

Ice sheet modulation of deglacial North American Monsoon intensification

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1 **The North American Monsoon, the dominant source of rainfall for much of the arid US**
2 **Southwest, remains one of the least understood monsoon systems. The late Pleistocene evolu-**
3 **tion of this monsoon is poorly constrained, largely because glacial changes in winter rainfall**
4 **obscure summer monsoon signatures in many regional proxy records. Here, we develop**
5 **deglacial records of monsoon strength from isotopic analyses of leaf wax biomarkers in ma-**
6 **rine sediment cores. Reconstructions indicate a regional decrease in monsoon rainfall during**
7 **the Last Glacial Maximum, and that the deglacial trajectory of the North American Mon-**
8 **soon closely tracks changes in North American ice cover. In climate model simulations, North**
9 **American ice cover shifts the westerlies southward, favoring mixing of cold, dry air into the**
10 **Southwest. This process, known as ‘ventilation’, weakens the monsoon by diluting the energy**
11 **fluxes required for convection. As the ice sheet retreats north, the monsoon strengthens, and**
12 **local ocean conditions may play a larger role in regulating its intensity. We conclude that on**
13 **glacial-interglacial timescales, ice sheet-induced reorganizations of atmospheric circulation**

¹⁴ have a dominant influence on the North American Monsoon.

¹⁵ Motivation

¹⁶ The North American Monsoon (NAM) provides over 50% of annual rainfall in arid regions of
¹⁷ the American Southwest and northwestern Mexico, sustaining a growing human population and
¹⁸ unique regional ecosystems^{1,2}. The NAM is a summertime circulation driven by the development
¹⁹ of a surface thermal low and upper level anti-cyclone over the desert Southwest, which draws in
²⁰ moist air along the Sierra Madre Occidental^{3,4}. The Gulf of California (GoC) and eastern Pacific
²¹ are likely the primary source of moisture in the core region⁵, although the Gulf of Mexico and
²² evapotranspiration may be important secondary sources^{6–9}. Seasonal warming of Gulf of Cal-
²³ ifornia (GoC) sea-surface temperatures (SSTs) are tightly correlated with NAM intensification,
²⁴ suggesting that they may help trigger convection^{3,10,11}. However, remote forcing and changes
²⁵ in the large-scale atmospheric circulation are also important, as intraseasonal breaks in monsoon
²⁶ activity are linked to southward displacements of the westerlies^{4,12,13}.

²⁷ The future trajectory of this monsoon in response to anthropogenic warming remains uncer-
²⁸ tain: climate models predict a delay in the NAM’s onset by the end of the 21st century, but disagree
²⁹ whether increases in late-season rainfall will compensate for early season deficits^{14–16}. This uncer-
³⁰ tainty highlights the need for an improved mechanistic understanding of this monsoon’s response
³¹ to past and present climate change. In particular, it is critical to assess the relative influence of
³² changes to local energy fluxes and the large-scale atmospheric circulation on the long-term history
³³ and variability of the NAM.

34 Paleoclimate histories provide an ideal opportunity to elucidate these mechanisms. For in-
35 stance, the dramatic changes in Earth's boundary conditions associated with Pleistocene glacia-
36 tions may be used to analyze NAM sensitivity to a cooler global climate¹⁷. However, the causes
37 of NAM changes during the Last Glacial Maximum (LGM, 21 ka BP) remain unclear. Previ-
38 ous paleoclimatic research hypothesized that the NAM collapsed during the LGM due to cooler
39 oceans and continental ice sheets¹⁸⁻²⁰. Drawing on observations of modern climate, researchers
40 have suggested a tight link between GoC SSTs and monsoon strength over the Holocene^{21,22}, but
41 it is unclear whether this SST-monsoon connection can explain NAM variability during full glacial
42 conditions. In contrast, several modeling studies suggest that ice sheet-induced displacements of
43 the westerlies during the LGM weakened or eliminated the NAM^{20,23}. This could have resulted
44 from 'ventilation,' or the mixing of cold, dry air into the NAM region²³. Evaluating the relative
45 importance of these oceanic versus atmospheric drivers of NAM strength requires high-resolution
46 records of the monsoon during the Pleistocene-Holocene transition.

47 Unfortunately, large winter rainfall changes almost wholly mask the NAM signal in many
48 deglacial proxy records²⁴⁻²⁶, resulting in disparate descriptions of late Pleistocene NAM behavior.
49 A composite speleothem record from southern Mexico suggests that rainfall is relatively insensi-
50 tive to glacial boundary conditions²⁷, contradicting lacustrine records in northern Mexico that point
51 to a weak monsoon prior to the Holocene^{28,29}. The persistence of monsoon-sensitive taxa during
52 the LGM indicates that the NAM circulation was not entirely suppressed through the last glacial
53 interval³⁰, contradicting hypotheses of NAM collapse at the LGM²⁰. Even among records that
54 show dry LGM conditions, records disagree on the relative magnitude of LGM versus deglacial

55 climate changes³¹. These divergent proxy interpretations highlight the need for quantitative paleo-
56 climatic records that constrain the magnitude of LGM NAM changes and the monsoon's deglacial
57 trajectory.

58 **Leaf wax-based monsoon reconstructions**

59 We generated new reconstructions of NAM hydroclimate using the isotopic composition of leaf
60 wax biomarkers preserved in marine sediment cores from across the modern NAM region (Figure 1). Analysis of a north-south transect of core top samples from the Gulf of California demon-
61 strates that leaf wax-based inferences of δD precipitation closely tracks changes in the proportion
62 of annual rainfall that comes from the monsoon (% July-September rainfall), showing that this
63 proxy method can provide novel constraints on NAM history (Figure 1, see Methods). We devel-
64 oped a Bayesian regression model to capture the strong linear relationship between leaf wax δD
65 and % July-September rainfall (Figure 1), and then applied this regression downcore to develop
66 continuous, quantitative estimates of NAM hydroclimate spanning the last $\sim 20,000$ years.
67

68 Our reconstructions reveal dramatic changes in the monsoon's contribution to regional hy-
69 drology across the Pleistocene-Holocene transition. In the modern climatology, the Guaymas Basin
70 receives approximately 70% of annual rainfall during the modern monsoon season. In contrast,
71 LGM data suggest a starkly different climatology, with the summer monsoon contributing only
72 42% of annual rainfall (Figure 2a). Similar deglacial hydroclimate trends are evident in the core
73 from the Mexican margin, although both records show different trajectories of monsoon change

