



Phytoplankton community, seasonality, and water chemistry modulate the lacustrine diurnal carbon engine and carbonate $\delta^{13}\text{C}$ values

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Abstract

Stable isotopic measurements of carbon in carbonate rocks ($\delta^{13}\text{C}_{\text{carb}}$) have long been used to investigate paleoclimate and global carbon cycle dynamics. Recent work documented the diurnal carbon engine effect, which describes the impact of daily photosynthetic cyclicality on dissolved inorganic carbon (DIC) and, in turn, $\delta^{13}\text{C}_{\text{carb}}$ and $\delta^{13}\text{C}_{\text{DIC}}$ values, in low-latitude shallow marine environments. Because carbonate precipitation is temporally structured over the diurnal cycle, the resulting $\delta^{13}\text{C}_{\text{carb}}$ reflects a precipitation-weighted bias rather than a simple daily average. However, this system has not yet been explored in lakes. Here we present an adapted diurnal carbon engine model and constrain this effect for three end-member lake systems: Great Salt Lake (UT, USA), Green Lake (NY, USA), and Morrison Lake (MT, USA). We document how differing lake water chemistry and phytoplankton communities modulate local diurnal carbon engines. Because of geochemical and ecological differences, the impact of the diurnal carbon engine on $\delta^{13}\text{C}_{\text{carb}}$ values varies greatly by lake system. We estimated $\Delta^{13}\text{C}$ offsets, which represent the daily variability expected resulting from the diurnal engine and precipitation weighting, of $\sim 0.36\text{‰}$ in Great Salt Lake and $\sim 0.24\text{‰}$ in Green Lake. We also modeled how seasonality impacts the diurnal carbon engine in Great Salt Lake, which resulted in $\sim 0.10\text{‰}$ of $\delta^{13}\text{C}_{\text{carb}}$ variability. Modeled diurnal variability is consistent with observed dynamics in Green Lake. Predicted $\delta^{13}\text{C}_{\text{carb}}$ values are consistent with those measured from most carbonate sedimentary facies in each lake. Lacustrine $\delta^{13}\text{C}_{\text{carb}}$ shifts may reflect local environmental and ecological conditions, rather than changes in long-term lake $\delta^{13}\text{C}_{\text{DIC}}$ values or the global carbon cycle, complicating lacustrine $\delta^{13}\text{C}_{\text{carb}}$ record interpretations.



39 1 Introduction

40 Sediments and sedimentary rocks formed in lacustrine (lake) environments preserve
41 valuable records of continental climate dynamics across Earth's history. One of the most
42 important types of geochemical records is the measurement of stable carbon isotope ratios of
43 carbonate sedimentary rocks (reported in delta notation as $\delta^{13}C_{carb} = \left[\frac{\left[\frac{^{13}C}{^{12}C} \right]_{carbonate}}{\left[\frac{^{13}C}{^{12}C} \right]_{standard}} - 1 \right] \times$
44 1000) from both marine and lacustrine settings (e.g., Berner, 1990; Kump and Arthur, 1999;
45 Zachos et al., 2001). While there are many ways that lacustrine $\delta^{13}C_{carb}$ records can be perturbed
46 from equilibrium, excursions in $\delta^{13}C_{carb}$ are often linked to environmental or climatic events
47 (e.g., Talbot, 1990; Valero-Garcés et al., 1997; Briner et al., 2006). These interpretations often
48 assume that the dominant driver of $\delta^{13}C_{carb}$ change in lakes is global change in the $\delta^{13}C$ values of
49 the marine–atmosphere system, which propagates into the dissolved inorganic carbon reservoir
50 (DIC) of lakes. These same assumptions underlie efforts to relate fluctuations in $\delta^{13}C_{carb}$ records
51 to variations in the global carbon cycle. Recent work has demonstrated that a number of other
52 local-scale processes can impact carbon isotope values and our ability to interpret these
53 measurements, including chemical alteration (e.g., Higgins et al., 2018; Swart and Oehlert,
54 2018), rock composition (i.e., sedimentary facies) variability (e.g., Ingalls et al., 2020; Geyman
55 and Maloof, 2021), variation in biological mineralizers (e.g., Gischler et al., 2009), and the so-
56 called “diurnal engine” (e.g., Geyman and Maloof, 2019; Trower et al., 2024).

57 The diurnal engine describes how daily fluctuations in the concentration of dissolved
58 inorganic carbon (DIC) and pCO_2 result from the cyclic intensity of photosynthesis (due to the
59 diel cycle). This in turn drives corresponding fluctuations in pH and carbonate mineral saturation
60 states (see Zeebe and Wolf-Gladrow, 2001), and ultimately causes the carbon isotopic



61 composition of DIC ($\delta^{13}\text{C}_{\text{DIC}}$) to change. The carbon isotopic composition of DIC becomes
62 relatively more enriched in ^{13}C during the day due to the preferential consumption of ^{12}C during
63 photosynthesis and relatively more depleted in ^{13}C throughout the night due to the release of ^{12}C -
64 enriched CO_2 through biological respiration (see **Figure 1**). Therefore, the diurnal engine
65 suggests that bulk $\delta^{13}\text{C}_{\text{carb}}$ values can be offset from long-term, whole-lake average $\delta^{13}\text{C}_{\text{DIC}}$
66 values (often described as background $\delta^{13}\text{C}_{\text{DIC}}$ values) by a greater magnitude than would be
67 predicted for carbonate minerals forming in equilibrium with long-term average DIC (often
68 described as background DIC), depending on biases in the timing of carbonate formation during
69 diel cycles; this effect complicates interpretations of $\delta^{13}\text{C}_{\text{carb}}$ records. These diurnal changes in
70 water chemistry and their resulting consequences on $\delta^{13}\text{C}_{\text{carb}}$ values have been documented in
71 several low-latitude shallow marine environments (Geyman and Maloof, 2019; Trower et al.,
72 2024).

73 A key consequence of the diurnal engine is that changes in depositional environment,
74 which can result in changes in primary productivity or isotopic fractionation and be recorded by
75 sedimentary facies, could result in different offsets between $\delta^{13}\text{C}_{\text{carb}}$ and $\delta^{13}\text{C}_{\text{DIC}}$ values. Some
76 secular fluctuations in $\delta^{13}\text{C}_{\text{carb}}$ records could therefore be entirely driven by the local diurnal
77 carbon engine and not, in fact, reflect any external (or correlative) processes (see Background
78 and **Figure 1**). The diurnal engine model implies that carbonate does not precipitate uniformly
79 through time, but is biased toward periods of elevated saturation state, potentially amplifying
80 isotopic offsets. Lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records are often interpreted to be driven by global carbon
81 cycle processes and thus reflect Cenozoic paleoclimate (e.g., Colman et al., 2007; Tanner, 2010;
82 Rugenstein and Chamberlain, 2018; Lettéron et al., 2018; Benavente et al., 2021), and
83 constraining how the diurnal engine effect modifies $\delta^{13}\text{C}_{\text{carb}}$ values is essential to being able to



