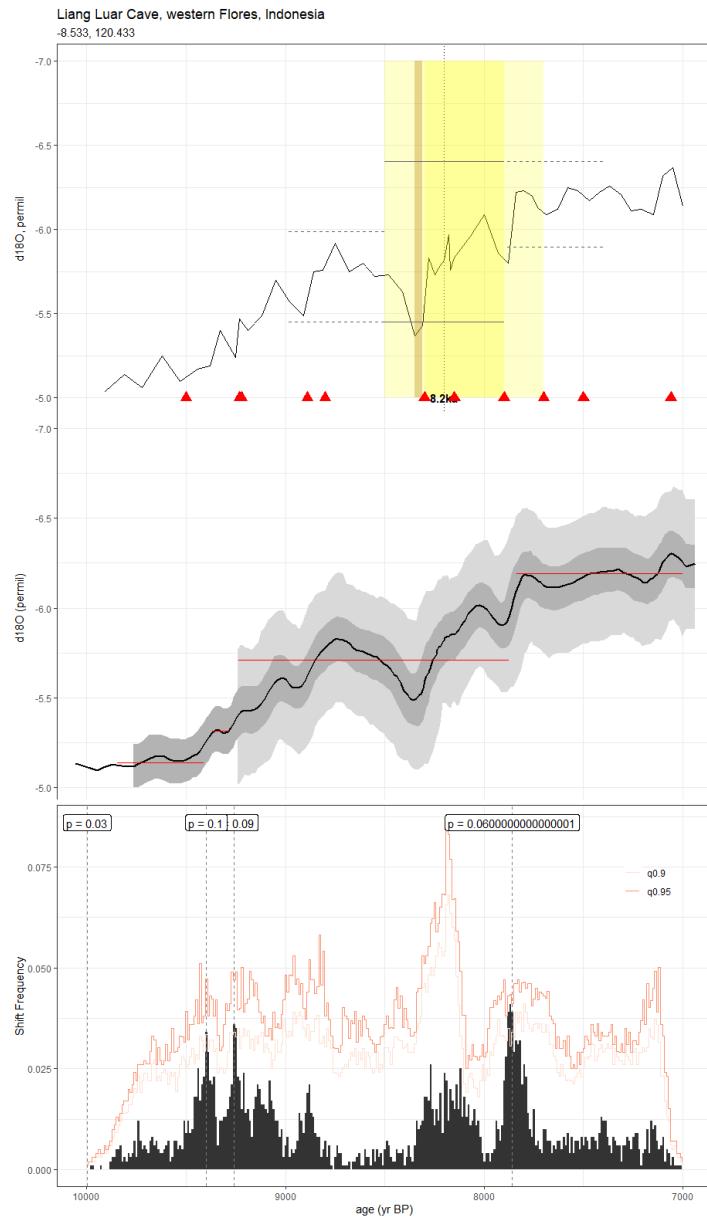
**Figure A32.** As in Fig. A3, but for the speleothem record of Strikis et al., 2011 (LG11). The original method used in construction of the published age model was unreported, but we leverage the BACON age ensemble published in SISALv2 for our analyses.



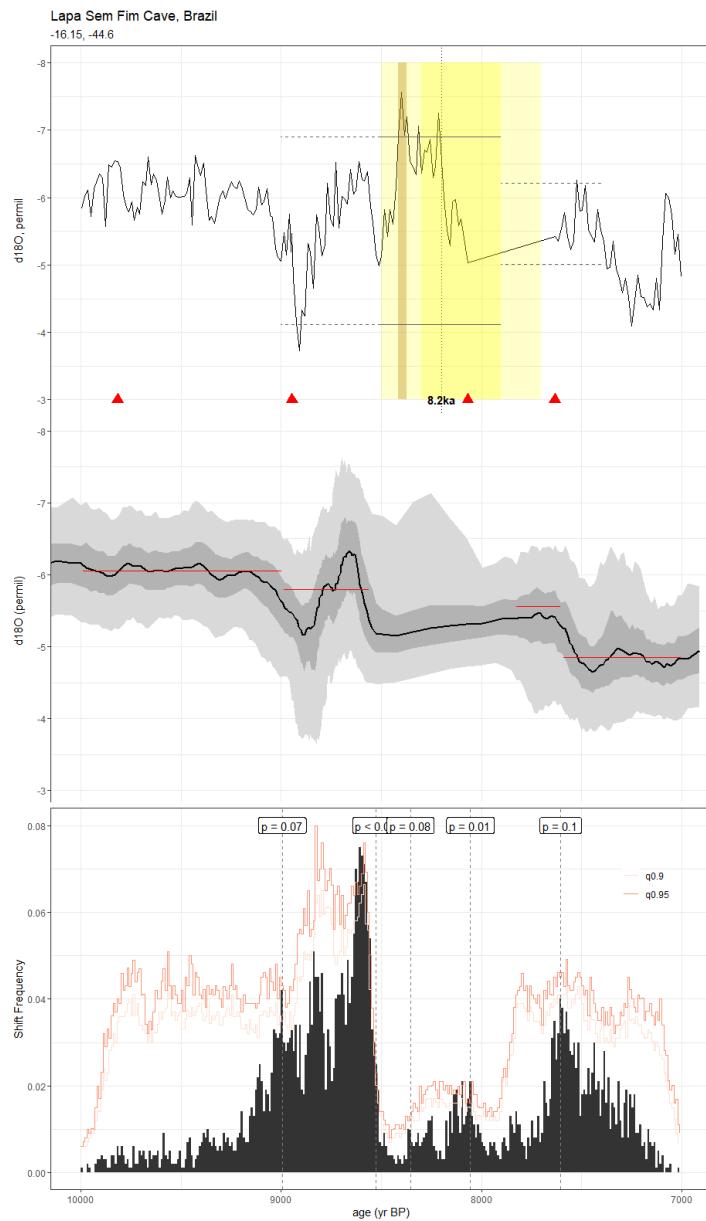
**Figure A33.** As in Fig. A3, but for the speleothem record of Zhang et al., 2013 (LH2). The published age model was generated by linearly interpolating between radiometric dates. Here, we employ the BACON age ensemble included in version 2 of the SISAL database.



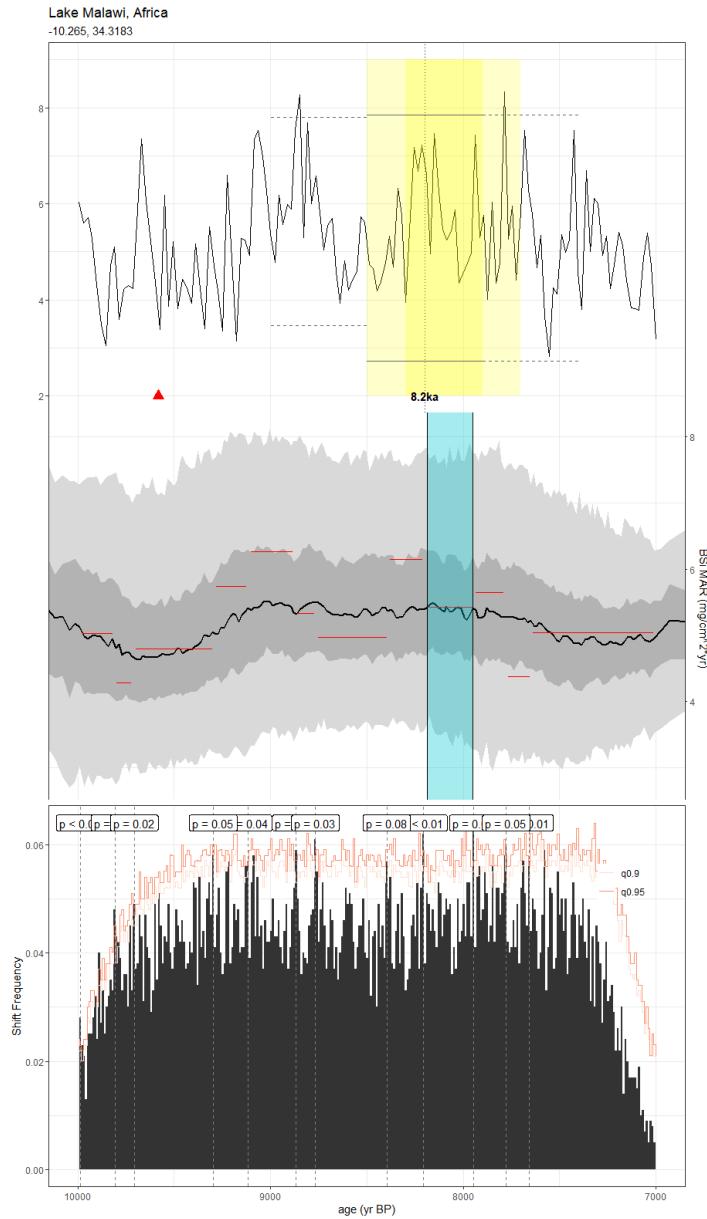
**Figure A34.** As in Fig. A3, but for the lacustrine sediment record of Bird et al., 2011 (LP). The published age model was created using CALIB 5.0 with an unreported calibration curve. Here, we construct our age ensemble in geoChronR using the BACON algorithm and SHCal20.



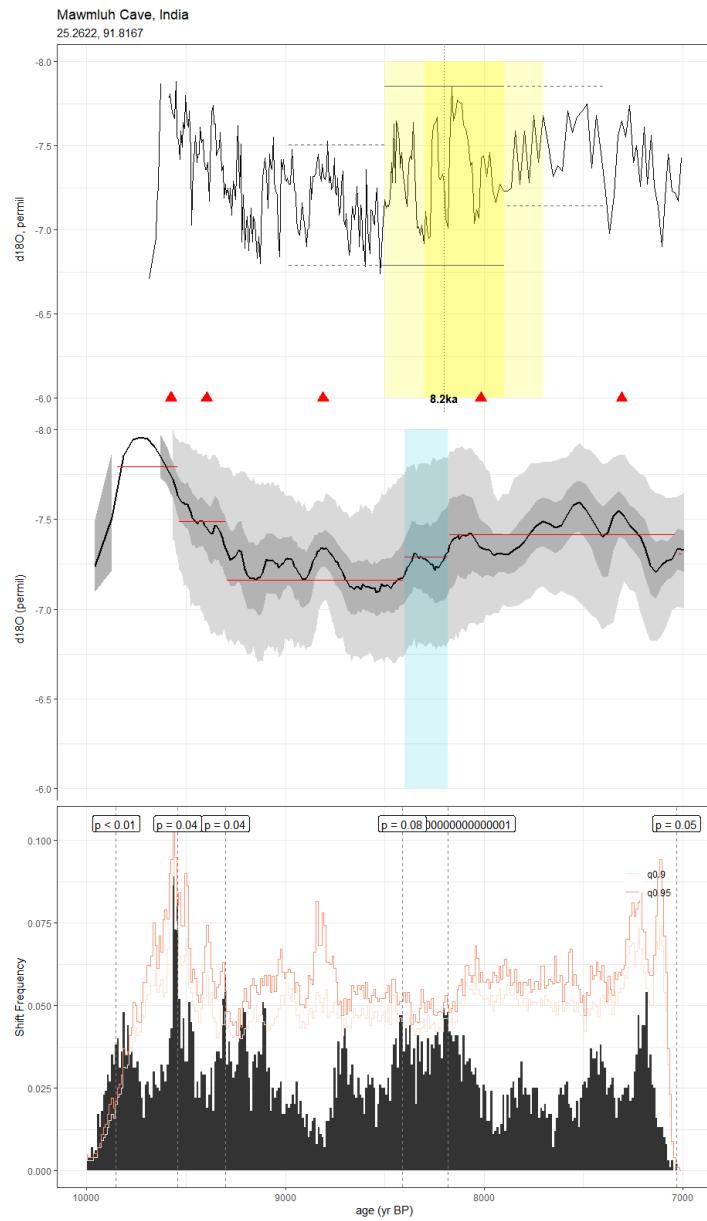
**Figure A35.** As in Fig. A3, but for the speleothem record of Ayliffe et al., 2013 (LR06\_B3\_2013). The published age model was constructed by linearly interpolating between radiometric dates. For our analyses, we leveraged the Bchron ensemble published in the SISALv2 dataset.