74 during the Holocene. This site receives 75% of annual rainfall between July and September, but
75 more negative values of leaf wax δD at the LGM indicate that the monsoon only contributed 45-
76 50% of annual rainfall at this time (Figure 2b, Supplementary Figure 1). Most notably, both
77 records suggest that while the NAM was weaker at the LGM, it still comprised greater than 20%
78 of regional rainfall. The leaf wax records also lack significant responses to known abrupt climate
79 change events during the last deglaciation (Heinrich Stadial 1, 17.5-14.5 ka; and the Younger
80 Dryas, 12.8-11.5 ka) (Figure 2a,b). This suggests that the deglacial NAM was not sensitive to
81 Gulf of Mexico moisture fluxes, since the latter region cooled significantly during the Younger
82 Dryas³². It also contrasts with oceanic proxies that indicate abrupt deglacial changes in upwelling
83 intensity and temperature in the Guaymas Basin^{21,33,34}, and some continental records of southwest
84 hydroclimate^{25,28}.

85 Compilations of proxy evidence from the western US show that changes in westerly-driven
86 storms increased winter rainfall during the LGM, which could account for the large seasonality
87 change in our reconstructions²⁶. We performed a sensitivity test to identify the magnitude of winter
88 rainfall changes needed to fully explain the LGM change in % July-September contribution (i.e.
89 assuming that the monsoon did not change and cool season (Nov-May) precipitation was the sole
90 driver) (Figure 3). At the Guaymas Basin, winter precipitation would have to increase between
91 200% and 430% to explain observed LGM proxy values; at the Mexican margin site it would
92 have to increase 450–1420%. Such large-magnitude increases in winter rainfall are unrealistic,
93 especially given that Clausius-Clapeyron scaling suggests that the cooler glacial atmosphere would
94 have held less water vapor. In addition, a suite of fully-coupled GCM simulations of the LGM

95 from the PMIP3 archive suggests that LGM winter precipitation was at most 180% greater than
96 pre-industrial values (Figure 3b)¹⁷. The inferred % July-September changes from our biomarker
97 data therefore require a substantial decrease in the strength of the summer monsoon during the
98 LGM.

99 **Proxy-model comparisons**

100 To evaluate the causal mechanisms responsible for monsoon changes, we plotted our % July-
101 September rainfall reconstructions against the temporal evolution of the Laurentide Ice Sheet³⁵
102 (Figure 2, 4). We also compared our Guaymas Basin % July-September record to SST reconstruc-
103 tions from the same site³⁴. Deglacial SST data does not yet exist for the Mexican Margin or other
104 regions of the GoC. % monsoon rainfall increases in step with the northward retreat of ice until
105 the Laurentide's southern latitude crosses 55°N in both proxy records (Figure 4a). This relation-
106 ship is statistically significant ($p < 0.05$, Supplementary Table 3) and can be described with a
107 second order polynomial, with a steeper rate of increase in % monsoon rainfall at the Guaymas
108 Basin than at the Mexican margin. The Guaymas site may be more sensitive to glacial climate
109 changes because it sits farther north within the NAM region (Supplementary Table 3). We also an-
110 alyzed the deglacial dynamics of the monsoon in a series of 'timeslice' simulations conducted with
111 the HadCM3 climate model^{36,37}. In these simulations, % July-September rainfall shows a similar
112 polynomial relationship with the extent of the Laurentide ice sheet as observed in our proxy recon-
113 structions (Figure 4b). The modeled relationship is driven by changes in monsoonal rainfall, and
114 shows little correlation with winter rainfall anomalies (Supplementary Figure 2), reinforcing our

115 interpretation that the leaf wax records ultimately reflect changes in NAM circulation.

116 The relationship between monsoon rain and SST at the Guaymas Basin changes sign over
117 the deglaciation. Prior to 15 ka, monsoon strength is negatively correlated with SST (Figure 4c,
118 Supplementary Table 3). Between 22 and 15 ka, the warmest SSTs coincide with the lowest %
119 monsoon values. After 15 ka, monsoon rainfall covaries positively with SSTs, plateauing at mod-
120 ern values once SSTs are greater than 22°C. The latter relationship can be described by a quadratic
121 polynomial (Figure 4c, Supplementary Table 3), and supports interpretations that GoC SSTs must
122 surpass a certain threshold to sustain a strong NAM circulation^{11,21}. However, the decoupling of
123 ocean temperature and the monsoon between 22 and 15 ka suggests that some other factor must
124 be responsible for the steady increase in monsoon strength in this interval. We suggest this may
125 result from changes to the large-scale environment that render it unfavorable for NAM convec-
126 tion, independent of relatively warm SSTs. HadCM3 simulations fail to capture this ‘threshold’
127 relationship between SSTs and the NAM (Figure 4d). This may reflect the fact that many GCMs
128 lack the resolution to explicitly simulate the GoC, which biases GCM predictions of the NAM’s
129 response to warming¹⁴. However, both models and proxies simulate a steady increase in monsoon
130 strength over the deglaciation, suggesting that this bias does not play a significant role in simula-
131 tions of deglacial climate. In sum, the proxy data suggest a close, consistent relationship between
132 the NAM and ice volume during full glacial conditions, with SSTs playing a role only after 15 ka.
133 Models support this proxy interpretation, showing a first-order influence of ice volume on changes
134 in NAM strength across the entire deglaciation.

135 **Influence of mid-latitude circulations on the glacial NAM**

136 The modeled relationship between NAM intensity and ice extent results from changes in the mid-
137 latitude westerlies. The high albedo of the Laurentide Ice Sheet creates a strong meridional gra-
138 dient of temperature, which drives a southward shift in the subtropical jet based on thermal wind
139 balance²³. Across the timeslice simulations, the strength of westerly wind anomalies over the
140 NAM region tracks changes in the southernmost latitude of ice over North America (Supplemen-
141 tary Figure 3a). Stronger westerly winds are associated with a large-scale environment that is less
142 favorable to monsoonal convection, as indicated by the negative correlation between moist static
143 energy and zonal wind strength over the NAM region (Supplementary Figure 3b). Decreases in
144 moist static energy result from ventilation, or the import of low-energy air into the NAM region by
145 the westerlies as well as the mixing influence of transient eddies²³. These processes act synergisti-
146 cally, since a southward shift in the mid-latitude westerlies generates baroclinic instability, which
147 would promote the growth of transient eddies²³.