84 accurately interpret these archives. For example, a relatively strong diurnal engine ($>1\%$ offset
85 from predicted $\delta^{13}\text{C}_{\text{carb}}$ without a diurnal engine) could result in $\delta^{13}\text{C}_{\text{carb}}$ excursions that are
86 erroneously interpreted as the result of global carbon cycle changes, but instead reflect a change
87 in phytoplankton community, which in turn could be associated with a change in sedimentary
88 facies. Said differently, an environmental change, recorded in sedimentary lithofacies, could
89 modify the overall productivity of that environment, ultimately shifting κ_p (the photosynthetic
90 forcing) and the local diurnal carbon engine. Likewise, an ecological shift in the dominant
91 phytoplankton community could change background isotopic fractionation (ϵ_p , where $\epsilon_p \approx$
92 $\delta^{13}\text{C}_{\text{CO}_2(\text{aq})} - \delta^{13}\text{C}_{\text{biomass}}$). Compared to shallow marine systems, lacustrine environments are
93 often characterized by stronger spatial and temporal variability in productivity and carbonate
94 saturation, suggesting that the magnitude and variability of the diurnal engine effect may be
95 greater and more heterogeneous. Therefore, both environmental and ecological drivers could
96 produce cryptic fluctuations in bulk $\delta^{13}\text{C}_{\text{carb}}$ values unrelated to sympathetic fluctuations in long-
97 term average $\delta^{13}\text{C}_{\text{DIC}}$ values (as is often invoked when interpreting $\delta^{13}\text{C}_{\text{carb}}$ records). Note that
98 environmental and ecological shifts are often related, further complicating the impact of carbon
99 isotope records. Several recent studies in modern lacustrine settings have reported microbialite
100 carbonate $\delta^{13}\text{C}_{\text{carb}}$ values that are up to 4% higher than predicted for carbonate minerals forming
101 in isotopic equilibrium with average lake water DIC (Warden et al., 2016; Belan et al., 2019;
102 White et al., 2021; Chen et al., 2022). These data provide compelling evidence that the diurnal
103 engine effect is likely significant in lacustrine carbonate strata, suggesting that this mechanism
104 may be responsible for and/or amplify or dampen some shifts in key lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records
105 (although these data could also be evidence for carbon limitation in microbial mats, changing ϵ_p).



106 At present, the parameters needed to model how diel cycling (and therefore the resulting
107 photosynthetic forcing) controls the magnitude of the diurnal engine have only been somewhat
108 constrained in marine environments (see Geyman and Maloof, 2019; Trower et al., 2024).
109 Lacustrine photosynthetic communities are distinct from those found in shallow marine
110 environments (e.g., Konopka and Brock, 1978; Hasle and Syversten, 1996), typically being
111 dominated by freshwater algae or cyanobacteria rather than zooxanthellate reefs, marine algae,
112 and marine cyanobacteria. Furthermore, many lakes have different physiochemical properties
113 than the ocean, such as pH and buffering capacity, meaning lake carbonate chemistry can more
114 easily shift between favoring dissolution and precipitation, ultimately impacting the timing of
115 carbonate formation and the potential biases of $\delta^{13}\text{C}_{\text{carb}}$ values. Previously published constraints
116 from marine systems are thus poor indicators for the lacustrine systems discussed herein.
117 Furthermore, the enhanced seasonality and high spatial and temporal variation in primary
118 productivity in carbonate-producing lakes relative to shallow tropical marine settings suggest
119 that the magnitude and variability of the effect may be greater in lacustrine settings (e.g.,
120 Marshall and Peters, 1989; Catalan et al., 2002; Muchanga and Sickingabula, 2021). Here, we
121 present a modified version of the Trower et al. (2024) diurnal engine model that facilitates new
122 lacustrine applications, includes a new $\Delta^{13}\text{C}$ framework, and compares well with observational
123 data. We employ this model to evaluate how productivity and alkalinity jointly control the
124 magnitude diurnal isotopic offsets across a range of lacustrine conditions with a specific focus on
125 seasonality and phytoplankton community. Together, these analyses provide a framework for
126 understanding how diurnal carbon cycling influences lacustrine carbonate isotope records and
127 how these effects may be distinguished from longer-term environmental change.

128



129 **2 Background**

130 **2.1 Lacustrine stable carbon isotope dynamics**

131 Traditionally, the fraction of organic carbon burial (f_{org}) (relative to the total carbon burial
132 flux) has been used to explain carbon isotope fluctuations in marine carbonates as a function of
133 environmental change (Kump and Arthur, 1999). For example, a global positive carbon isotope
134 excursion could be explained by increased burial of organic carbon (i.e., higher f_{org}), which
135 sequesters a greater proportion of ^{12}C in the lithospheric carbon reservoir. This, in turn, increases
136 the $\delta^{13}\text{C}$ value of DIC ($\delta^{13}\text{C}_{\text{DIC}}$) in the ocean. While this framework for interpreting marine $\delta^{13}\text{C}$
137 excursions can be useful, work over the last couple of decades has demonstrated that $\delta^{13}\text{C}$
138 dynamics are often more complicated, with many other processes and carbon sources (e.g.,
139 methane hydrates or authigenic carbonate formation; Bjerrum and Canfield, 2011; Schrag et al.,
140 2013) having the potential to drive large magnitude and global-scale perturbations in $\delta^{13}\text{C}$
141 records.

142 Lacustrine $\delta^{13}\text{C}$ records reflect many of the same processes and carbon sources as marine
143 $\delta^{13}\text{C}$ records, allowing geochemists to apply the substantial body of research from marine
144 settings to lacustrine settings. However, there are key differences, such as a substantially smaller
145 DIC reservoir, distinct flora and fauna, and a greater importance of some carbon cycle processes
146 (such as bedrock interaction and evaporation) (e.g., Leng and Marshall, 2004). Because of the
147 large size of the oceanic reservoir, marine $\delta^{13}\text{C}$ variability is often assumed to reflect global
148 processes (facilitating global chemostratigraphic correlation), with ocean mixing posing a
149 minimum constraint on the temporal resolution (~ 1000 yr) of $\delta^{13}\text{C}$ records. In contrast, any given
150 lake is significantly smaller than the ocean, and most carbonate-producing lakes are closed-
151 basins, in which regional-scale evaporation and concomitant effects on organic matter



152 preservation can drive strong covariation between $\delta^{13}\text{C}_{\text{carb}}$ and $\delta^{18}\text{O}_{\text{carb}}$ isotopes, complicating
153 global interpretations of $\delta^{13}\text{C}_{\text{carb}}$ records (e.g., Leng and Marshall, 2004). Processes on relatively
154 short time scales (< 1000 yr) are rarely considered when assessing (marine or lacustrine) $\delta^{13}\text{C}$
155 records, much less daily variability (as discussed herein).

156

157 **2.2 The diurnal carbon engine model for daily $\delta^{13}\text{C}_{\text{carb}}$ variability**