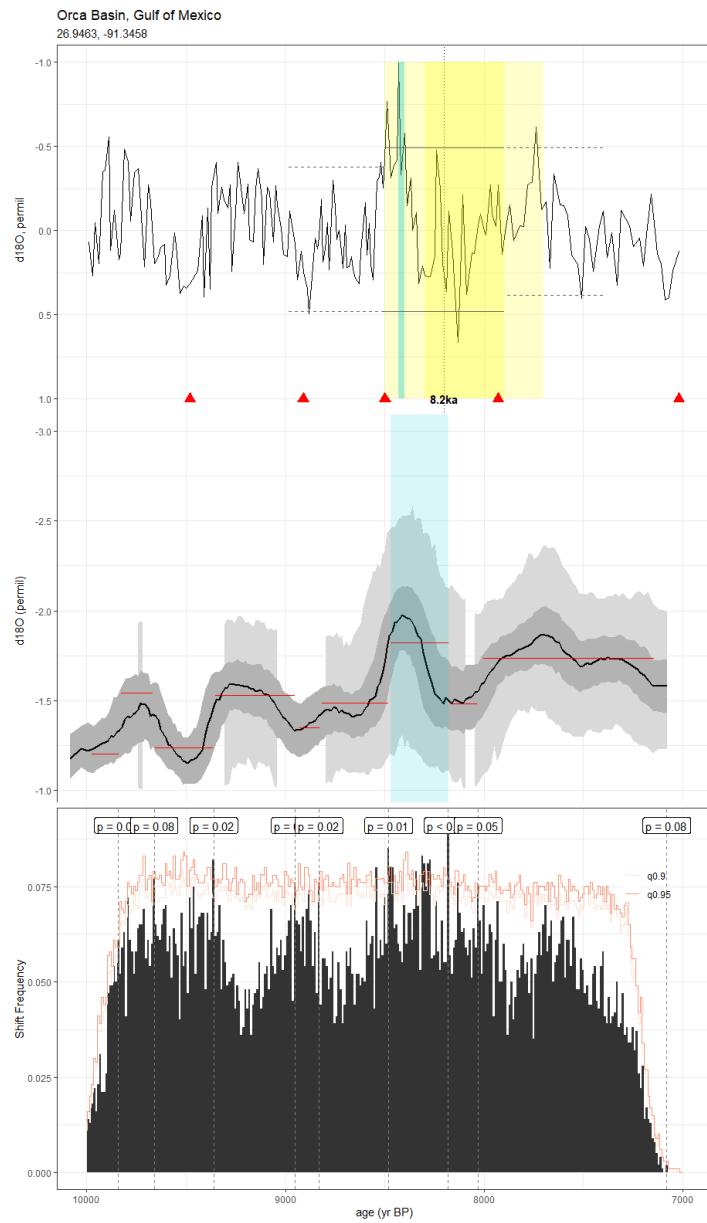
**Figure A36.** As in Fig. A3, but for the speleothem record of Azevedo et al., 2021 (LSF19). The original method used in the construction of the published age model was unreported, but we use the BACON ensemble supplied in version 2 of the SISAL database for our analyses.



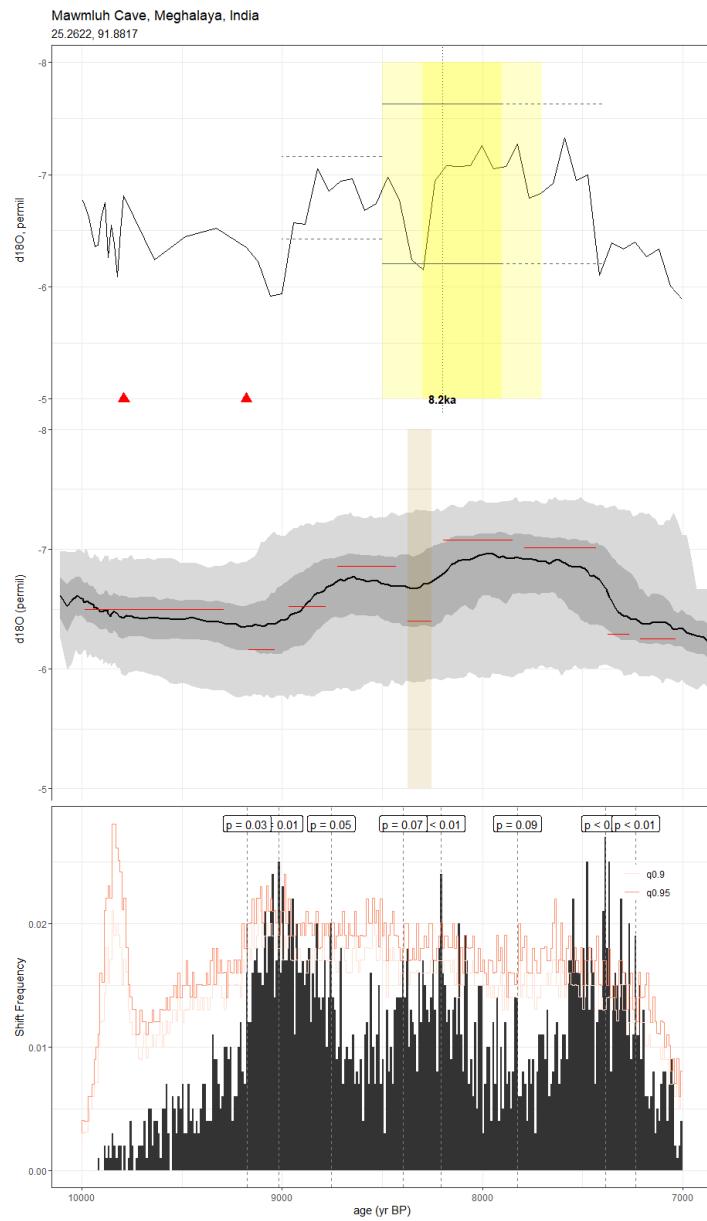
**Figure A37.** As in Fig. A3, but for the lake sediment BSi MAR record of Johnson et al., 2003 (M981P). CALIB 4.3 was used in the construction of the published age model, with a reservoir age correction of -450 years applied to the radiometric dates. Here, we constructed the age ensemble using the BACON algorithm in geoChronR.



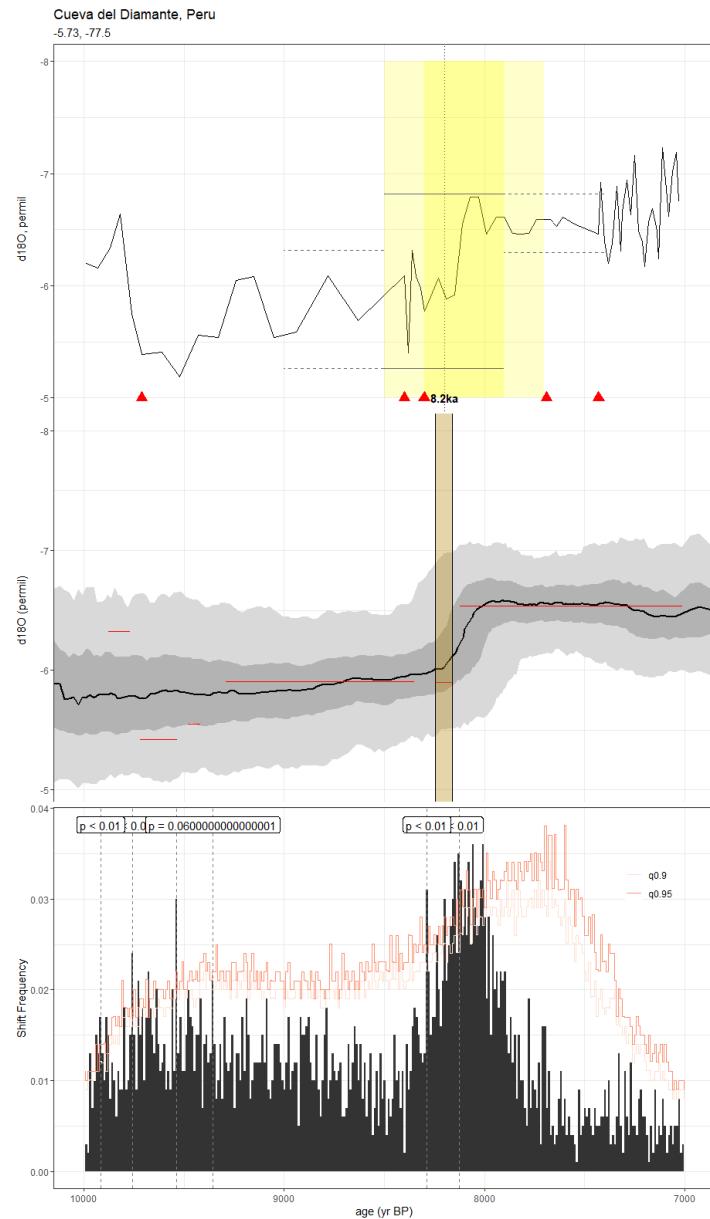
**Figure A38.** As in Fig. A3, but for the speleothem record of Lechleitner et al., 2017 (MAW6). The published age model was constructed using copRa. Here, we employed the Bchron ensemble included in SISALv2.