148 Our work provides novel quantitative constraints on the glacial NAM circulation. The new
149 reconstructions help resolve disagreement between evidence of a strong monsoon at the LGM³⁰
150 and studies that hypothesized the total collapse of the NAM²⁰ by showing that the circulation
151 was much weaker but still present. HadCM3 simulations indicate that extensive continental ice
152 sheets regulate the NAM by strengthening and shifting the mid-latitude westerlies, favoring the
153 mixing of low energy air into the monsoon region. This ‘ventilation effect’ inhibits convection and
154 decreases the strength of the NAM. As continental ice disintegrates, the westerlies weaken and shift

155 northward, allowing the monsoon to intensify. This mechanism explains the strong relationship
156 between the NAM and ice cover observed in both proxy data and model simulations. Local SSTs
157 may have played a role in NAM strengthening towards the end of Termination 1 but they cannot
158 explain NAM changes prior to 15 ka. The primacy of ice sheet influence distinguishes the NAM
159 from the Indian and west African monsoons, where abrupt deglacial SST changes are dominant
160 drivers of monsoon responses^{38,39}. This strong influence of ventilation on the NAM may be tied
161 to its geographical setting; the lack of topographic barriers against the intrusion of mid-latitude
162 circulations, as well as the monsoon's proximity to cold, dry air over the northeast Pacific⁴ may
163 create a strong sensitivity to changes in mid-latitude atmospheric circulation²³.

164 Our results support the conclusions of new research that link mid-latitude circulations to
165 the dynamics of monsoon regimes. Paleoclimate model simulations show that ventilation may
166 regulate the poleward extent of the west African monsoon⁴⁰. In addition, Cenozoic changes in Hi-
167 malayan topography may have strengthened the south Asian monsoon by blocking ventilation by
168 dry air from mid-latitude deserts^{41,42}. Proxy-model syntheses have revealed a tight link between
169 east Asian monsoon variability and the seasonal cycle of the westerly jet⁴³ as a result of enhanced
170 energy advection downstream of Tibet⁴⁴. Similarly, both our proxy reconstructions and model sim-
171 ulations support a strong role for mid-latitude circulations in modulating North American monsoon
172 behavior on glacial-interglacial timescales. Together, this body of evidence suggests that predict-
173 ing monsoon responses to past and future climate change requires careful consideration of how
174 changes in the large-scale atmospheric flow may alter the energetic environment for convection.

175 **Methods**

176 **Age Models.** The composite record from the Guaymas Basin is based on three sediment cores:
177 MD02-2515, JPC-56, and MD02-2517. MD02-2515 and JPC-56 have published age models based
178 on planktic foraminifera and bulk organic carbon^{33,46}. Our age model updates previous efforts by
179 using the most recent marine calibration curve (Marine13) and a reservoir age correction based
180 on ref.⁴⁷ (Supplementary Figure 4). The age model for MD02-2517 is based on eight radiocarbon
181 dates on benthic and planktic foraminifera, to which we applied published benthic-planktic and
182 reservoir age corrections^{46,47} (Supplementary Table 1). The age model for NH-8P, the sediment
183 core from the Mexican margin, is based on published radiocarbon dates on bulk organic matter⁴⁸.
184 For all sites, we used the Bayesian age modeling program BACON to explicitly model sedimen-
185 tation rates and quantify age uncertainty (Supplementary Figure 4)⁴⁹. Median offsets between
186 previously published and revised age models are 100-200 years.

187 **Leaf Wax Analyses.** Lipids were extracted from freeze-dried and homogenized sediments using
188 an accelerated solvent extractor system (ASE, Dionex 350), at a temperature and pressure of 100°C
189 and 1500 psi, respectively. The total lipid extract (TLE) was evaporated under a steady stream of N₂
190 gas. We separated fatty acids from other lipid compounds using an aminopropylsilyl gel column,
191 eluting the neutral fraction with a dichloromethane:isopropanol (2:1) and the acid fraction with
192 4% acetic acid in dichloromethane. We methylated the acids in a solution of acidified methanol
193 (50°C, overnight). The resultant FAMEs (fatty acid methyl esters) were purified again over silica
194 gel using dichloromethane. We focused subsequent analyses on the C₃₀ fatty acid, since it is
195 exclusively derived from terrestrial plants⁵⁰.

196 FAME concentrations were determined using a GC-FID system. Hydrogen and carbon iso-
197 topic composition were measured via gas chromatography-isotope ratio-mass spectrometry (GC-
198 IR-MS) using a Thermo Delta V Plus mass spectrometer. Reference H₂ and CO₂ gases were cali-
199 brated to an *n*-alkane standard (A6 mix provided by Arndt Schimmelmann at Indiana University),
200 and a synthetic mix of FAMEs was analyzed every 5-7 samples to monitor drift. Samples were
201 run in quadruplicate for δ D to obtain a precision better than 2‰, and triplicate for $\delta^{13}\text{C}$ to obtain
202 a precision better than 0.2‰. To account for the added methyl group during methylation, the δ D
203 and $\delta^{13}\text{C}$ of the methylation methanol was determined by methylating a phthalic acid standard of
204 known isotopic composition obtained from Arndt Schimmelmann at Indiana University. A mass
205 balance correction was then applied to the δ D and $\delta^{13}\text{C}$ values of our FAMEs. Down-core mea-
206 surements of δ D were corrected for ice volume changes during glacial intervals. Benthic oxygen
207 isotope data was used as a proxy for ice volume⁵¹, and was then scaled assuming that 1‰ of the
208 increase in $\delta^{18}\text{O}$ (8‰ in δ D) at the LGM is due to ice alone⁵². The correction is therefore:

$$\delta D_{corrected} = \frac{1000 + \delta D_{initial}}{((8 * \delta^{18}\text{O}_{scaled})/1000) + 1} - 1000 \quad (1)$$

209

210

211 **NAM Reconstructions.** We used modern coretop data from the Gulf of California to develop a
212 quantitative relationship between rainfall seasonality and leaf wax δ D. Sediment trap studies sug-
213 gest that waxes in near-shore marine settings primarily reflect local vegetation⁵³. We first converted
214 δD_{wax} values to estimates of δD_{precip} . While leaf wax hydrogen isotopes closely track the isotopic