158 The diurnal engine describes the daily transition from local carbon cycling dominated by
159 aquatic photosynthetic organisms (e.g., green algae, cyanobacteria) during the day to carbon
160 cycling dominated by heterotrophy at night (e.g., Geyman and Maloof, 2019; Trower et al.,
161 2024). During the day, $\delta^{13}\text{C}_{\text{DIC}}$ values and the carbon isotope ratios of the carbonate minerals that
162 precipitate from that DIC pool ($\delta^{13}\text{C}_{\text{carb}}$) increase, in addition to the rate of carbonate mineral
163 formation, while $\delta^{13}\text{C}_{\text{DIC}}$ values, $\delta^{13}\text{C}_{\text{carb}}$ values, and the rate of carbonate mineral formation all
164 decrease at night (**Figure 1**). These changes occur due to indirect effects of photosynthesis and
165 aerobic respiration on the residual pool of DIC. The net effect of this process is that more
166 carbonate precipitates when $\delta^{13}\text{C}_{\text{DIC}}$ values are elevated, meaning that bulk $\delta^{13}\text{C}_{\text{carb}}$ values of
167 precipitated minerals will tend to more closely match daytime $\delta^{13}\text{C}_{\text{DIC}}$ values rather than daily
168 averages (**Figure 1**). Although phototrophs and heterotrophs are abundant in most lakes and
169 shallow marine environments, the magnitude of net photosynthetic forcing can vary significantly
170 between and within ecosystems (e.g., Verduin, 1956; Carpenter et al., 1987; Howarth, 1988;
171 Lewis Jr, 2011). This forcing affects the magnitude of change in $\delta^{13}\text{C}_{\text{DIC}}$ values, so when this
172 forcing is small, $\delta^{13}\text{C}_{\text{carb}}$ values closely match the values predicted for carbonate minerals
173 precipitating in equilibrium with average $\delta^{13}\text{C}_{\text{DIC}}$ values (see Geyman and Maloof, 2019; Trower
174 et al., 2024). This key observation forms the basis of using $\delta^{13}\text{C}_{\text{carb}}$ values as proxies for



175 environmental or paleoclimatic change. In contrast, when the magnitude of photosynthetic
176 forcing is large, like in modern coral reef environments (see Geyman and Maloof, 2019), the
177 diurnal engine can lead to large differences ($\geq 6\text{‰}$) between actual and predicted $\delta^{13}\text{C}_{\text{carb}}$ values
178 (e.g., Geyman and Maloof, 2019). These large differences complicate interpretations of this
179 important geochemical proxy, meaning that an understanding of local diurnal carbon cycling and
180 high-resolution, paired $\delta^{13}\text{C}_{\text{carb}}\text{--}\delta^{13}\text{C}_{\text{org}}$ measurements may be necessary to tease apart local and
181 global $\delta^{13}\text{C}_{\text{carb}}$ signals. Furthermore, because carbonate precipitation rates vary through the
182 diurnal cycle as a result of changes in carbonate mineral saturation state and temperature,
183 resulting bulk carbonate isotopic compositions reflect a precipitation-weighted integration of
184 $\delta^{13}\text{C}_{\text{DIC}}$ values rather than simply a daily average.

185 To date, the diurnal engine effect has only been quantified in equatorial, shallow marine
186 carbonate islands (Geyman and Maloof, 2019; Trower et al., 2024), but this work has
187 demonstrated that some positive $\delta^{13}\text{C}_{\text{carb}}$ excursions could be entirely driven by shifts in
188 photosynthetic forcing rather than shifts in globally averaged marine $\delta^{13}\text{C}_{\text{DIC}}$ values as typically
189 argued. Geyman and Maloof (2019) used a compilation of geochemical datasets from modern
190 coral reef environments to posit a likely range of photosynthetic forcing in modern shallow
191 marine environments. However, recent work conducted by Trower et al. (2024) found that a
192 location on a modern carbonate platform (characterized by grainy sediment and sparse calcifying
193 algae) resulted in photosynthetic forcing lower than almost all observed modern coral reefs.
194 Therefore, local productivity appears to play a significant role in modulating the diurnal engine
195 effect. Additionally, these findings demonstrate that the coral reef photosynthetic forcing
196 parameters derived from the original diurnal engine study are not broadly representative of
197 shallow marine environments with different photosynthetic communities, much less lacustrine



environments. Photosynthetic forcing could also vary significantly over small spatial scales (< 1 m) due to significant differences in productivity related to different photosynthetic communities (and environments, nutrients, etc.). In addition, because of the importance of local productivity in lakes, one cannot simply apply the marine diurnal engine values from previous work to lacustrine settings. In this contribution, we use a range of photosynthetic forcing values more representative of lakes to broadly explore the diurnal engine effect in lacustrine systems and suggest that future work seek to constrain this forcing in a variety of lacustrine systems.

2.3 Phytoplankton & carbon concentrating mechanisms

Primary producers are important to lake ecosystems, making them key to understanding biogeochemical and ecological change. Green algae and cyanobacteria are common primary producers in lakes with differing evolutionary backgrounds and ecological roles (e.g., Rockwell et al., 2014). Green algae are eukaryotes that rely on oxygenic photosynthesis, using chlorophylls a and c to capture light energy and convert carbon dioxide into organic compounds (e.g., Leliaert et al., 2012).

Cyanobacteria are divided into two groups (α -cyanobacteria and β -cyanobacteria) based on the type of carbon concentrating mechanism (CCM) they possess and the form of ribulose-1,5-bisphosphate carboxylase/oxygenase (RuBisCO) they encapsulate (1A and 1B RuBisCO, respectively). α -cyanobacteria and β -cyanobacteria exhibit adaptations that reflect their ecological niches and evolutionary histories (Badger et al., 2002). Broadly speaking, α -cyanobacteria are typically found in nutrient-poor, oligotrophic environments (e.g., marine) (e.g., Badger and Price, 2003). In contrast, β -cyanobacteria are broadly more commonly associated



220 with nutrient-rich, eutrophic waters (e.g., lakes) (e.g., Badger and Price, 2003). They also
221 commonly form dense blooms (e.g., Huisman et al., 2018).

222 Both types of cyanobacteria and green algae are distinct in how they fix carbon.
223 Specifically, cyanobacteria and green algae exhibit distinct CCMs that enhance their
224 photosynthetic efficiency by increasing the local concentration of carbon dioxide around the
225 enzyme responsible for fixing CO₂, RuBisCO. In cyanobacteria, the CCM is called a
226 carboxysome, which is a proteinaceous microcompartment that encapsulates RuBisCO and
227 carbonic anhydrase (Rae et al., 2013; Hill et al., 2020; Huffine et al., 2023). These cyanobacteria
228 CCMs efficiently convert bicarbonate (HCO₃⁻) into CO₂ directly at the site of RuBisCO, thereby
229 minimizing photorespiration and enhancing carbon fixation efficiency even in low CO₂
230 conditions (Badger and Price, 2003). Additionally, cyanobacteria actively transport inorganic
231 carbon species (CO₂ and HCO₃⁻) into the cell using specialized transporters, further boosting
232 intracellular CO₂ levels.

233 Green algae represent a broad designation with a variety of CCM strategies that are
234 generally less understood when compared to cyanobacteria. Many green algae utilize pyrenoids,
235 which are specialized structures within the chloroplasts that concentrate CO₂ around RuBisCO
236 (Mackinder et al., 2016; Wang et al., 2023). These structures often work in conjunction with the
237 active transport of HCO₃⁻ and CO₂ across cellular membranes, facilitated by carbonic anhydrases
238 located in various cellular compartments, including the chloroplast stroma (Giordano et al.,
239 2005). The exact mechanisms can vary significantly among different groups of green algae,
240 reflecting their evolutionary diversity.

241 Distinct carbon concentrating mechanisms between α -cyanobacteria, β -cyanobacteria,
242 and green algae result in different ranges of carbon isotopic fractionation (ϵ_P). α -cyanobacteria



243 containing alpha-carboxysomes and Form 1A RuBisCO are thought to produce a small range of
244 ϵ_P values ($\sim 15.5\text{--}18.9\text{‰}$; Popp et al., 1998). In contrast, β -cyanobacteria containing beta-
245 carboxysomes and Form 1B RuBisCO inhabit a larger range of environments and produce a
246 larger range of ϵ_P values ($\sim 7.7\text{--}22.7\text{‰}$; Hurley et al., 2021). These diverse CCMs can cause
247 variable magnitudes of fractionation under different environmental conditions, as they often
248 involve multiple steps of carbon transport and conversion, which can preferentially select for ^{12}C
249 or ^{13}C depending on the species and the specific pathway utilized (Giordano et al., 2005;
250 Reinfelder, 2011).