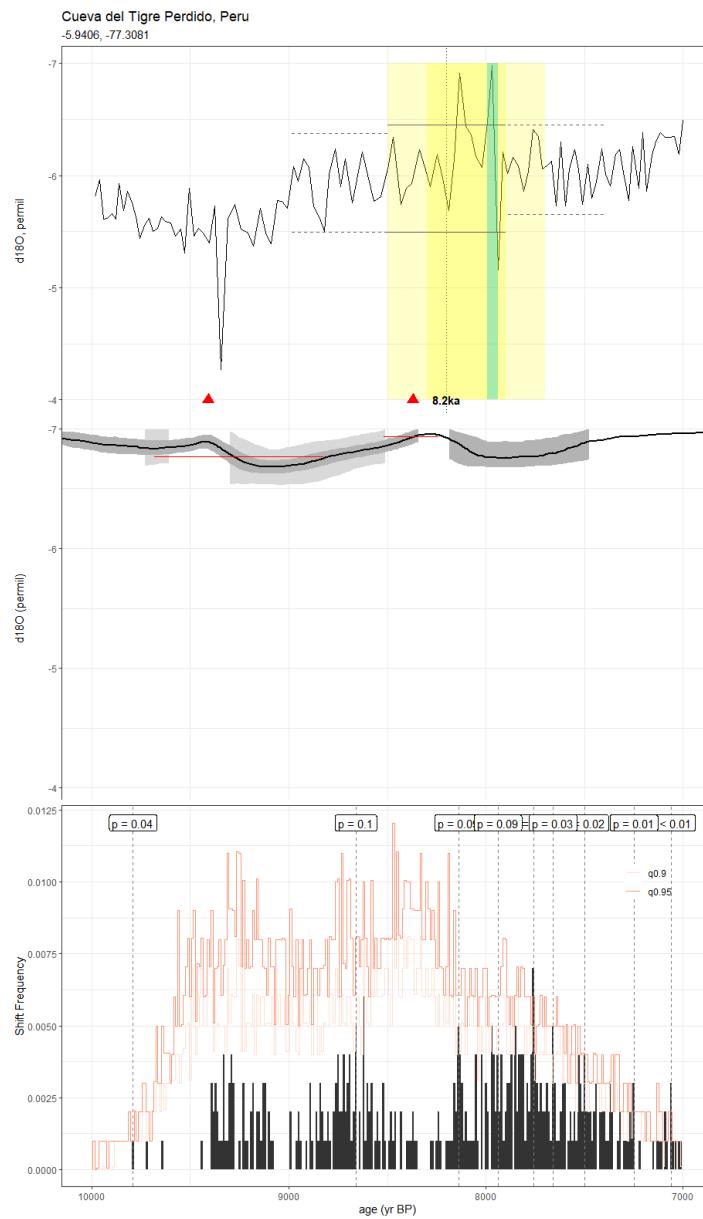
**Figure A39.** As in Fig. A3, but for the foraminifera record of LoDico et al., 2006 (MD022550). The published age model was constructed using CALIB 5.0, with a 400 year reservoir age correction applied. Here, we used the BACON algorithm included in geoChronR to create the age ensemble.



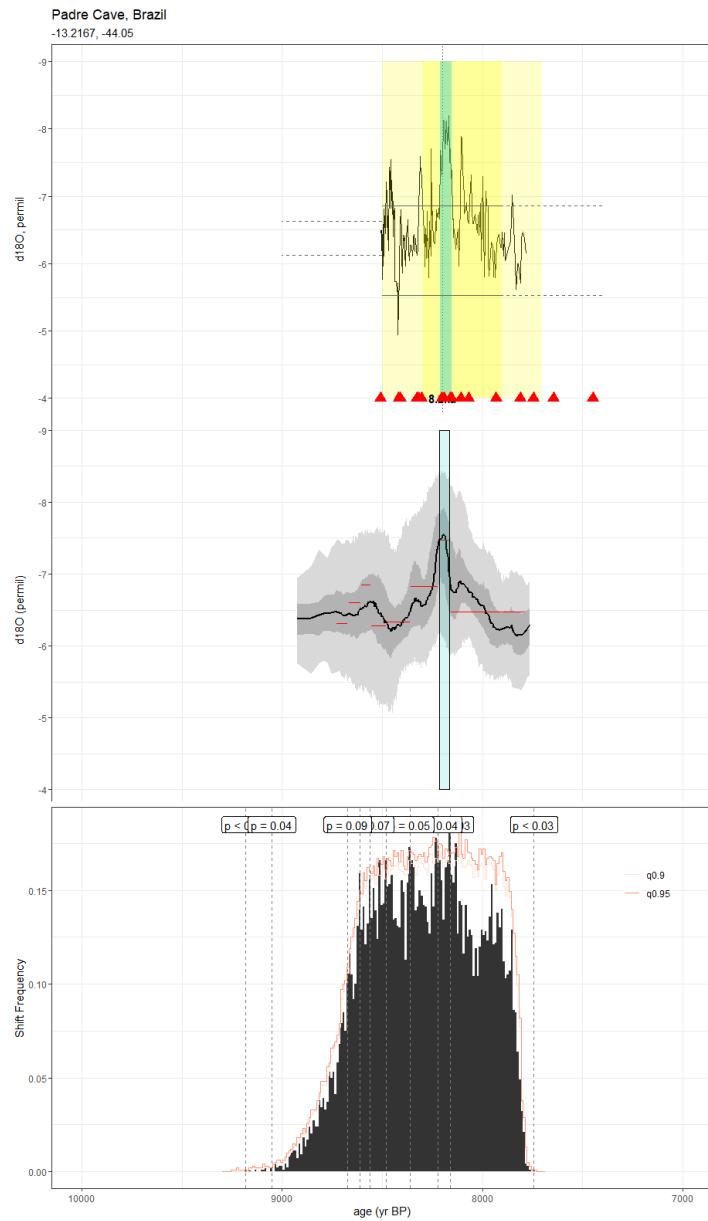
**Figure A40.** As in Fig. A3, but for the speleothem record of Dutt et al., 2015 (MWS1). The published age model was created using the StalAge algorithm. Here, we use the Bchron ensemble from SISALv2.



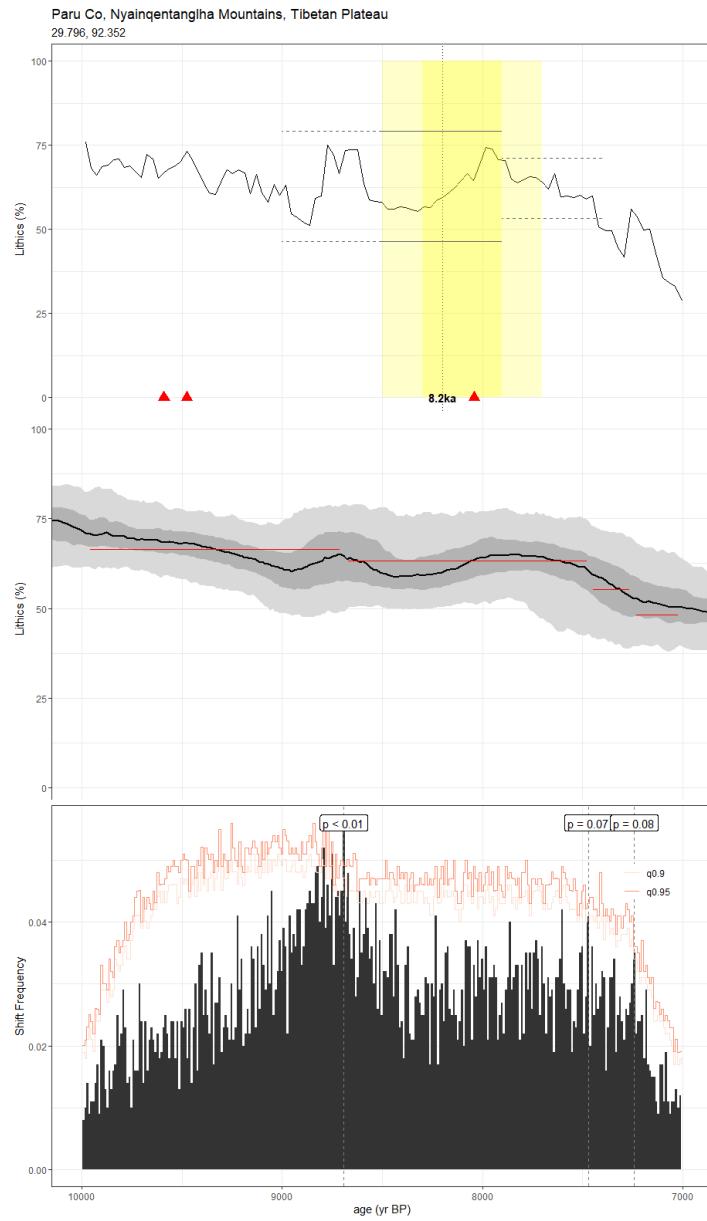
**Figure A41.** As in Fig. A3, but for the speleothem record of Cheng et al., 2013 (NARC). The published age model was constructed by linearly interpolating between radiometric dates, but here, we leverage the copRa ensemble from the SISALv2 dataset.



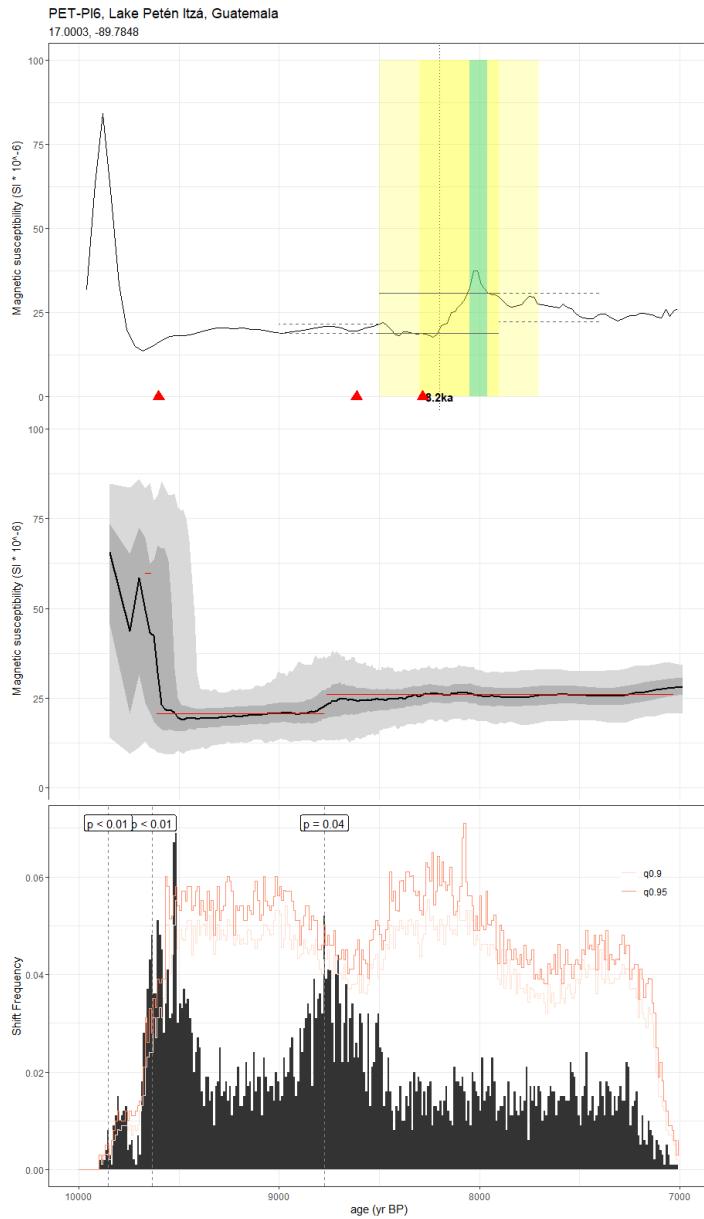
**Figure A42.** As in Fig. A3, but for the speleothem record of van Breukelen et al., 2008 (NCB). Isoplot 3 was used to construct the published age model, however, we use the SISALv2 BACON age ensemble for our analyses.