215 composition of environmental water and therefore δD_{precip} , this relationship can be confounded by
 216 variability in the apparent offset between δD_{wax} and δD_{precip} , or $\varepsilon_{water-wax}$, across different plant
 217 taxa. In particular, waxes synthesized by graminoids are more depleted relative to δD_{precip} than
 218 those produced by eudicots⁵⁰, perhaps as a result of variations in the timing of seasonal leaf wax
 219 production or in the intermediate hydrocarbon compounds used in leaf wax synthesis⁵⁴. Limited
 220 measurements of ε values are available for members of Cactaceae, which use the Crassulaic Acid
 221 Metabolism (CAM), but this family may not be a major contributor to sedimentary leaf waxes.
 222 Carbon isotopes track changing proportions of graminoids vs. eudicots on the landscape, since
 223 herbs and shrubs in the NAM region primarily use the C₃ photosynthetic pathway, while most
 224 grasses are C₄ taxa⁵⁰. In turn, C₃ and C₄ photosynthetic pathways result in different values of leaf
 225 wax $\delta^{13}\text{C}$, with a more enriched carbon isotopic signature in C₄ taxa.

226 We used paired carbon isotopes to identify the proportion of leaf waxes that come from C₄
 227 grasses vs. eudicots, and then applied appropriate $\varepsilon_{water-wax}$ offsets to infer δD_{precip} from δD_{wax} .
 228 In this approach, we represent ε values as proportional to the fraction of C₄ (f_{C4}) taxa in a given
 229 sample of sedimentary leaf waxes (Eq. 2). This ε value is then used to adjust δD_{wax} values to
 230 obtain δD_{precip} (Eq. 3)⁵⁵.

$$\varepsilon = f_{C4} \cdot \varepsilon_{C4} + (1 - f_{C4}) \cdot \varepsilon_{C3} \quad (2)$$

$$\delta D_{precip} = \frac{1000 + \delta D_{wax}}{(\varepsilon/1000) + 1} - 1000 \quad (3)$$

231 Equation 2 requires us to identify a probability distribution of values of f_{C4} given a certain
232 number of C_4 leaf waxes in each leaf-wax sample (Y), or $P(f_{C4}|Y)$. Y can be inferred from $\delta^{13}\text{C}$
233 data, using equation 4, where N is the number of leaf waxes in a sample (assumed to be a large
234 number).

$$Y = \left(\frac{\delta^{13}\text{C}_{\text{wax}} - \delta^{13}\text{C}_{C3}}{\delta^{13}\text{C}_{C4} - \delta^{13}\text{C}_{C3}} \right) \cdot N \quad (4)$$

235 Inference proceeds in a Bayesian framework following ref. ⁵⁵. End-member $\delta^{13}\text{C}$ values

236 were obtained from our own measurements of Sonoran desert taxa at the Arizona-Sonora Desert
237 Museum (ASDM), near Tucson, AZ (Supplementary Table 2). Leaf samples from Sonoran desert
238 species were gathered by student participants in the ASDM's Junior Docents program on May 17
239 and June 25, 2016. During each sampling effort, several leaves from separate specimens were
240 collected and homogenized to average across individual variability. Upon transport to the lab,
241 samples were freeze-dried and cut into 0.4 g samples. n -acids were extracted and measured
242 following the same methods applied to the sedimentary leaf waxes. In total, 3 C_4 taxa and 17 C_3
243 taxa are included in our analysis. A full list of taxa sampled, as well as C_{30} n -acid concentrations
244 and isotopic values, is provided in Supplementary Table 3. These values of $\delta^{13}\text{C}$ inferred from
245 modern plants were corrected for isotopic changes in atmospheric carbon associated with the Suess
246 effect. Corrected and uncorrected values are in Extended Tables 2 and 3.

247 Estimates of $\varepsilon_{\text{water-wax}}$ for C_4 and C_3 taxa are obtained from the data compilation in ref. ⁵⁰.

248 ε values are based on the C_{29} n -alkane rather than the C_{30} n -acid; given that existing research dis-

249 agrees about whether there are significant offsets between *n*-acid and *n*-alkane apparent fraction-
250 ations we assume that these are equivalent^{50,54}. The corrections were performed using a Bayesian
251 framework, as detailed in ref. ⁵⁵. This carbon correction had a minimal impact on the overall trends
252 in our data since down-core $\delta^{13}\text{C}$ changes are not large; indeed, raw δD_{wax} is strongly correlated
253 with inferred δD_{precip} ($r = 0.92$). Down-core δD_{precip} values are shown in Supplementary Figure
254 1.

255 We analyzed the correlation between coretop δD_{precip} values and climate data from adja-
256 cent land areas using the high-resolution (0.3°) North American Regional Reanalysis (NARR)⁵⁶.
257 Because of known biases in NARR evaporation and moisture transports⁵⁷, we only use the precip-
258 itation and temperature, which are directly assimilated. Seasonally-averaged NARR precipitation
259 exhibits little bias over northwest Mexico⁵⁸. For each coretop site, terrestrial grid cells within 1°
260 of latitude and 2° of longitude were used to determine the average climate of the region contribut-
261 ing leaf waxes to the coretop site. The strongest relationship is with % July-September (% JAS,
262 $r = 0.79$) rainfall (Figure 1). This relationship reflects differences in the isotopic composition
263 of winter and summer rainfall: the Global Network of Isotopes in Precipitation (GNIP) station in
264 Tucson, AZ shows that summer rainfall is more enriched in deuterium relative to winter rainfall
265 (Supplementary Figure 5), likely as a result of differences of temperature, strength of convective
266 updrafts, and water vapor source region for each season^{59,60}. Thus, sites with increased monsoon
267 rainfall exhibit more positive rainfall δD values. This interpretation is distinct from the ‘amount
268 effect,’ which would predict a more depleted isotopic signature as rainfall rates increase moving
269 south along the Mexican Margin⁶¹. We suggest that seasonality dominates our isotopic signa-

ture because this region features a bimodal rainfall distribution with distinct isotopic compositions for winter and summer precipitation that swamp any signal from the amount effect. Rainfall data from Tucson reveals that the isotopic offset between winter and summer rainfall is larger than the magnitude of isotopic variability associated with interannual changes in rainfall amount⁵⁹.