251

252 **3 Methods**

253 **3.1 Diurnal engine model framework**

254 To evaluate the influence of diurnal variability on carbonate chemistry and isotopic
255 composition, we employed a modified version of the diurnal carbon engine model presented by
256 Trower et al. (2024) and based on the original description by Geyman and Maloof (2019). This
257 model simulates coupled carbon cycling, gas exchange, and carbonate precipitation over a 24-
258 hour period and is designed to capture transient, non-equilibrium behavior driven by diurnal
259 biological forcing. The model presented here is not intended to exactly reproduce site-specific
260 conditions, but rather to explore mechanistic relationships between environmental forcing and
261 carbonate system dynamics. Specific model modifications from the Trower et al. (2024) code
262 include: utilizing PHREEQC (Parkhurst and Appelo, 1999) for all water geochemistry
263 calculations, as opposed to CO2SYS (Lewis and Wallace, 1998), because CO2SYS is explicitly
264 designed for modern marine seawater and cannot be adapted for non-marine settings; translation



265 from MATLAB into R, a free and open-source language; and model parameter changes
266 appropriate for lacustrine systems. For details about the full set of governing equations and
267 model structure, please refer to Trower et al. (2024). This model is best described as a transient
268 box model, designed to capture evolving, non-equilibrium conditions resulting from the
269 dynamic, time-varying processes described herein. By design, the model is not forced to be
270 mass- or isotope-conservative over each day, so it can be used to explore scenarios in which the
271 system experiences hysteresis (conditions at the end of the day match those at the beginning of
272 the day) and scenarios under which photosynthetic forcing might cause a net change in the
273 aqueous chemistry of the system (conditions at the end of the day do not match those at the
274 beginning of the day).

275 The model represents a well-mixed surface water layer subject to sinusoidal forcing of
276 photosynthetic activity, parameterized by a productivity scaling term (κ_p). This forcing drives
277 diurnal variability in dissolved inorganic carbon (DIC) through photosynthetic uptake during
278 daylight hours and reduced uptake at night. Gas exchange with the atmosphere is included as a
279 continuous process that moderates both DIC concentrations and isotopic composition. Carbonate
280 precipitation is calculated as a function of carbonate mineral saturation state (Ω_{carb}), with
281 precipitation occurring only when $\Omega > 1$. Precipitation rates increase nonlinearly with
282 supersaturation following a kinetic formulation, capturing the strong dependence of carbonate
283 formation on Ω_{carb} values and temperature. As a result, carbonate precipitation is not uniformly
284 distributed through time, but instead preferentially occurs during periods of elevated saturation
285 state. The isotopic composition of DIC ($\delta^{13}\text{C}_{\text{DIC}}$) is tracked dynamically, incorporating
286 fractionation associated with gas exchange, biological uptake, and carbonate precipitation. The
287 isotopic composition of precipitated carbonate is calculated as



288 $\delta^{13}\text{C}_{carb} = \delta^{13}\text{C}_{DIC} + \varepsilon_{carb}$ Eq. 1

289 where ε_{carb} represents the equilibrium fractionation between DIC and carbonate. To quantify the
290 influence of diurnal variability of carbonate isotope signatures, we define the diurnal isotopic
291 offset ($\Delta^{13}\text{C}$) as

292 $\Delta^{13}\text{C} = \overline{\delta^{13}\text{C}_{carb,w}} - (\bar{\delta}^{13}\text{C}_{DIC} + \varepsilon_{carb})$ Eq. 2

293 where $\overline{\delta^{13}\text{C}_{carb,w}}$ is the precipitation-weighted mean isotopic composition of carbonate over the
294 diurnal cycle and $(\bar{\delta}^{13}\text{C}_{DIC} + \varepsilon_{carb})$ is the daily average instantaneous $\delta^{13}\text{C}_{carb}$ value. This
295 formulation quantifies the isotopic bias introduced by preferential carbonate precipitation during
296 daytime enrichment of DIC, relative to the expected $\delta^{13}\text{C}_{carb}$ predicted from the daily mean DIC
297 composition. For comparison with observational data, modeled time series were aligned to
298 observed peak saturation conditions to account for differences in temporal reference frames
299 between modeled and measured datasets.

300

301 **3.2 ε_p according to phytoplankton community and $p\text{CO}_2$**

302 Carbon isotopic fractionation varies substantially among phytoplankton (e.g., Popp et al.,
303 1998; Burkhardt et al., 1999; Marty and Planas, 2008; Brandenburg et al., 2022). Specifically, ε_p
304 varies according to daylength, growth rate, $p\text{CO}_2$, cell geometry, and physiology (Popp et al., 1998;
305 Burkhardt et al., 1999; Brandenburg et al., 2022). For example, ε_p can vary from 1.6–26‰ among
306 diatoms, 17–25‰ among haptophytes, and 9.9–25‰ among freshwater green algae (Popp et al.,
307 1998; Burkhardt et al., 1999; Marty and Planas, 2008).



Many different processes can impact photosynthetic carbon isotope fractionation, with growth rate and CO₂ availability being two of the more commonly discussed variables (e.g., Hollander and McKenzie, 1991; Botz et al., 1996; ŠantRůČková et al., 2000; Zhang et al., 2001; de Kluijver et al., 2014; Penger et al., 2014; Esposito et al., 2019; Hurley et al., 2019, 2021). For example, in Hurley et al. (2021) ϵ_p values (expressed as $\epsilon_p = \delta^{13}C_{CO_2(aq)} - \delta^{13}C_{biomass}$) in a β -cyanobacterium varied largely as a function of CO₂ concentrations. Based on the data presented in Hurley et al. (2021), we used the following equation to estimate ϵ_p (in ‰) for β -cyanobacteria:

$$\epsilon_p = \left(\frac{1}{[CO_2]} * -107.64 \right) + 22.538 \quad \text{Eq. 3}$$

In the event that CO₂ concentration ([CO₂]) falls outside of the range of values reported in Hurley et al. (2021), we adopt the minimum and maximum ϵ_p values reported in Hurley et al. (2021) when [CO₂] is below and above their reported range, respectively. In contrast, ϵ_p in α -cyanobacteria appears to be independent CO₂ availability (Popp et al., 1998). We therefore adopted a single mean value for ϵ_p (17.3‰) for α -cyanobacteria.

For this contribution, we arbitrarily adopt a general ϵ_p value of 14.5‰ for green algae (an average of multiple green algae ϵ_p values reported in the literature) but emphasize that the specific algae in question could lead to large differences in ϵ_p values and therefore modeled carbon isotope values (see **Figure S1** for a model sensitivity test varying ϵ_p from 9.9–25‰). Accurate modeling of the diurnal engine effect for a specific system should incorporate detailed knowledge of the photosynthetic community.