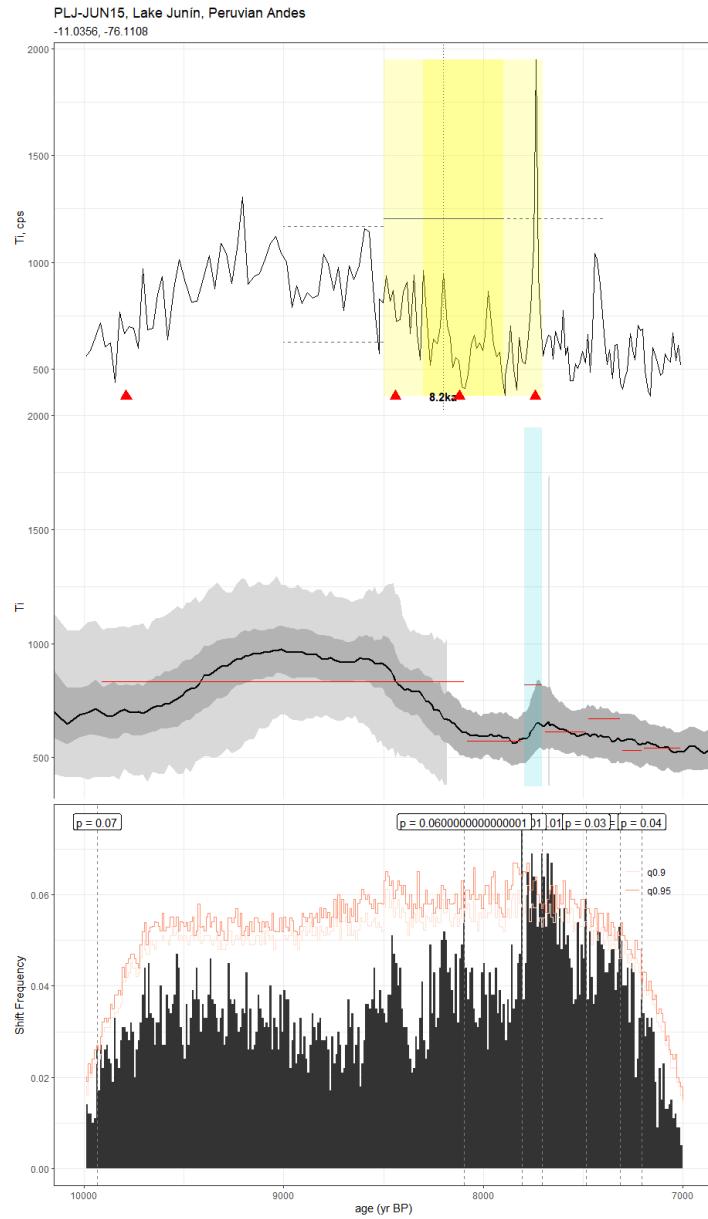
**Figure A43.** As in Fig. A3, but for the speleothem record of Cheng et al., 2009 (PAD07). The original age modeling method used in the construction of the published time series is unknown. Here, we present an age ensemble using the BACON algorithm provided by geoChronR.



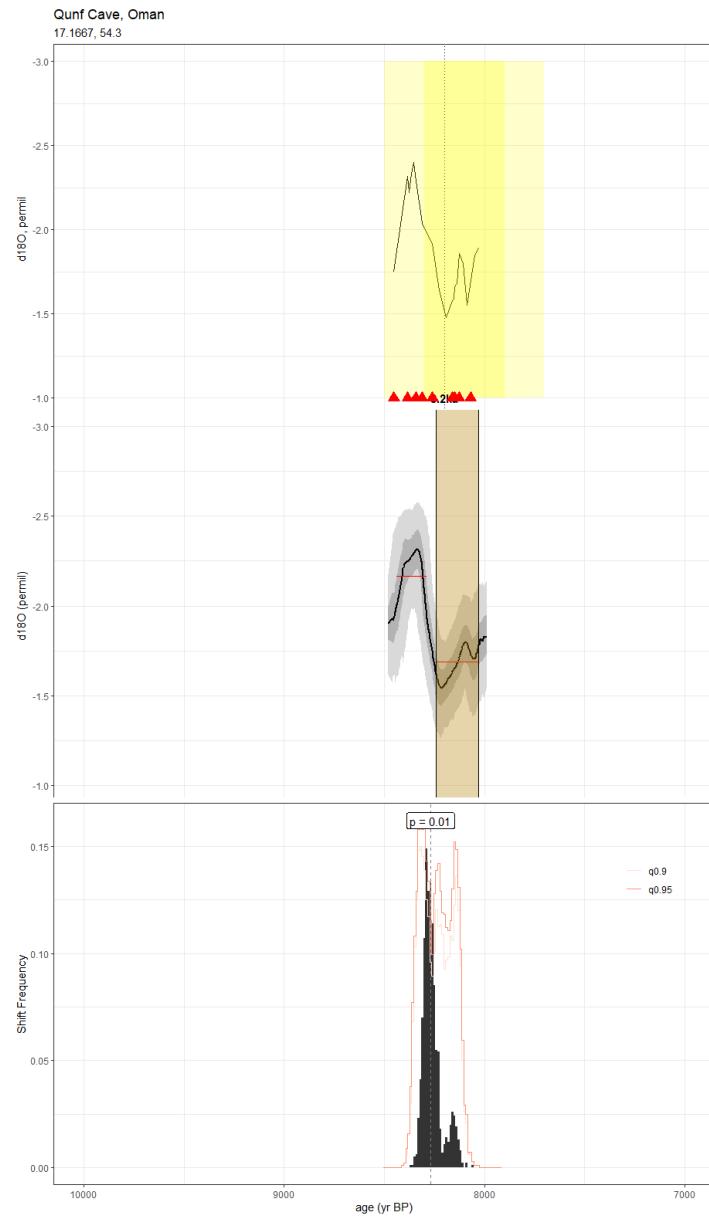
**Figure A44.** As in Fig. A3, but for the lake sediment (percent lithics) record of Bird et al., 2014 (ParuCo). CALIB 6.0 and the IntCal09 calibration curve were used in the construction of the published age model. We construct our age ensemble using BACON and IntCal20 via geoChronR.



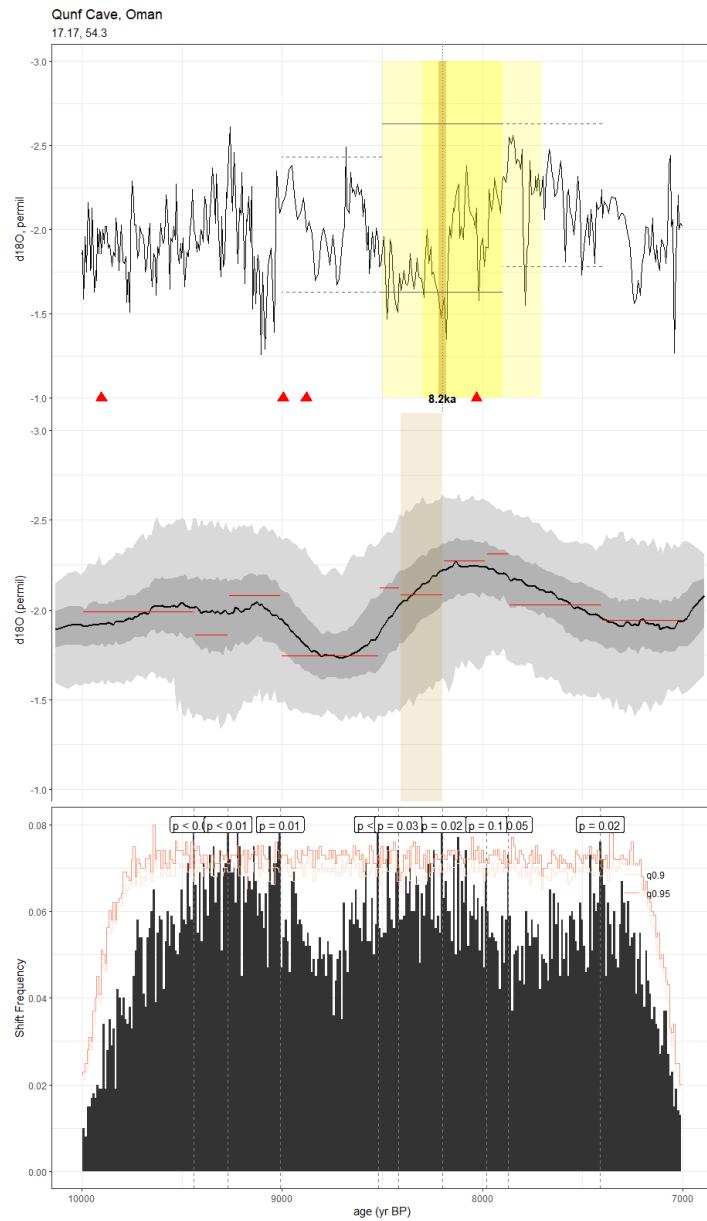
**Figure A45.** As in Fig. A3, but for the lake sediment magnetic susceptibility record of Escobar et al., 2012 (PETPI6). The published age model was generated using the OxCal algorithm with IntCal09 calibration curve. Here, we show an age ensemble created using the BACON algorithm with IntCal20 calibration curve generated by geoChronR.



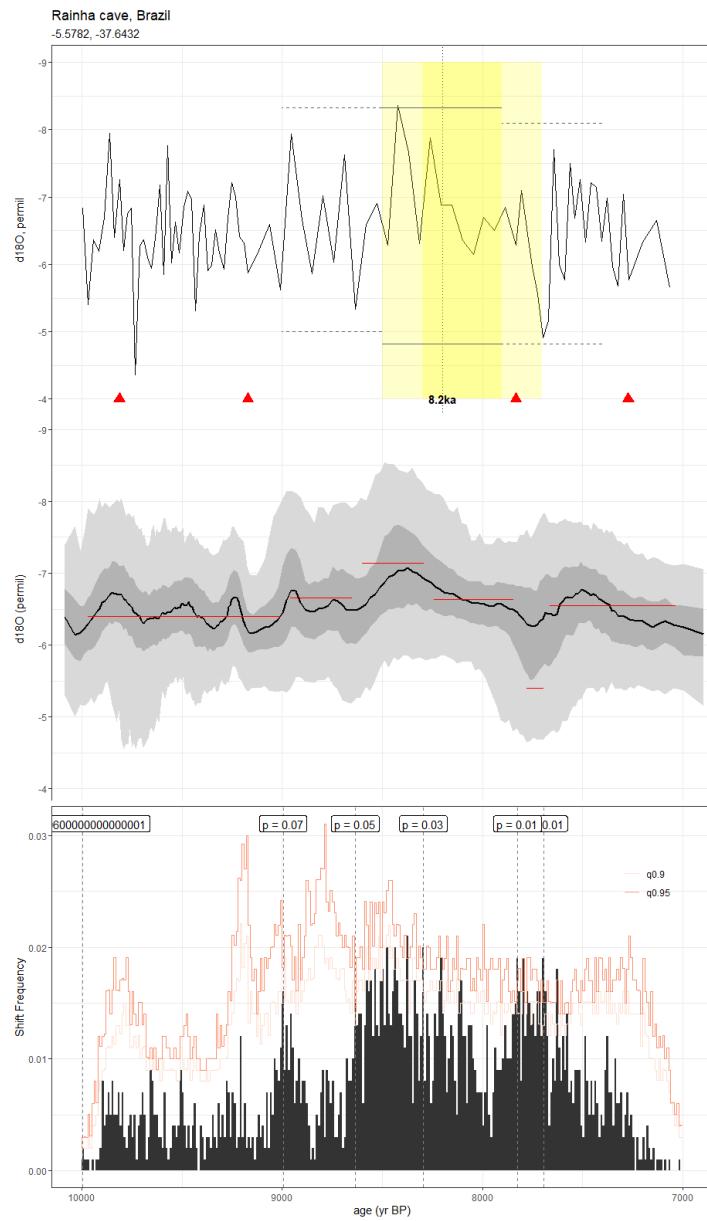
**Figure A46.** As in Fig. A3, but for the lake sediment titanium content record of Woods et al., 2020 (PLJJUN15). The published age model was created using the BACON algorithm with IntCal13 calibration curve. Here, we reconstruct a BACON ensemble using the IntCal20 curve in geoChronR.