Inferred δD_{precip} might reflect evaporative enrichment of leaf water, as ε in individual taxa has been shown to vary across aridity gradients^{62,63}. However, this effect may be attenuated in sedimentary leaf waxes, which are bulk mixtures of multiple plant taxa that each have different environmental sensitivities to leaf water enrichment. In addition, plants in regions with a greater proportion of %July-September rainfall may register greater leaf water enrichments due to higher temperatures during wax synthesis. Thus, even if our δD signal does not record precipitation isotopes with perfect fidelity, evaporative enrichment may amplify the statistical link between wax isotopic composition and %July-September rainfall. Changes in annual temperature may influence δD values at the LGM, but any such influence is minor, as SST reconstructions from the Guaymas Basin suggest a 4°C cooling at the LGM relative to early Holocene values³⁴, implying a 4‰ decrease in δD .

A quantitative relationship between %JAS rainfall and δD_{precip} , as well as reconstructions of % monsoon rainfall over the deglaciation, were built using Bayesian inference following the methods outlined in ref.⁵⁵. We developed a Bayesian regression relating δD_{precip} (Y) to % July-September rainfall (first column of X ; second column of X consists of ones), where:

$$Y = X \cdot \beta + \epsilon \quad (5)$$

$$\epsilon \sim N(0, \tau^2), IID \quad (6)$$

289 β is a vector of regression coefficients and ϵ is the model error, with independent and iden-
 290 tically normally-distributed errors (IID) and a variance of τ^2 (Eq. 6). We estimate the parameters
 291 using Bayes' Rule, using Normal prior distributions for β and an Inverse Gamma prior for τ^2 ⁵⁵.
 292 The conditional posterior distributions for each parameter are conjugate with the prior distribu-
 293 tions, meaning that they follow the same distributional form⁶⁴. Calculation of the full posterior
 294 solution for each parameter proceeds via a Gibbs sampler⁶⁴, where initial values for τ^2 and β are
 295 specified, and values are sequentially sampled from their respective conditional posterior distribu-
 296 tions. Supplementary Figure 6 shows the prior and posterior distributions for the regression model
 297 parameters.

298 Our ultimate goal is to infer % JAS values from down-core measurements of δD_{precip} . To do
 299 so, we must invert the regression model developed above using another application of Bayes' rule
 300 relating % JAS (X) to δD_{precip} (Y):

$$P(X|Y) \propto P(Y|X) \cdot P(X) \quad (7)$$

301 $P(Y|X)$ is our regression model, and $P(X)$ is a specified prior distribution. The prior dis-

302 tribution is Normal ($N(\mu_p, \sigma_p^2)$). The prior mean is set to modern climatological % JAS at each
303 core site, while σ_p is set to 12, to match the observed variability in % July-September rainfall in
304 the NARR dataset, which we take as a plausible range of values. Inference of % July-September
305 rainfall proceeds by drawing from the suite of inferred δD_{precip} values and the posterior distribu-
306 tion of parameters from the regression model. This process iterates through all previously inferred
307 down-core δD_{precip} values, so the variance at each point in our final data includes the full range of
308 error from both our δD_{precip} reconstructions and the reconstructions of % July-September rainfall.
309 Comparison of the prior and posterior distributions of modern inferred % July-September rainfall
310 at each core site shows that the posterior distribution is much narrower as a result of the incor-
311 poration of the information from the δD_{precip} data (Supplementary Figure 7). In Figure 2, error
312 envelopes represent the 68% and 95% confidence intervals for each datapoint, and also incorporate
313 age uncertainty. For full details of the Bayesian approach, see ref⁵⁵.

314 **HadCM3 Model Simulations.** We explored the drivers of deglacial NAM changes using a series
315 of ‘timeslice’ simulations conducted with the model HadCM3. HadCM3 is an Earth system model
316 developed by the UK Meteorological Office, and consists of an atmospheric model (HadAM3)
317 coupled to an ocean and sea ice model, as well as a land surface model⁶⁵. The atmosphere, ocean,
318 and sea ice components are run at a relatively coarse resolution of 2.5° latitude by 3.75° longitude
319³⁷. Despite the coarse resolution, the pre-industrial control produces a NAM with rainfall rates
320 comparable to instrumental data (Supplementary Figure 8). The land surface model, the Met Of-
321 fice Surface Exchange Scheme (MOSES2.1), includes a vegetation and terrestrial carbon model
322 (TRIFFID)⁶⁶. However, atmospheric concentrations of greenhouse gases (CO₂, CH₄, and N₂O)

323 were prescribed based on gas concentrations obtained from ice core data³⁶. Orbital parameters and
324 ice sheets were also prescribed, with the latter derived from the ICE 5G dataset⁶⁷. Freshwater forc-
325 ing in the North Atlantic associated with abrupt deglacial climate events is not included in these
326 simulations.

327 ‘Timeslice’ model simulations including dynamic vegetation for each 1,000-year timestep
328 were initiated from a previous set of simulations and allowed to run to equilibrium³⁶. The timesteps
329 between 21 ka and 0 ka were used for the analyses in this paper. We analyzed NAM domain av-
330 erages (18 – 33°N and 102 – 112°W) of precipitation, sea surface temperature, air temperature,
331 geopotential height, zonal and meridional winds, and specific humidity. We also calculated moist
332 static energy (MSE), which measures an air column’s energetic content, integrating information
333 from the geopotential height (parcel’s height z multiplied by the gravitational constant g), temper-
334 ature (T multiplied by the specific heat C_p), and moisture content (specific humidity q multiplied
335 by the latent heat of vaporization L_v) (Equation 8). MSE is conserved in adiabatic processes and
336 broadly measures the favorability of atmospheric conditions for convection⁶⁸. This approach is
337 supported by the fact that the correlation between summer precipitation anomalies and changes in
338 MSE over the NAM region is 0.86 across timeslice simulations.

$$MSE = C_p \cdot T + L_v \cdot q + g \cdot z \quad (8)$$

339 In these simulations, we find that % July-September rainfall, the parameter we reconstruct
340 from leaf wax data, strongly covaries with changes in total summertime rainfall ($r^2 = 0.81$), and is

341 only weakly related to winter rainfall amount ($r^2=0.04$) (Supplementary Figure 2). This supports
342 our inference that % July-September rainfall changes in the NAM domain are ultimately reflecting
343 summertime circulation.