330 3.3 Modeling specific end-member environments

331 We created four different model configurations that incorporate geochemistry from four
332 different environments and three different model phytoplankton taxa. These configurations
333 include: water chemistry from the Turks and Caicos Islands with α -cyanobacteria (model
334 scenario 1); water chemistry from the South Arm of Great Salt Lake with β -cyanobacteria
335 (model scenario 2); water chemistry from Green Lake with green algae (model scenario 3); and
336 estimated water chemistry for Morrison Lake with green algae (model scenario 4). These
337 environments were chosen to represent different end-member environments where the diurnal
338 engine likely operates including a marine carbonate platform (Caicos Platform, Turks and Caicos
339 Islands), a high salinity lake (Great Salt Lake, Utah, USA), a freshwater lake (Green Lake, New
340 York, USA), and a high elevation lake (Morrison Lake, Montana, USA). Furthermore, the
341 phytoplankton communities present in these systems differ, with some being dominated by
342 cyanobacteria (benthic mats and/or planktonic blooms) and others by green algae (benthic
343 charophytes and planktonic green algae) (Culver and Brunskill, 1969; Koch and Madden, 2001;
344 Larsen and Belovsky, 2013). Importantly, each of these systems produces carbonate (aragonite
345 or calcite), allowing for the comparison of the diurnal engine across geochemical conditions and
346 phytoplankton communities. We used mineral precipitation kinetics specific to the primary
347 carbonate mineral forming in each environment.

348

349 3.3.1 Caicos Platform

350 The Caicos Platform (CP) is an archipelago southeast of the Bahamas (21.6940° N,
351 71.7979° W). Because the CP is a marine environment, the carbonate produced there is



352 aragonite, and the salinity of the water is ~35 g/kg. Note that most of the CP is a grainy
353 sediment-dominated environment stabilized by algae (e.g., Wanless and Dravis, 1989; Trower et
354 al., 2018) with modest productivity associated with green algae and patch reefs, while some
355 protected intertidal lagoons host highly productive microbial mat communities with abundant
356 and diverse α -cyanobacteria (e.g., Trembath-Reichert et al., 2016; Gomes et al., 2020; Lingappa
357 et al., 2022; Stein et al., 2023; Trower et al., 2024). When modeling this environment, we used
358 the carbonate and major ion chemistry configuration of Trower et al. (2024) (**Table 1**).

359

360 3.3.2 Great Salt Lake, Utah

361 Great Salt Lake (GSL) is a large hypersaline lake northwest of Salt Lake City, Utah
362 (41.1158° N, 112.4768° W). Its salinity ranges from ~80–280 g/kg and it is a carbonate
363 (aragonite) producing lake. Previous work has suggested that GSL microbial mat communities
364 are dominated by cyanobacteria, the most abundant of which are halophilic coccoidal
365 cyanobacteria *Eubhalothece* sp. (Lindsay et al., 2017; Ingalls et al., 2020). Seasonal planktonic
366 blooms of the green alga *Dunaliella viridis* are also an important mode of primary production in
367 the lake (Baxter, 2018). Note that we are specifically modeling the South Arm of GSL, as the
368 North Arm of GSL has even higher salinity and predominantly photoheterotrophic organisms
369 and the halophilic green alga *Dunaliella salina* (Baxter, 2018). When modeling this
370 environment, we used carbonate and major ion chemistry published by Ingalls et al. (2020)
371 (**Table 1**) and represented primary production as dominated by β -cyanobacteria.

372

373



374 *3.3.3 Green Lake, New York*

375 Green Lake (GL) is a freshwater meromictic lake in Fayetteville, New York (43.0512° N,
376 75.9659° W) in which mixing does not occur, leading to the formation of deep, anoxic bottom
377 waters (Brunskill and Ludlam, 1969). In addition to the photosynthetic microbial communities
378 that contribute to growth of carbonate mounds on the steep basin margins, the calcifying algae,
379 *Chara*, lines the shorelines (Thompson et al., 1990). Calcified thalli of these freshwater
380 autotrophs comprise a significant fraction of the shallow, shoreline sediments (Leapaldt et al.,
381 2024). The GL water column experiences an annual ‘whiting’ event (episodic water column
382 precipitation of fine-grained calcium carbonate) that can last from one to several months
383 (Stanton et al., 2023). The euhedral whiting crystals that form during these events are thought to
384 be produced by a synergy in which the picocyanobacteria *Synechococcus* increases calcite
385 saturation via photosynthesis and the diatoms (*Synechococcus*) and their associated EPS provide
386 nucleation sites for carbonate growth (Stanton et al., 2023). In contrast to GSL (which
387 precipitates aragonite), GL precipitates calcite (Brunskill, 1969). When modeling this
388 environment, we used carbonate and major ion chemistry published by Leapaldt et al. (2024)
389 (**Table 1**) collected in shallow (~1 m depth) water along shoreline adjacent to one of the GL
390 thrombolite mounds.

391

392 *3.3.4 Morrison Lake, Montana*

393 Morrison Lake (ML) is a freshwater alpine lake in southwestern Montana (44.6019° N,
394 113.0367° W). ML is a high elevation lake (~2500 m.a.s.l.) that produces carbonate (calcite).
395 ML is a closed basin lake fed by groundwater and underlain by carbonate bedrock (e.g., Thomas,



2019). These characteristics differentiate ML with GSL and GL, representing a different lake system endmember in terms of climate and seasonality. Furthermore, due to its latitude and elevation, ML is ice-covered for many months each year and likely is characterized by stronger seasonality than GSL and GL. There are no previously published lake chemistry data for ML. As such, we utilized pH data from ML and GL chemistry data as an approximation for ML (both GL and ML are freshwater carbonate-producing lakes) (**Table 1**) (**Data Set S1**). We did, however, manually set alkalinity higher at a mid-point value between the known values for GL and GSL (**Table 1**), as we found it unlikely that alkalinity is as low as GL at ML given the surrounding carbonate bedrock. ML, much like GL, experiences seasonal whittings that contribute carbonate to the sediment column (e.g., Stanton et al., 2023), but Leapaltd et al. (2024) demonstrated that carbonate precipitation occurs outside of whiting events in GL and we consider that this is likely the case for ML as well.

408

409 **3.4 Model sensitivity and seasonality tests**

In addition to the model scenarios described above, we also used model scenario 1 (representing the CP) to conduct model parameter sensitivity tests and model scenario 2 (representing the South Arm of GSL) to conduct phytoplankton community and seasonality tests. For the sensitivity tests, we varied the value of key model parameters over a wide range to assess its impact on key variables, such as [DIC] and $\delta^{13}\text{C}_{\text{carb}}$. The goal of these tests was to identify which model parameters have the largest impact on these key variables and are thus most important to constrain the diurnal carbon engine effect. For the phytoplankton community tests, we used identical GSL chemistry and diurnal carbon engine parameters, only changing the value of ϵ_p according to phytoplankton taxa (see Section 3.2) such that the impact of ϵ_p could be



419 assessed. For the seasonality tests, we modified temperature, wind speed, and the photosynthetic
420 forcing function to simulate different seasons for model scenario 2. We used average daily
421 temperatures and wind speeds reported for GSL in December, March, June, and September to
422 represent winter, spring, summer, and fall, respectively. We modified the photosynthetic forcing
423 function by increasing and decreasing its magnitude during the summer and winter, respectively
424 (the function magnitude was not altered for spring or fall), and by adjusting its period according
425 to the number of daylight hours at that time of year and latitude. We did not have separate
426 geochemistry suites for each season and present this seasonality test as a broad preliminary result
427 that should be followed up with additional data collection and model validation as part of future
428 work.