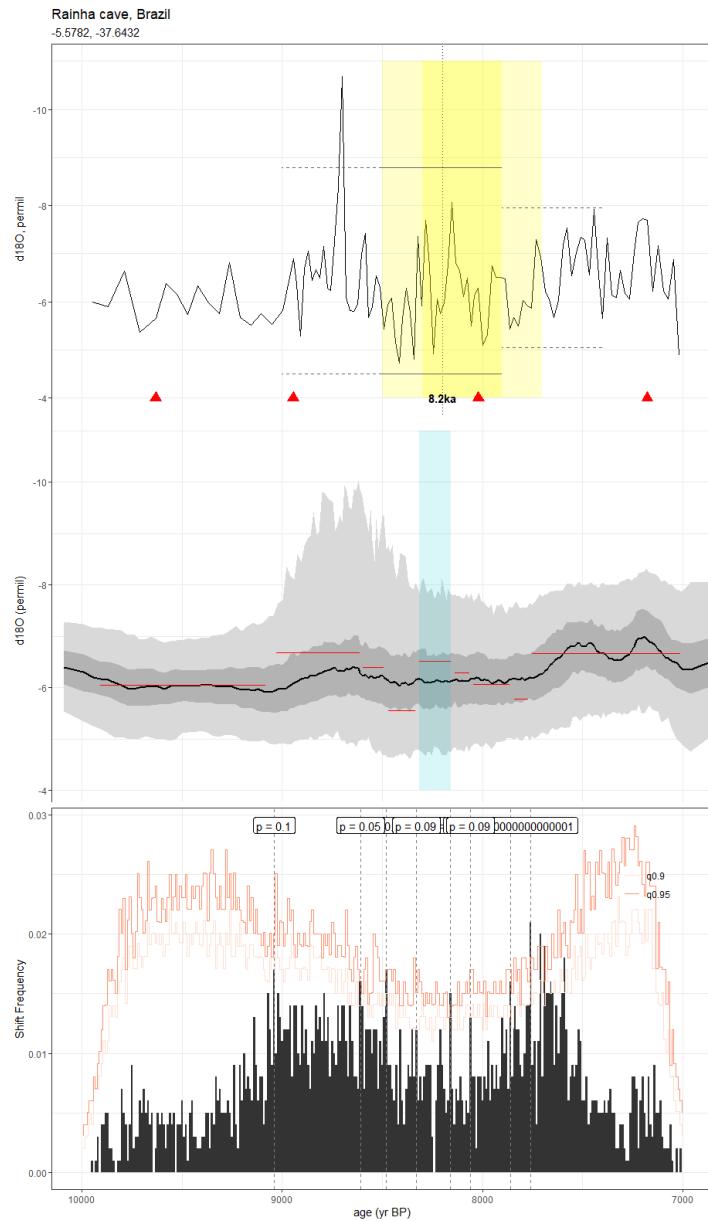
**Figure A47.** As in Fig. A3, but for the speleothem record of Cheng et al., 2009 (Q5Cheng). The method used in the construction of the published age model is unreported; here, we use BACON in geoChronR to generate our age ensemble.



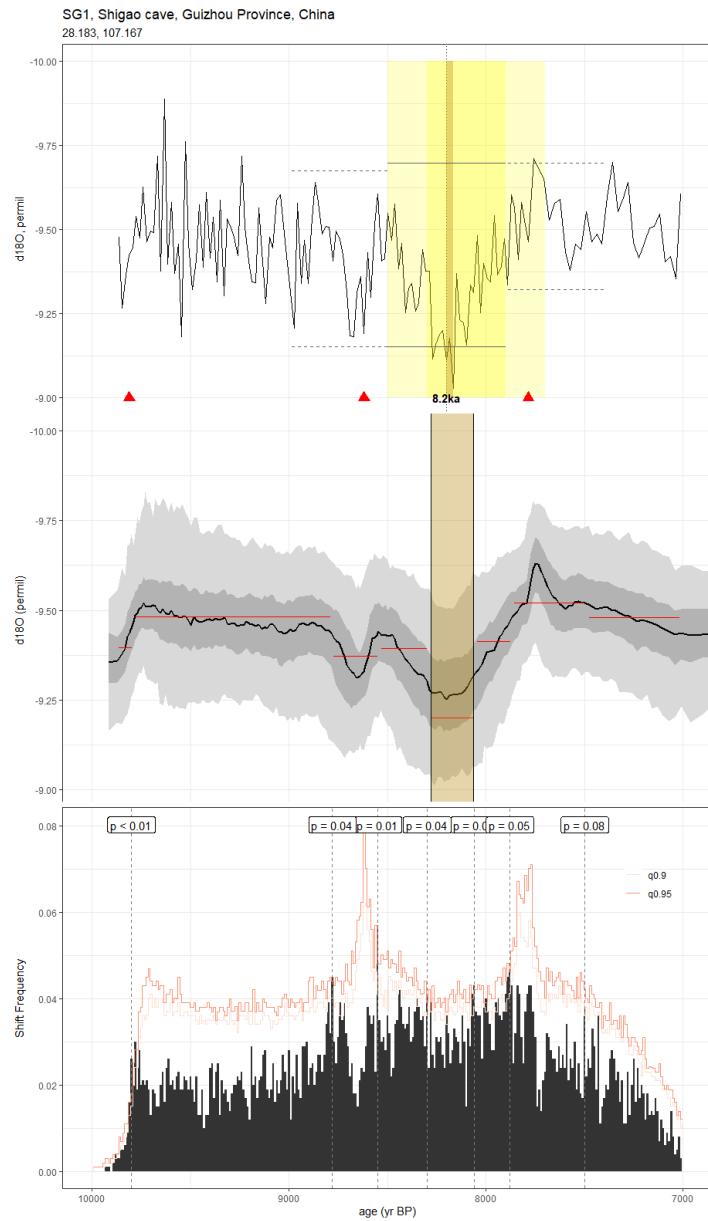
**Figure A48.** As in Fig. A3, but for the speleothem record of Fleitmann et al., 2007 (Q52007). The published age model was created via a polynomial fit to the age-depth curve of the Th–U data. Our age ensemble leverages the BACON algorithm included in geoChronR.



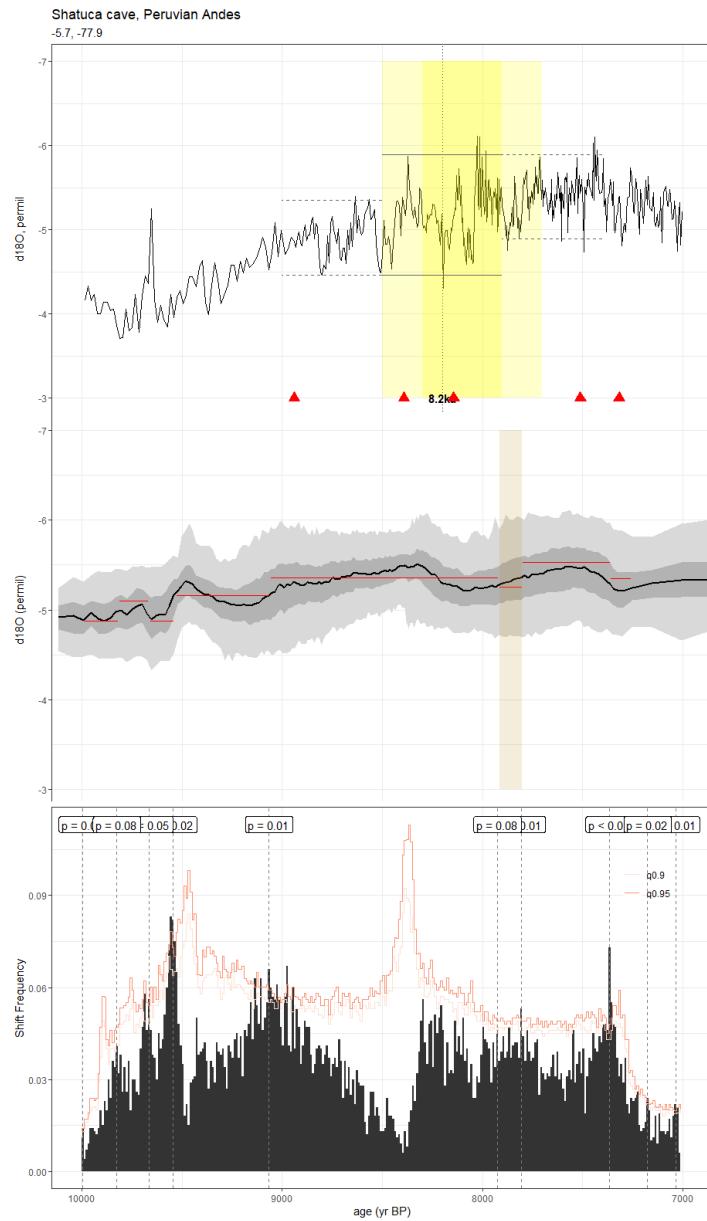
**Figure A49.** As in Fig. A3, but for the speleothem record of Cruz et al., 2009 (RN1). The method used in the construction of the published age model was unreported, but we leverage the Bchron ensemble supplied in version 2 of the SISAL database for our analyses.