344 Recent studies have suggested local land surface evapotranspiration and albedo may influ-
345 ence the strength of the NAM circulation ^{3,6,7}. The inclusion of dynamic vegetation in this model
346 allows us to analyze the impact of changes in land surface characteristics on the NAM. In HadCM3,
347 land surface evapotranspiration has little correlation with deglacial changes in % monsoon rain-
348 fall, and the surface albedo remains constant across simulations (Supplementary Figure 2). This
349 suggests land surface processes have little impact on monsoon strength, but further testing of these
350 hypotheses would benefit from regional climate model simulations.

351 **Comparisons between reconstructed NAM changes, ice sheet extent, and SST.** Our compar-
352 ison between reconstructed % July-September rainfall and ice sheet extent (Figure 4a) accounts
353 for age model and reconstructed uncertainty in a Monte Carlo fashion. We both 1) resample from
354 the suite of age models for each sediment core to identify the full range of possible values of leaf
355 wax δD for a given absolute calendar age interval during the deglaciation; and 2) resample the full
356 range of error in our reconstructions of % July-September rainfall for each core site. This approach
357 yields the complete range of possible values of % July-September rainfall for any absolute calendar
358 age interval.

359 Reconstructions of ice sheet variables from the LGM to the present were obtained from the
360 ICE6G dataset, which uses geodetic measurements of crustal motion and a dynamical model to

361 constrain the chronology of ice sheet retreat^{35,69}. The resultant model of ice sheet change is tuned
362 using exposure dating and detailed sea-level records. We obtained values for the southernmost
363 latitude of ice over North America for 500-year intervals from 21,000 years BP to 1,000 years
364 BP. Since the output comes from a dynamic model, we do not account for age uncertainty. The
365 southernmost latitude of ice correlates strongly with other metrics of ice extent, including ice area
366 ($r = 0.93$).

367 After obtaining distributions of possible % July-September rainfall values for each 100-year
368 interval during the deglaciation, we binned the results based on their corresponding value of south-
369 ernmost latitude of ice from the ICE6G dataset, and constructed the 95% confidence intervals
370 shown in Figure 4a. To illustrate the shape of the relationship, we fit these values with a second-
371 order polynomial. Error estimates on polynomial coefficients are based on fitting a suite of poly-
372 nomials to the full suite of output from our Monte Carlo simulations (Supplementary Table 4). We
373 also calculated coefficients for HadCM3 timeslice simulations, but were not able to calculate coef-
374 ficients errors since only one set of model runs were available. We assessed statistical significance
375 by calculating the adjusted R^2 and performing a separate F-test.

376 Deglacial records of Gulf of California SST only exist for the Guaymas Basin, so we restrict
377 ourselves to analyzing the relationship between SST and monsoon rainfall at the Guaymas Basin
378 site. Future research, including the development of quantitative deglacial SST reconstructions
379 from the Mexican Margin, will permit more rigorous testing of the relationship between regional
380 ocean temperature and monsoon convection. The reconstruction for the Guaymas Basin is based
381 on alkenones obtained from MD02-2515, the same core from which we extracted leaf waxes³⁴. We

382 restrict our analyses to the leaf wax reconstructions from MD02-2515, since we can control for age
383 uncertainty when comparing measurements from the same core. In this case, we construct a scat-
384 terplot of SST vs. % July-September rainfall, and error bars in this case represent the propagation
385 of analytical error in both leaf wax data and the alkenone data and uncertainty from our Bayesian
386 reconstruction of % July-September rainfall.

387 **Data Availability Statement** New leaf wax-based reconstructions will be made available for
388 download from NOAA's National Center for Environmental Information Paleoclimatology Database
389 <https://www.ncdc.noaa.gov/data-access/paleoclimatology-data>, and by contacting the correspond-
390 ing author. Previously published alkenone data from the Guaymas Basin are also available via
391 the NOAA database. HadCM3 model code is available at [http://cms.ncas.ac.uk/code_](http://cms.ncas.ac.uk/code_browsers/UM4.5/UMBrower/index.html)
392 [browsers/UM4.5/UMBrower/index.html](http://cms.ncas.ac.uk/code_browsers/UM4.5/UMBrower/index.html). Simulations used in this paper³⁷ are avail-
393 able at [https://www.paleo.bristol.ac.uk/ummodel/scripts/papers/Davies-Barnard_](https://www.paleo.bristol.ac.uk/ummodel/scripts/papers/Davies-Barnard_et_al_2017.html)
394 [et_al_2017.html](https://www.paleo.bristol.ac.uk/ummodel/scripts/papers/Davies-Barnard_et_al_2017.html). ICE6G data are available at [http://www.atmosp.physics.utoronto.](http://www.atmosp.physics.utoronto.ca/~peltier/data.php)
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579 **Author Contributions** T.B. and J.E.T. designed the study, collected the data, and performed the data
580 analysis. J.A.A. and J.W.M. provided access to samples and assisted in data collection. T.B. and J.E.T.
581 wrote the manuscript.

582 **Competing Interests** The authors declare that they have no competing financial interests.

583 **Correspondence** Correspondence and requests for materials should be addressed to T.B. (email: tripti@email.arizona.edu)