429

430 **4 Results**

431 **4.1 A flexible, free, and open-source diurnal carbon engine model**

432 Using previously published and best estimates for lake geochemistry (see Methods), we
433 used the diurnal engine model presented herein to model and estimate the magnitude of this
434 effect in these four carbonate-producing settings (CP, GSL, GL, and ML). This diurnal engine
435 model is very similar to the one presented by Trower et al. (2024) with only minor modifications
436 made to handle lake geochemistry (see Methods). This new version is also written in R, making
437 it free and open source to use. Note that this version of the diurnal engine model produced
438 identical results for the marine CP (model scenario 1) as the model presented by Trower et al.
439 (2024) (**Figure 2**).

440



441 **4.2 Diurnal carbon engine model sensitivity tests**

442 We conducted a series of sensitivity tests on model parameters and found that most
443 parameters do not have a strong impact on the resulting magnitude of the diurnal carbon engine
444 effect. However, we find that the photosynthetic forcing (referred to as the κ_p value following
445 Trower et al. (2024) and Geyman and Maloof (2019)) has a strong control on the resulting
446 carbonate chemistry and the magnitude of the isotopic response to the diurnal carbon engine
447 (**Figure 3**). Solution chemistry also plays an important role, with lacustrine settings having
448 displayed a marked difference from equatorial shallow marine settings (the only areas where this
449 effect has been previously documented) (see **Figures 2 and 3**; model scenarios 1–2). Depending
450 on the selected value for κ_p , $\delta^{13}\text{C}_{\text{DIC}}$ values fluctuated by $\geq 4\text{‰}$ in the Turks and Caicos Islands
451 simulation (**Figure 3**). This sensitivity is consistent with observed diurnal variability in natural
452 systems (see Section 4.6; Geyman and Maloof, 2019), where moderate productivity conditions
453 reproduce observed carbonate system amplitudes.

454

455 **4.3 The diurnal carbon engine varies between lacustrine systems**

456 We also found that our diurnal carbon engine model predicts pronounced differences in
457 the projected diurnal engine effect across the three lakes (model scenarios 2–4; **Figures 4 and 5**).
458 Our model scenarios suggested that GSL is likely characterized by a smaller diurnal engine
459 effect, in terms of the $\delta^{13}\text{C}_{\text{carb}}$ response, than GL and ML (**Figure 4**). Our results also suggested
460 that seasonality plays a smaller role in shaping the effect in ML and GSL than in GL (**Figure 4**).
461 However, it is important to note that ML is ice-covered in the winter, which likely alters the
462 diurnal carbon engine; ice-coverage is not incorporated into these model scenarios, so our results



do not consider this effect. It would be valuable to collect field data during ice coverage to better understand how the carbonate system behaves when gas exchange and light penetration are very limited. We also found that the predicted carbonate saturation state (Ω_{carb}) was an order of magnitude higher in GL and ML than in GSL ($\sim 3.5\text{--}7$ in GSL versus $\sim 5\text{--}90$ in GL and ML; **Figure 4**), although some of this difference results from mineralogy (aragonite in GSL; calcite in GL and ML) and ionic strength (ion activities in GSL are much more impacted by complexation due to high ionic strength than in the other lakes). In terms of the modeled carbon fluxes, the photosynthetic flux (F_{photo}) was identical across the model scenarios (as designed; **Figure 5**), but the carbonate precipitation flux (F_{carb}) and the gas exchange flux (F_{gas}) differed dramatically between the model scenarios. In terms of F_{carb} , GSL had relatively low values across the day ($\sim 0\text{--}0.6 \mu\text{mol/kg/hr}$) while GL has values ranging up to $\sim 8 \mu\text{mol/kg/hr}$ and ML had values exceeding $20 \mu\text{mol/kg/hr}$ (**Figure 5**). Importantly, F_{carb} was always positive throughout the entire model day in all three model scenarios, indicating that carbonate was never dissolving (**Figure 5**; which is expected, based on the Ω_{carb} values shown in **Figure 4**). Although carbonate precipitation occurs throughout the diurnal cycle in these simulations, rates vary strongly with saturation state, resulting in preferential precipitation during periods of elevation Ω . Lastly, F_{gas} also differed substantially between these three model scenarios, with values ranging between ~ 0 to $-1.1 \mu\text{mol/kg/hr}$ in GSL, between -1 to $10 \mu\text{mol/kg/hr}$ in GL, and -1 to $1 \mu\text{mol/kg/hr}$ in ML (**Figure 5**). These model results predict that GSL is always a net CO_2 sink (lake $p\text{CO}_2 <$ atmospheric $p\text{CO}_2$), while GL and ML are degassing at least some of the time (lake $p\text{CO}_2 >$ atmospheric $p\text{CO}_2$). In summary, we found large, predicted differences between GSL, GL, and ML (model scenarios 2–4), which all in turn differ from the CP (model scenario 1).



485 We tested the impact of phytoplankton community, in terms of ϵ_p , on the diurnal carbon
486 engine effect by systematically varying ϵ_p while holding all other model parameters constant
487 (**Figure 6**; for these model simulations we used GSL chemistry values). We found that β -
488 cyanobacteria produced the largest diurnal carbon engine effect, driving $\delta^{13}\text{C}_{\text{DIC}}$ 0.46‰ and
489 $\delta^{13}\text{C}_{\text{carb}}$ 0.27‰ more positive than green algae, which resulted in the smallest positive shifts
490 (**Figure 6**). α -cyanobacteria resulted in an intermediate magnitude diurnal carbon engine effect
491 (in terms of $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$; **Figure 6**). These results demonstrate the importance of
492 constraining the phytoplankton community (and its associated ϵ_p value) in order to best estimate
493 the impact of the diurnal carbon engine on $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ records. These differences
494 highlight that the magnitude and temporal structure of the diurnal carbon engine are strongly
495 system-dependent, reflecting differences in buffering capacity and carbonate precipitation
496 dynamics.

497

498 **4.4 The impact of seasonality on the diurnal carbon engine in Great Salt Lake**

499 Many lacustrine systems, both modern and ancient, are not equatorial and are likely
500 characterized by substantial seasonality. As depicted in **Figures 4 and 5**, seasonality does appear
501 to impact the diurnal carbon engine. We conducted additional seasonality simulations for GSL
502 (**Figure 7**), as we have more extensive geochemical and meteorological data for GSL in
503 comparison with GL and ML. In these simulations, we varied temperature ($T = 25^\circ\text{C}$, 18°C , -
504 3°C , and 5°C ; based on historical data for Salt Lake City, UT, from WeatherSpark), wind speed
505 ($u = 4.1$ m/s, 3.8 m/s, 3.3 m/s, and 4.3 m/s; based on historical data for Salt Lake City, UT, from
506 WeatherSpark), and productivity ($\kappa_p = 190$, 150 , 110 , and 150 $\mu\text{mol/kg}$; estimates based on
507 Trower et al., 2024) according to season (summer, autumn, winter, and spring, respectively;



508 **Figure 7**). In terms of the modeled fluxes, summer had greatest magnitudes of F_{photo} values,
509 followed by autumn and spring (identical, by design), and then winter (**Figure 7**); these
510 differences are expected based on the chosen values for κ_p . Summer had the highest F_{carb} values,
511 followed by autumn, and spring and winter (which were nearly identical; **Figure 7**). Lastly, for
512 F_{gas} , winter had the largest magnitude flux, followed closely by spring, and then autumn and
513 summer, which had significantly smaller fluxes (**Figure 7**). This finding suggests that, for our
514 model conditions, GSL is always a CO_2 sink, even considering seasonality, but this net
515 absorption of CO_2 is considerably stronger in the winter and spring months. In terms of the
516 carbon isotopic response to a seasonally variable diurnal carbon engine in GSL, summer is
517 always characterized by larger diurnal changes in both $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ values, followed by
518 autumn, and then very closely spring, and lastly winter (**Figure 7**). Note that the autumn and
519 spring carbon isotopic responses are nearly indistinguishable, and that the net carbon isotopic
520 response is only positive during the summer when the diurnal engine is strongest (**Figure 7**).
521 Overall, our model simulations suggested that seasonality accounts for up to $\sim 0.28\text{‰}$ and
522 $\sim 0.10\text{‰}$ of variability in $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ values in GSL, respectively. Note that the initial
523 $\delta^{13}\text{C}_{\text{DIC}}$ value does not exert a strong control on these results, with a sensitivity test indicating a
524 $\sim 0.05\text{‰}$ and $\sim 0.01\text{‰}$ change in the seasonal variability of $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ values,
525 respectively, for each 1‰ shift in the initial $\delta^{13}\text{C}_{\text{DIC}}$ value (see **Figure S2**).