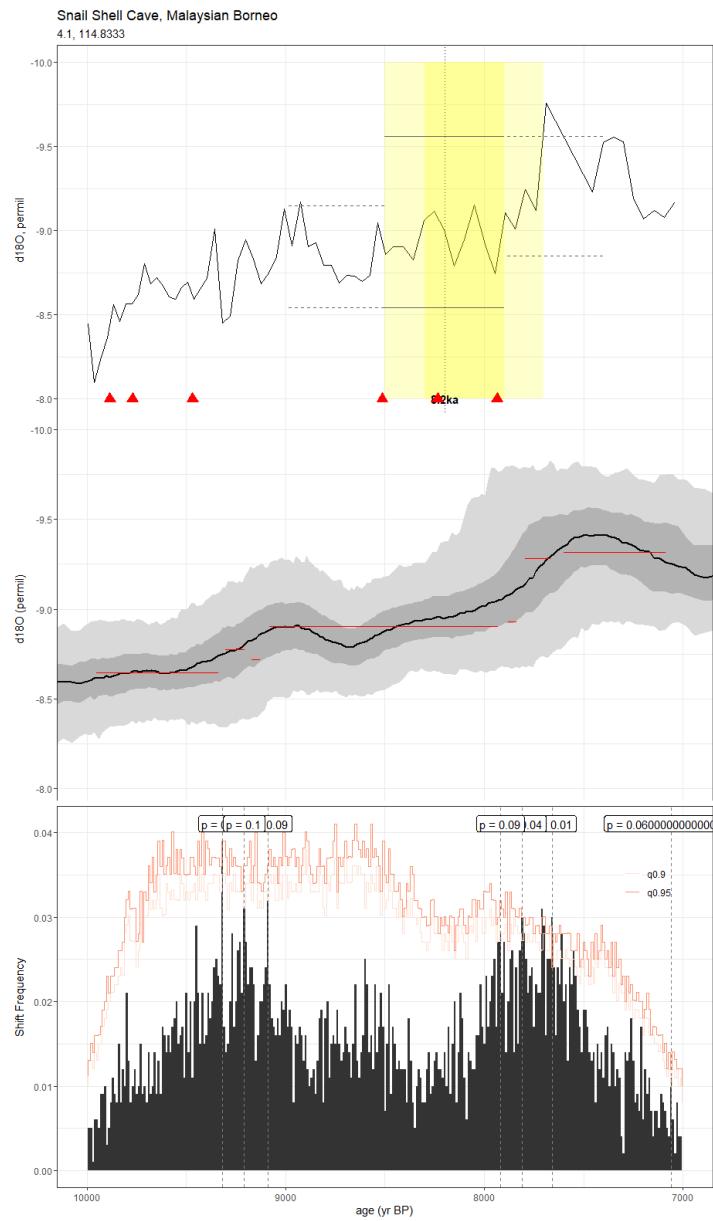
**Figure A50.** As in Fig. A3, but for the speleothem record of Cruz et al., 2009 (RN4). The method used in the construction of the published age model was unreported, but we leverage the Bchron ensemble supplied in version 2 of the SISAL database for our analyses.



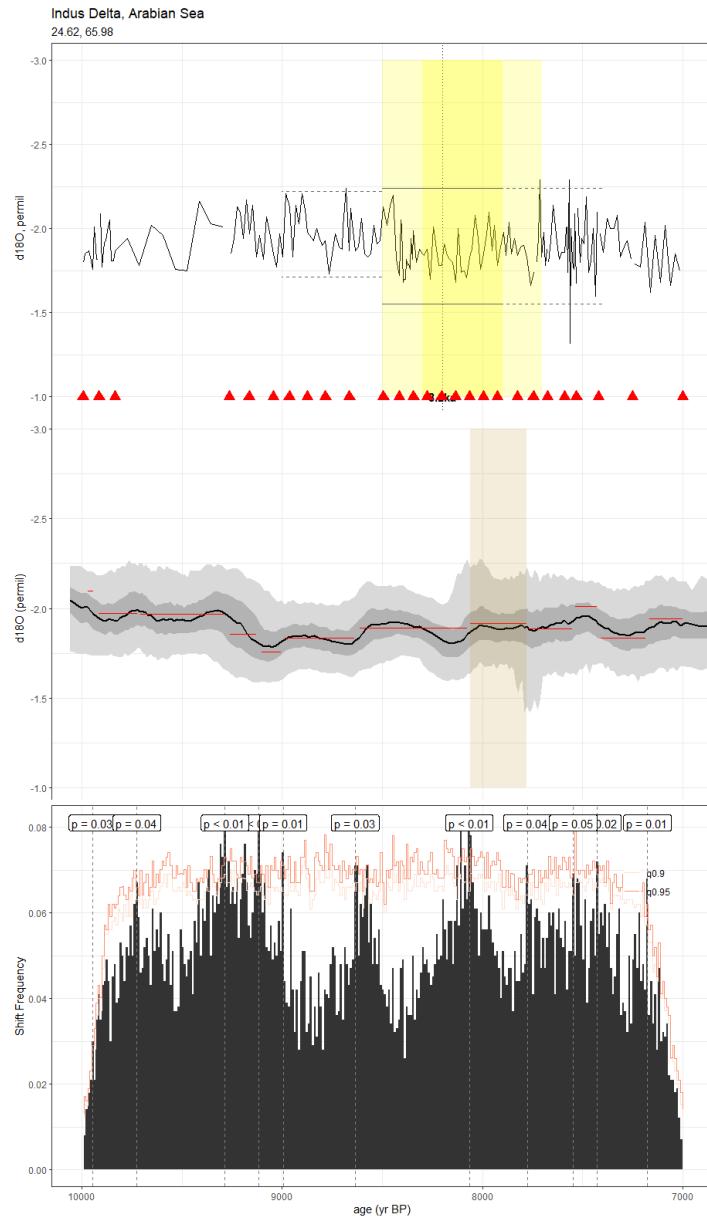
**Figure A51.** As in Fig. A3, but for the speleothem record of Jiang et al., 2012 (SG1). The published age model was constructed by linear interpolation between U/Th dates. Here, we leverage the BACON ensemble from SISALv2 for our analyses.



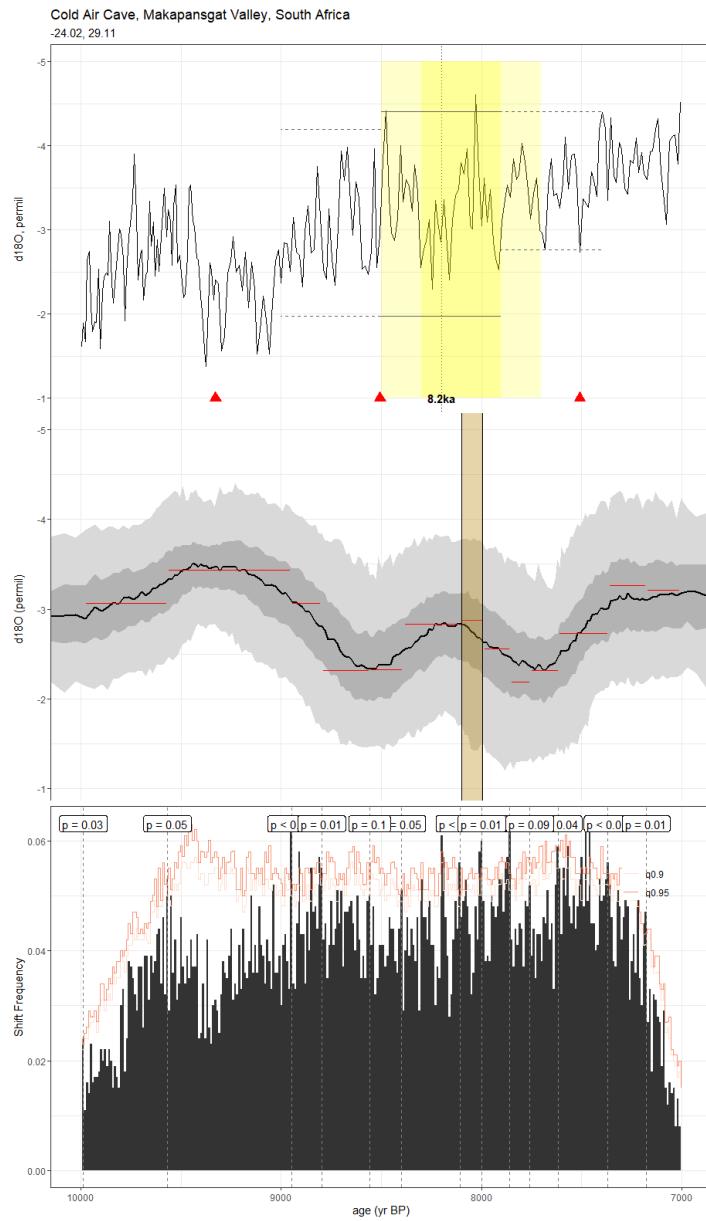
**Figure A52.** As in Fig. A3, but for the speleothem record of Bustamante et al., 2016 (Sha3). The published age model was developed using copRa. Here, we present the Bchron age ensemble generated for SISALv2.



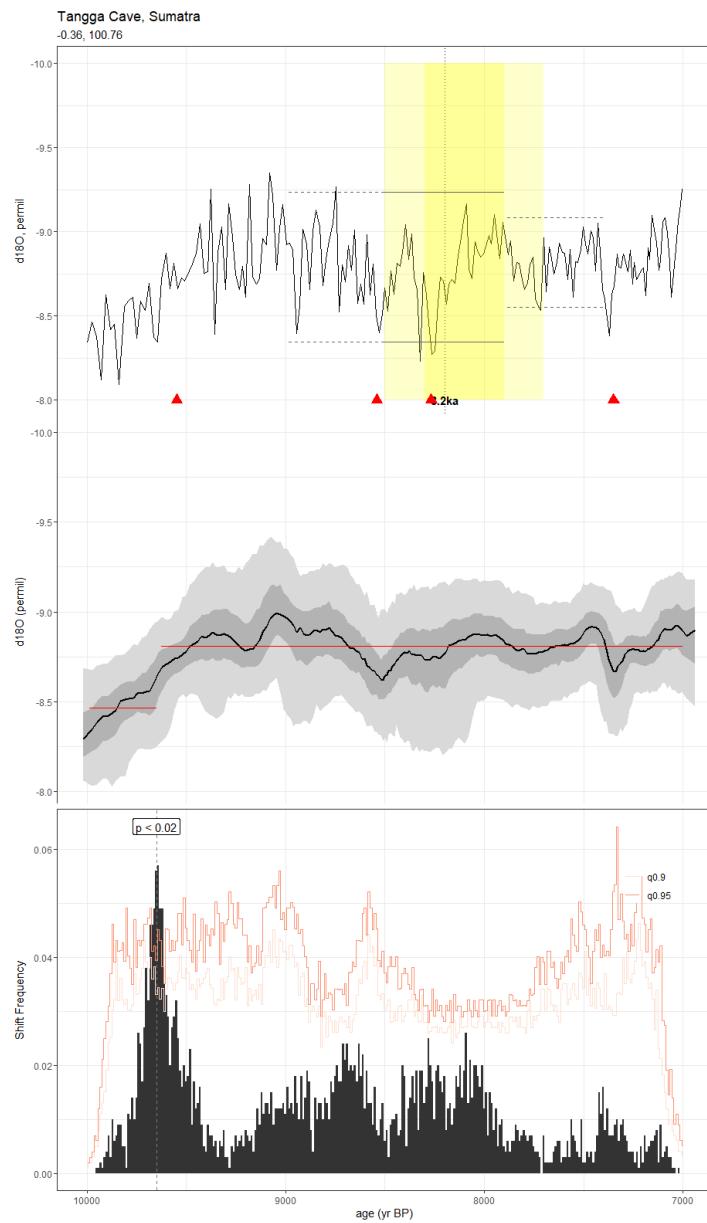
**Figure A53.** As in Fig. A3, but for the speleothem record of Carolin et al., 2016 (SSC01). StalAge was used to construct the published age model. We used the BACON algorithm included in geoChronR to generate our age ensemble.