584 1 Main Text Figure Captions

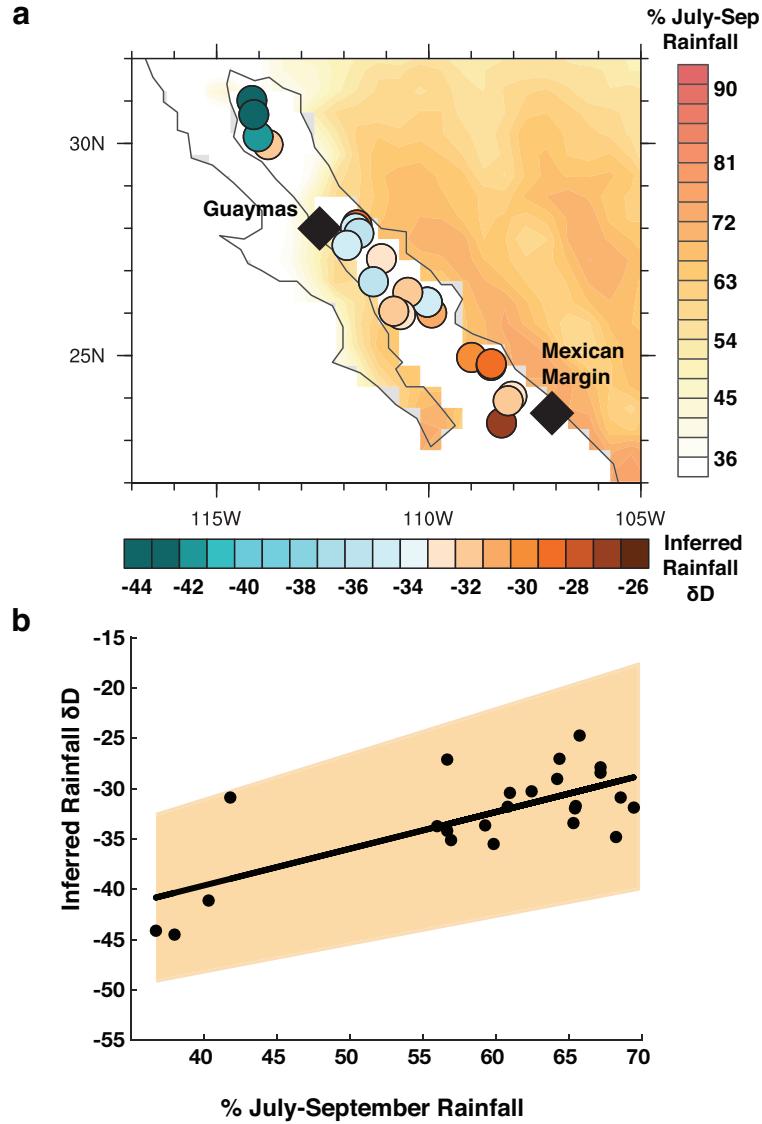


Figure 1: Relationship between leaf waxes δD and monsoon rainfall. a) core sites (diamonds) and coretop data (circles) with leaf wax-inferred δD of precipitation (colors of circles). Color contours show climatological % July-September (monsoon season) rainfall ⁴⁵. b) Bayesian regression between coretop leaf-wax inferred δD_{precip} and %July-September rainfall, with the 95% confidence interval.

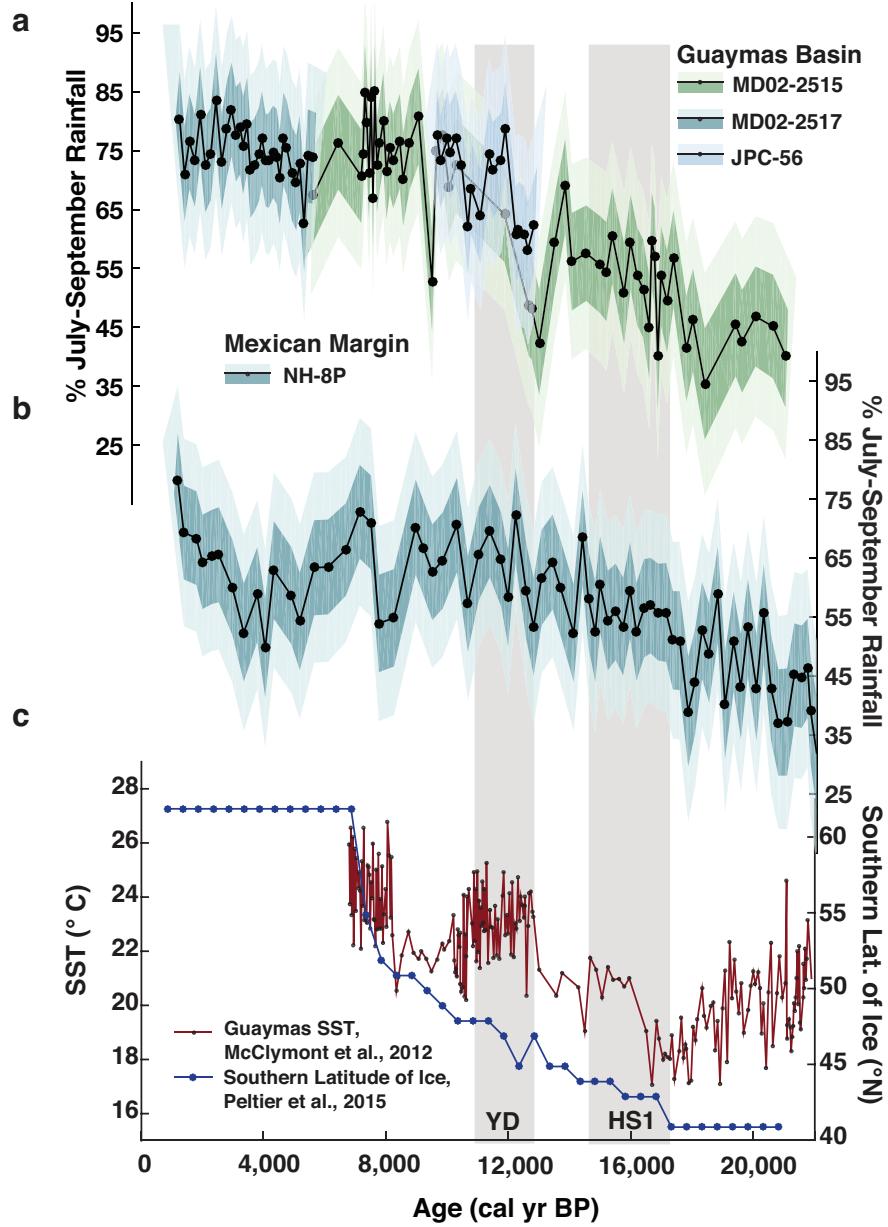


Figure 2: Leaf-wax based reconstructions of % July-September rainfall at the Guaymas Basin

(a) and Mexican Margin (b) sites. Dark and light error bars indicate 1σ and 2σ uncertainties in the reconstructions respectively, while the central black line indicates the mean value. c) shows changes in regional forcings, including the southernmost latitudinal extent of the Laurentide ice sheet³⁵ and Guaymas Basin SSTs³⁴. Gray bars highlight the Younger Dryas (YD; 12.8-11.5 ka) and Heinrich Stadial 1 (HS1; 17.5-14.5 ka).