526

527 **4.5 The diurnal carbon engine modulates $\delta^{13}\text{C}_{\text{carb}}$ records**

528 Previous work has documented the potential importance of the diurnal carbon engine in
529 shaping marine $\delta^{13}\text{C}_{\text{carb}}$ records (Geyman and Maloof, 2019; Trower et al., 2024). Our results
530 have demonstrated that the same is true for lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records (**Figures 2–8**). Assuming



constant major ion chemistry and summer temperature (25°C), we projected the carbon isotopic impact of the diurnal carbon engine across a range of productivity and alkalinity values in GSL (precipitating aragonite) and GL (precipitating calcite) (**Figure 8**). Aragonite (or calcite) precipitated with typical fractionation from DIC and the predicted aragonite (or calcite) precipitated with an active diurnal carbon engine differ. Using a modest productivity level ($\kappa_p = 150 \mu\text{mol/kg}$) and modern alkalinity for each lake (see red dots in the contour plots shown in **Figure 8**), we then estimated the predicted isotopic offset ($\Delta^{13}\text{C}$), defined as the difference between precipitation-weighted mean $\delta^{13}\text{C}_{\text{carb}}$ and the daily average instantaneous $\delta^{13}\text{C}_{\text{carb}}$ value (expected daily variability). This $\Delta^{13}\text{C}$ offset ranged from ~ 0 – 0.9‰ in GSL and ~ 0 – 3.9‰ in GL, with a much stronger dependence on productivity (κ_p) than alkalinity (**Figure 8**). For GSL, we calculated a precipitation-weighted mean $\delta^{13}\text{C}_{\text{carb}}$ value of $\sim 2.81\text{‰}$ and found an estimated $\Delta^{13}\text{C}$ offset of 0.03‰ (**Figure 8**). For GL, we calculated a precipitation-weighted mean $\delta^{13}\text{C}_{\text{carb}}$ value of $\sim -5.64\text{‰}$ and found an estimated $\Delta^{13}\text{C}$ offset of 0.27‰ (**Figure 8**). Note that we could not conduct these same tests for ML due to a lack of $\delta^{13}\text{C}_{\text{DIC}}$ data, but future field and lab work could allow for ML to be compared with GSL and GL.

4.6 Comparison to observed diurnal carbonate dynamics

To evaluate whether the model reproduces observed diurnal carbonate system behavior, we compared modeled diurnal variability in calcite saturation state (Ω) with digitized observations from Green Lake (Effler et al., 1981) (**Figure 9**). The model reproduces the timing and overall structure of the observed diurnal cycle, including nighttime minima and daytime maximum associated with photosynthetic CO_2 uptake. The magnitude of variability is sensitive to the productivity parameter (κ_p), with moderate values ($\kappa_p = 100$) producing amplitudes



554 comparable to observed summer conditions, while higher values produce larger amplitudes
555 (**Figure 9**). This comparison demonstrates that the model captures first-order controls on
556 biologically mediated carbonate chemistry in lacustrine systems while highlighting the
557 importance of productivity in regulating diurnal variability.

558

559 **5 Discussion**

560 **5.1 Timing of carbonate precipitation**

561 While it is difficult to know when most carbonate is precipitating throughout the year for
562 a given lake, it is clear that it is not only during seasonal whiting events (e.g., Stanton et al.,
563 2023; Leapaldt et al., 2024). Our results indicate that the magnitude of isotopic offsets is
564 controlled not simply by the extent of diurnal variability, but also by the timing of carbonate
565 precipitation relative to that variability. Because precipitation rates are highest during periods of
566 elevated saturation state, carbonate formation is inherently biased toward daytime conditions.
567 Even if carbonate precipitation did only occur over a few weeks per year during these whiting
568 events, diurnal variability would still matter during those events and ultimately influence $\delta^{13}\text{C}_{\text{carb}}$
569 values. If all carbonate precipitation occurred during a whiting (or ‘bloom’), κ_p would likely be
570 large at this time, meaning that the diurnal engine would result in an even stronger impact on
571 $\delta^{13}\text{C}_{\text{carb}}$ values than it would if averaged out over the course of an entire year. Our standard
572 model simulation assumes a temperature of 25°C meant to simulate summer conditions (unless
573 otherwise stated, like for the seasonality tests). This is because in lacustrine settings, most carbon
574 precipitates in the summer and precipitation rates are faster at warmer temperatures (e.g., Burton
575 and Walter, 1987; Lopez et al., 2009; Romanek et al., 2011). Due to temperature–productivity



relationships (Konopka and Brock, 1978), κ_p would also likely be higher in summer. For these reasons, summer is when the diurnal engine is most likely to have an impact on sediment $\delta^{13}\text{C}_{\text{carb}}$ values.

5.2 Cryptic $\delta^{13}\text{C}_{\text{carb}}$ variability

Overall, we found large, predicted differences between CP, GSL, GL, and ML (model scenarios 1–4), which demonstrates that solution geochemistry, phytoplankton community, and seasonality all play important roles in shaping the diurnal carbon engine effect and thus $\delta^{13}\text{C}_{\text{carb}}$ records. A change in environmental conditions, as recorded in sedimentary lithofacies, could result in a change in the overall productivity of that environment, ultimately shifting κ_p and modifying the local diurnal carbon engine. Because κ_p reflects net photosynthetic forcing, it likely integrates multiple environmental controls such as light availability, nutrient supply, and phytoplankton community composition. Likewise, an ecological change resulting in a shift in the dominant phytoplankton community could change background isotopic fractionation (ϵ_p). Therefore, our results indicate that both environmental and ecological change (internal lacustrine system dynamics) could drive cryptic fluctuations in bulk $\delta^{13}\text{C}_{\text{carb}}$ values that are not related to sympathetic fluctuations in long-term average $\delta^{13}\text{C}_{\text{DIC}}$ values (external global dynamics; as is often invoked when interpreting $\delta^{13}\text{C}_{\text{carb}}$ records). As our modeling results demonstrated, the diurnal carbon engine can vary substantially between different lake environments due to lake geochemistry and phytoplankton community. As a result, instantaneous $\delta^{13}\text{C}_{\text{carb}}$ values vary throughout the course of each day (due to the diurnal carbon engine) and time of year (due to seasonality). This scale of temporal variability makes it difficult to tie sedimentary samples to specific environmental conditions. However, in some cases it may be possible to use proxies



(e.g., different isotopic signatures, lipid biomarkers) to better understand depositional environments and constrain diurnal engine impact at a given locality (e.g., Klatt et al., 2025; Olsen-Valdez et al., 2025). These findings underline the need to interpret lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records in the context of changes in sedimentary facies, phytoplankton community, and seasonality, as κ_p (and therefore diurnal carbon engine strength) varies according to these environmental-ecological parameters and modulates $\delta^{13}\text{C}_{\text{carb}}$ records. Given the differences we predict in the diurnal carbon engine across these three lake systems, our findings also emphasize the need to ground truth the diurnal carbon engine in many different lake systems to establish suitable analogs for paleolacustrine $\delta^{13}\text{C}_{\text{carb}}$ record interpretation: κ_p is an empirical parameter that, to date, has only been estimated for zooxanthellate reefs and marine green algae in shallow tropical marine environments.