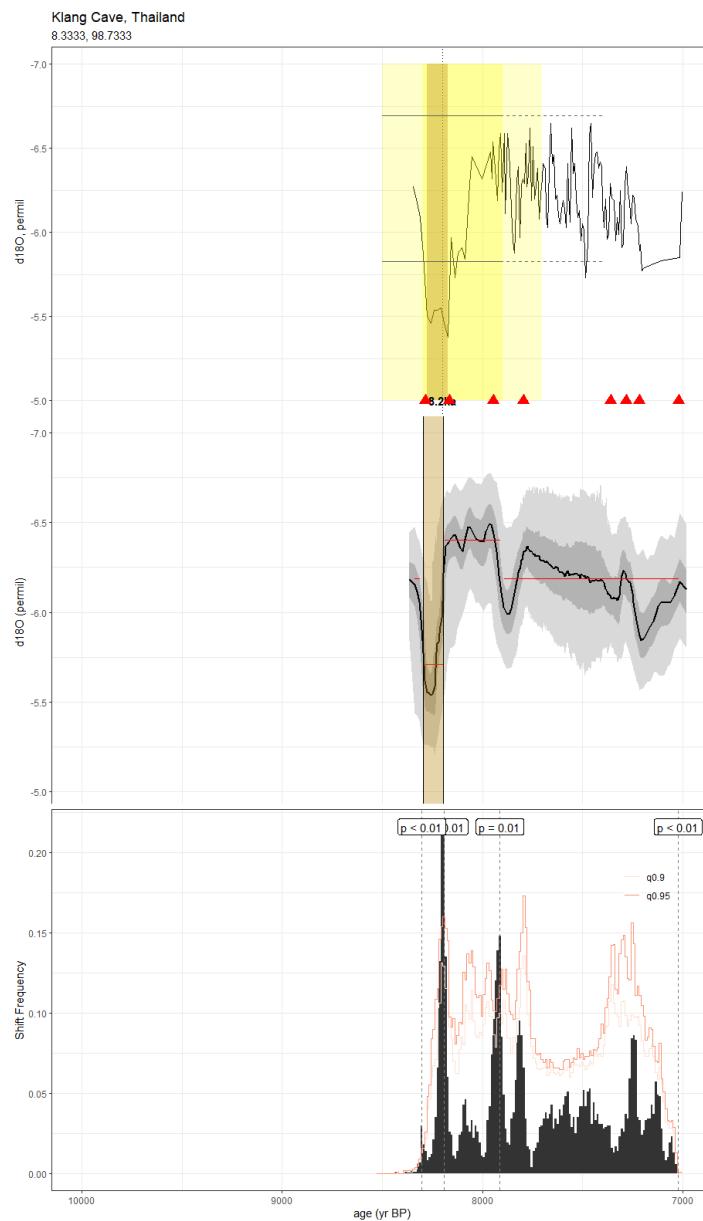
**Figure A54.** As in Fig. A3, but for the foraminifera record of Staubwasser et al., 2003 (Staubwasser63KA). The published age model was generated via a least-squares regression between  $^{14}\text{C}$  dates using the IntCal98 calibration curve. Here, we constructed our age ensemble using the BACON algorithm and IntCal20 calibration curve in geoChronR.



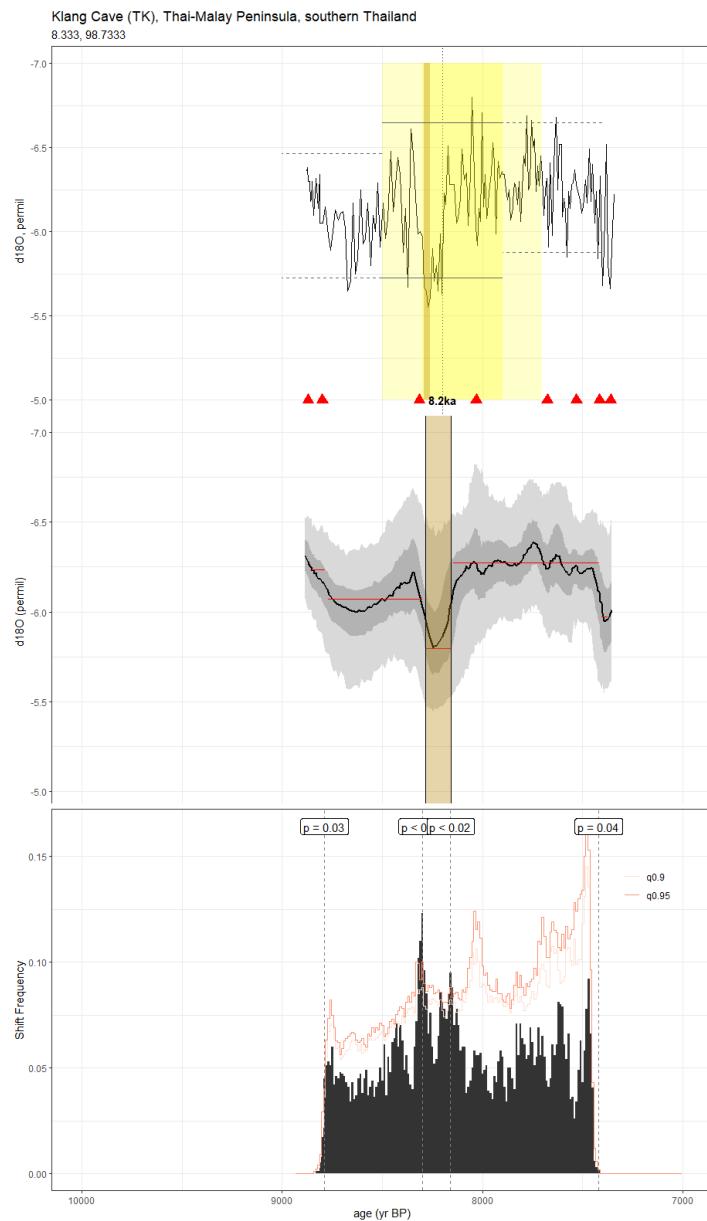
**Figure A55.** As in Fig. A3, but for the speleothem record of Holmgren et al., 2003 (T8). The published age model was constructed via linear interpolation between dates. Here, we construct our ensemble using the BACON age model algorithm in geoChronR.



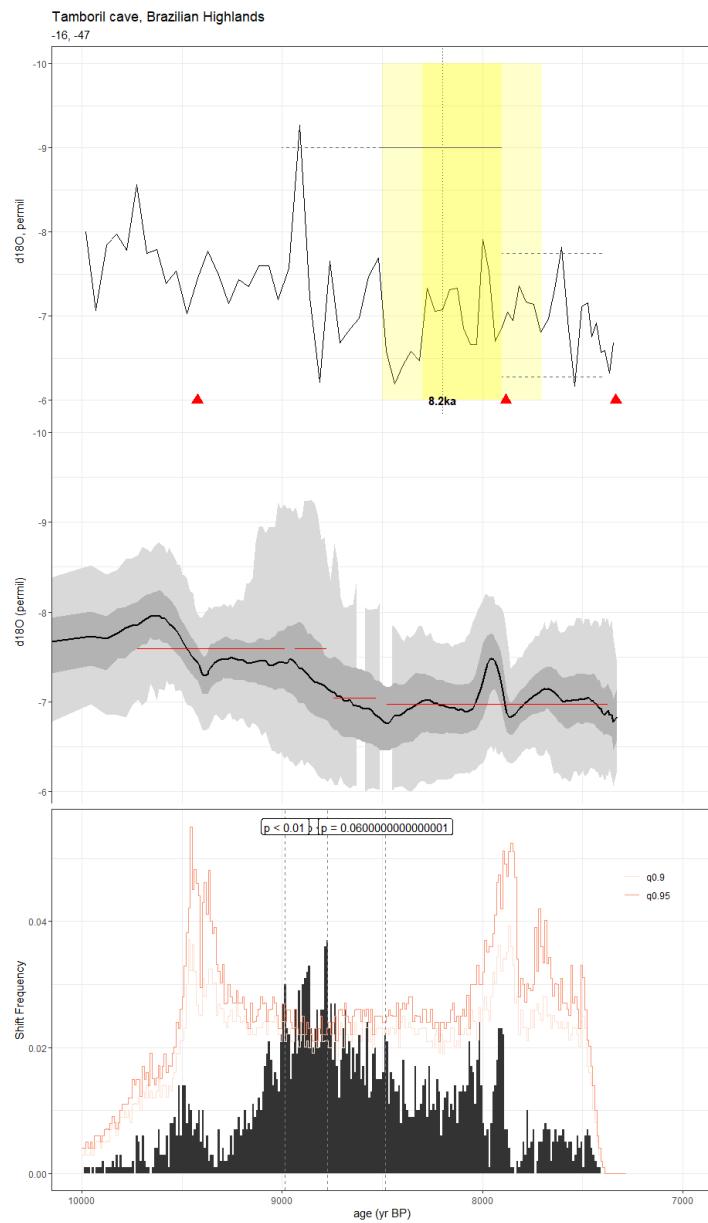
**Figure A56.** As in Fig. A3, but for the speleothem record of Wurtzel et al., 2018 (TA122). The published age model was constructed using the BACON algorithm. Here, we used the copRa ensemble generated for SISALv2.



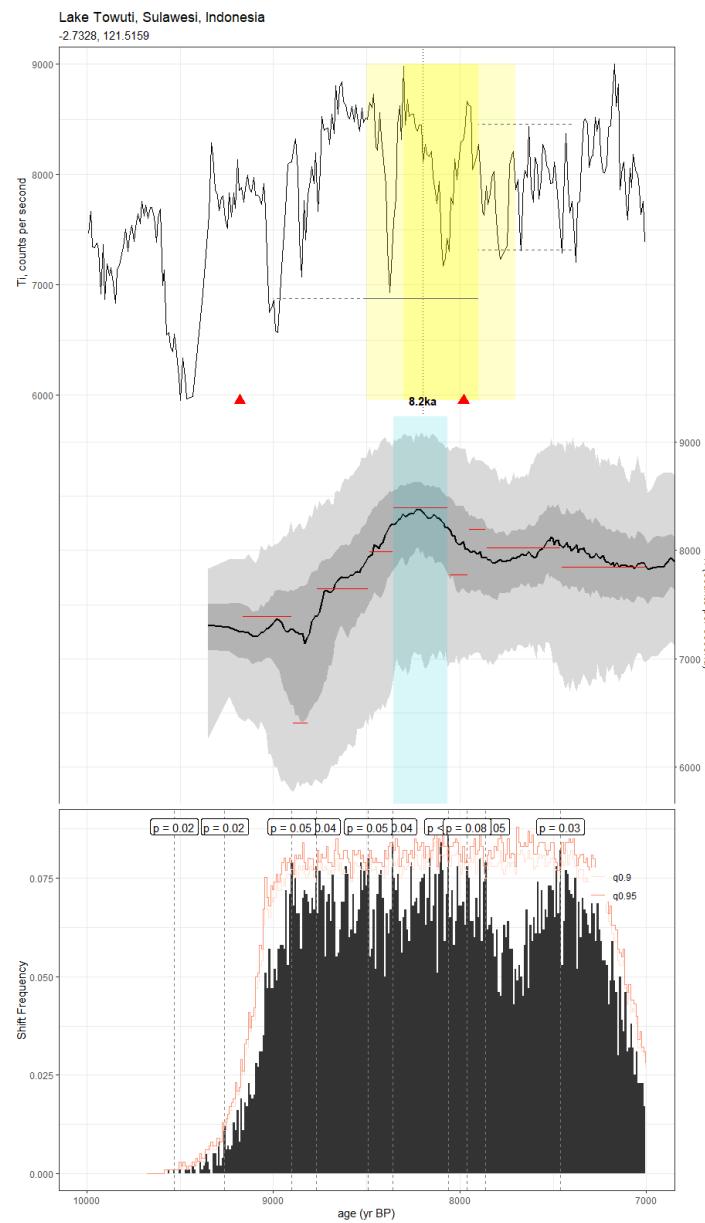
**Figure A57.** As in Fig. A3, but for the speleothem record of Chawchai et al., 2021 (TK07). The published age model was constructed using the BACON algorithm. Here, we used the BACON age ensemble supplied in the SISALv2 database.