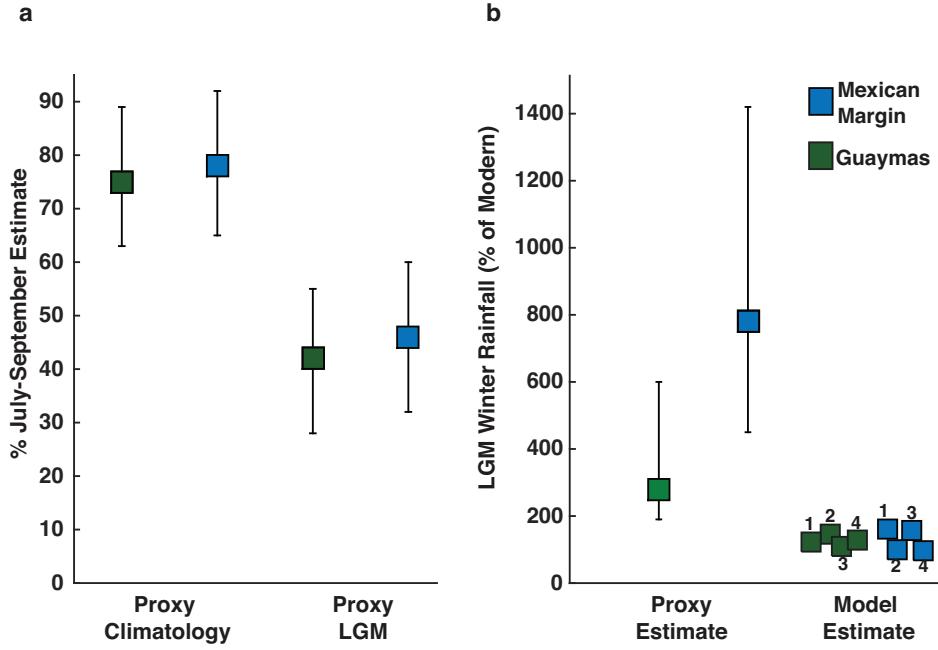


Figure 3: Sensitivity of monsoon reconstructions to winter rainfall changes. a) Modern proxy-estimated values of % July-September rainfall for each core site versus proxy estimates of LGM values (averaged between 22 and 18 ka). b) shows % change in November-March rainfall needed if winter rainfall was the sole driver of deglacial % JAS changes. Error bars represent 95% confidence intervals. Model estimates of changes in winter rainfall are included from PMIP3 archive ¹⁷, with 1=CNRM-CM4, 2=CCSM4, 3=MPI-ESM, and 4=MIROC.

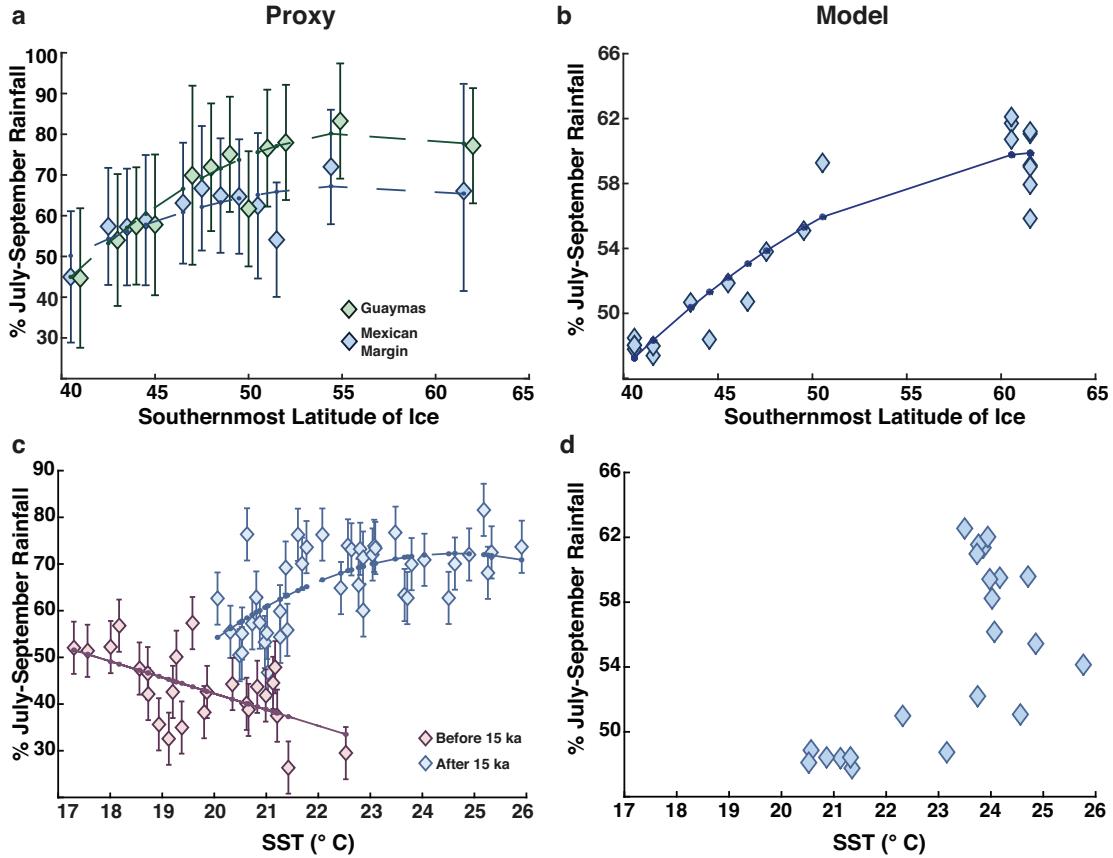


Figure 4: Atmospheric and oceanic drivers of glacial NAM changes. a) % July-September rainfall vs. the southernmost latitude of ice cover. Error bars represent 95% confidence intervals for % July-September values. b) Modeled NAM-region % JAS sensitivity to ice extent across HadCM3 simulations. c) Guaymas Basin % July-September rainfall vs. SSTs, in contrast to d) modeled relationship between SST and % July-September rainfall averaged between 102-112 $^{\circ}$ W and 20-35 $^{\circ}$ N.