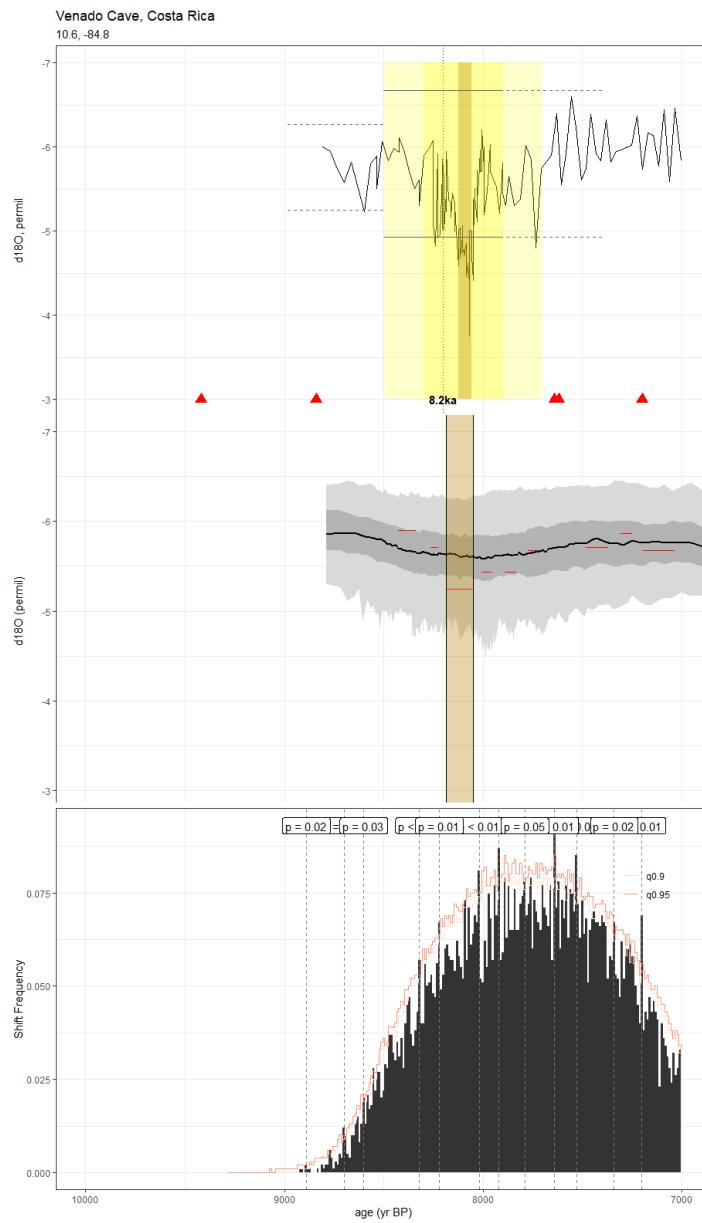
**Figure A58.** As in Fig. A3, but for the speleothem record of Chawchai et al., 2021 (TK20). The published age model was constructed using the BACON algorithm. Here, we used the BACON age ensemble supplied in the SISALv2 database.



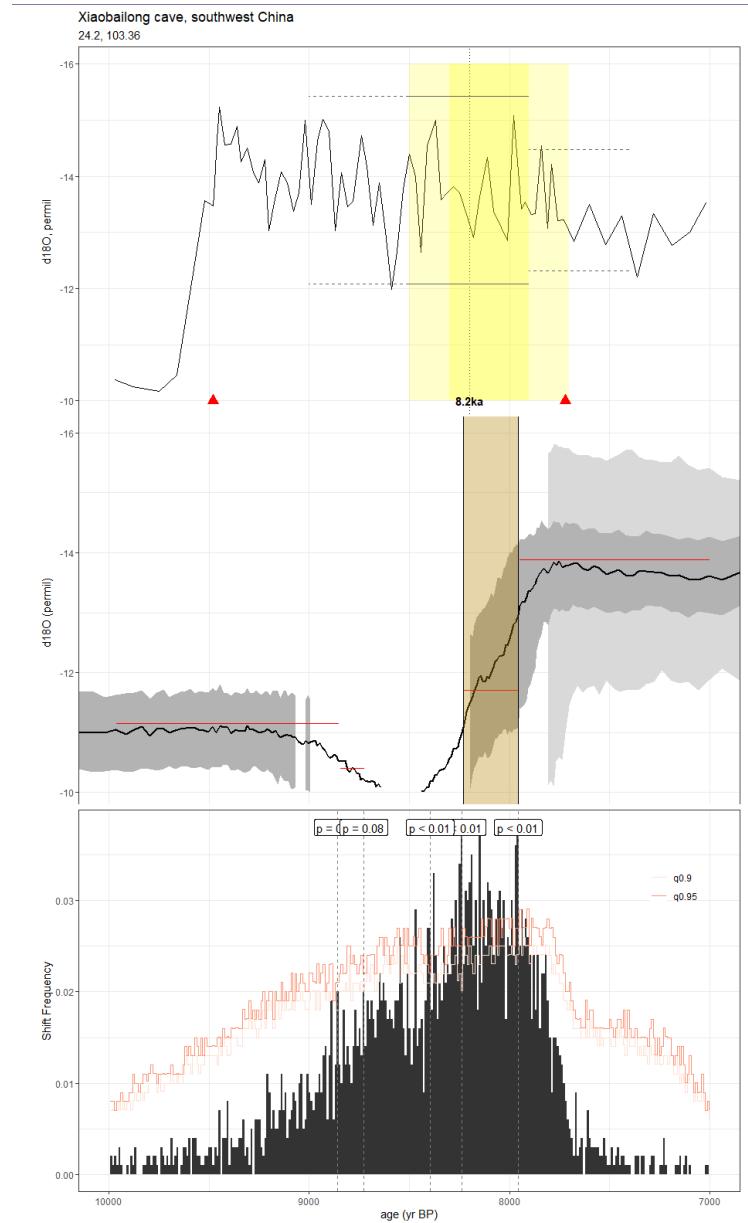
**Figure A59.** As in Fig. A3, but for the speleothem record of Ward et al., 2019 (TM6). The published age model was constructed using the copRa algorithm, though we use the BACON age ensemble supplied in the SISALv2 database for our analyses.



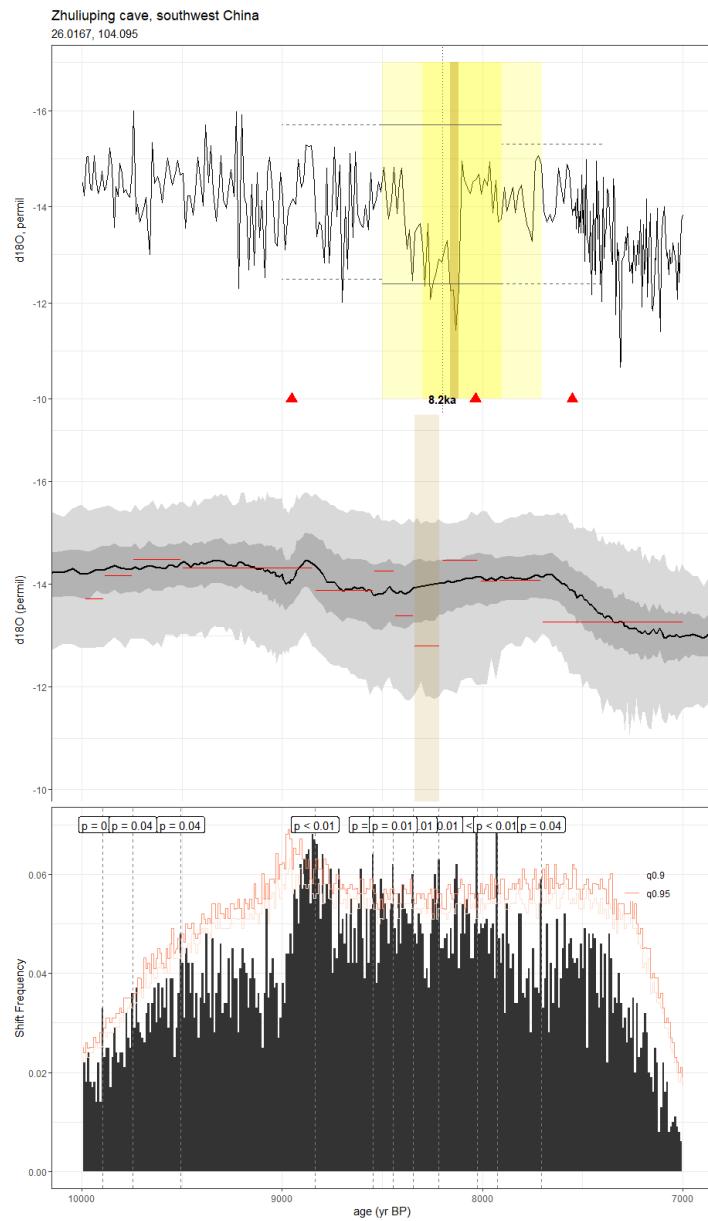
**Figure A60.** As in Fig. A3, but for the lacustrine sediment record of Russell et al., 2014 (TOW109B). CALIB 6.0 was used to construct the published age model, though we leveraged the BACON algorithm included in geoChronR to generate our age ensemble.



**Figure A61.** As in Fig. A3, but for the speleothem record of Lachniet et al., 2004 (V1). The published time series was aligned to a fifth-order polynomial best-fit age model between isochron dates. We employ the BACON ensemble provided by SISALv2 for our analyses.



**Figure A62.** As in Fig. A3, but for the speleothem record of Cai et al., 2015 (XBL29). The published age model was derived from linear interpolation between radiometric dates. For our analyses, we leveraged the SISALv2 Bchron age ensemble.



**Figure A63.** As in Fig. A3, but for the speleothem record of Huang et al., 2016 (ZLP1). The published age model was derived from linear interpolation between radiometric dates. For our analyses, we leveraged the SISALv2 BACON age ensemble.