Comparing predicted $\delta^{13}\text{C}_{\text{carb}}$ values with those measured from GSL and GL, we can begin to better interpret these $\delta^{13}\text{C}_{\text{carb}}$ records. For GSL, our diurnal carbon engine model predicts typical $\delta^{13}\text{C}_{\text{carb}}$ values of $\sim 2.8\text{‰}$, based on modern lake chemistry and a modest κ_p estimate of $150 \mu\text{mol/kg}$ (**Figure 8**). Comparing this prediction with the sedimentary facies-specific $\delta^{13}\text{C}_{\text{carb}}$ values reported by Ingalls et al. (2020), we find that the prediction compares reasonably well with $\delta^{13}\text{C}_{\text{carb}}$ values measured from ooid-dominated facies (**Figure 8**). However, our predicted $\delta^{13}\text{C}_{\text{carb}}$ value does not compare well with those reported from mud-rich facies and carbonate coatings on anthropogenically-delivered silicate sand grains and microbialite samples (**Figure 8**). The ooid isotopic range reported by Ingalls et al. (2020) is a bit higher than our estimate ($\sim 4\text{‰}$ vs $\sim 2.8\text{‰}$), and this is likely because ooids form slowly over long periods of time; in GSL, ooids have been accumulating over the past ~ 6000 years, with most of the carbonate precipitating prior to the Industrial Revolution (Paradis et al., 2024). For this reason,



the comparison of the $\delta^{13}\text{C}_{\text{carb}}$ values of these samples with modern $\delta^{13}\text{C}_{\text{DIC}}$ is complicated by the ^{13}C Suess Effect (Keeling, 1979), the phenomenon that the $\delta^{13}\text{C}$ value of atmospheric CO_2 (and DIC in surface fluids that equilibrate with the atmosphere) have increased due to anthropogenic carbon release. Atmospheric CO_2 $\delta^{13}\text{C}$ values have declined by about 2‰ since 1850 (Graven et al., 2020), implying that $\delta^{13}\text{C}_{\text{DIC}}$ values could have been 2‰ higher prior to 1850 when the carbonate in GSL ooids was precipitating. If we run the GSL model with an initial $\delta^{13}\text{C}_{\text{DIC}}$ value that is 2‰ higher, the model prediction ($\sim -4.78\text{‰}$) compares better with the observed $\delta^{13}\text{C}_{\text{carb}}$ values of GSL ooids ($\sim -4\text{‰}$). The microbialite sample $\delta^{13}\text{C}_{\text{carb}}$ values are even more positive than the ooid grainstones, and while the ^{13}C Suess Effect may also play a role in elevating these $\delta^{13}\text{C}_{\text{carb}}$ values, it cannot explain the larger magnitude of the positive offset of these samples. Instead, we interpret that it is more likely that microbial mats locally elevate productivity (and therefore κ_p), meaning that our relatively modest κ_p value of $150 \mu\text{mol/kg}$ is too low to accurately predict the $\delta^{13}\text{C}_{\text{carb}}$ values of these samples. Increasing κ_p would result in more comparable predicted $\delta^{13}\text{C}_{\text{carb}}$ values, and indeed, could allow us to place constraints on the productivity levels present when these microbialite samples precipitated.

In GL, measured $\delta^{13}\text{C}_{\text{carb}}$ values from a mud-rich charophyte setting and from a microbialite setting range from ~ -4 to -1‰ (Leapaldt et al., 2024). Both settings are relatively productive (dense charophyte beds and microbialite bioherms), so it is not surprising that the measured $\delta^{13}\text{C}_{\text{carb}}$ values are more positive than those predicted by our model ($\sim -5.64\text{‰}$; this was also the case for the microbialite samples from GSL). To date, very few age constraints exist for the carbonate grains in GL, so we cannot directly constrain how much of this positive measured-predicted offset could be the result of the ^{13}C Suess Effect. We suspect that the ^{13}C Suess Effect plays an important role in shaping GL $\delta^{13}\text{C}_{\text{carb}}$ values, accounting for at least some



645 of the positive measured-predicted $\delta^{13}\text{C}_{\text{carb}}$ offset. As was the case for the measured GSL $\delta^{13}\text{C}_{\text{carb}}$
646 values, we interpret that it is likely that these charophyte and microbialite environments locally
647 elevate productivity (and κ_p) and that our modest κ_p value of $150 \mu\text{mol/kg}$ is too low to
648 accurately predict these GL $\delta^{13}\text{C}_{\text{carb}}$ values. Takahashi et al. (1968) present a dated carbonate
649 core from GL, which suggests a reservoir age on the order of ~ 6000 yr. However, it is possible
650 that this age reflects a deeper part of the lake which could have a different reservoir age (water
651 depth at which the sediment core was collected is not reported), while the carbonate we are
652 comparing our model predictions with formed in shallow depths (Leapaldt et al., 2024).
653 Additional radiocarbon constraints from GL would be helpful in better understanding this
654 system.

655 Because of the lack of $\delta^{13}\text{C}_{\text{DIC}}$ data from ML, we cannot confidently interpret the
656 resulting $\delta^{13}\text{C}_{\text{carb}}$ values. Surface sediments from ML have been measured to have a $\delta^{13}\text{C}_{\text{carb}}$
657 value of -1.36‰ (S. Schoenemann, unpublished data), though we do not have the $\delta^{13}\text{C}_{\text{DIC}}$ and
658 major ion chemistry measurements needed to accurately model the diurnal engine effect in ML.
659 These additional geochemical data would greatly improve our understanding of ML. Comparison
660 of modeled diurnal carbonate dynamics with observations from Green Lake further supports our
661 interpretations (**Figure 9**). The model reproduces the timing and structure of observed diurnal
662 variability, while the magnitude of variability is sensitive to productivity. This agreement
663 suggests that the processes captured by this model are representative of real lacustrine systems,
664 and that variability in productivity can explain differences in observed diurnal amplitude.

665

666



667 **5.3 Model equilibrium**

668 Depending on model scenario, $\delta^{13}\text{C}$ values do not always return to their starting values by
669 the end of a diurnal cycle (e.g., Figure 4). This suggests that a baseline shift occurred during the
670 day, and depending on model set up, could lead to a runaway effect if the model were run for
671 many days. There are multiple explanations for this, depending on the specific model scenario.

672 First, the input model parameters may not have represented equilibrium (or realistic)
673 conditions, forcing the model to end at different values than it started at. For example, the
674 photosynthetic forcing function in diurnal engine models is not well constrained, as discussed
675 earlier, and model scenarios with an especially strong photosynthetic forcing typically do not
676 return to initial $\delta^{13}\text{C}$ values. Thus, a strong photosynthetic forcing could either be realistic and
677 anomalous (e.g., a whiting event) for a given system, leading to a baseline geochemistry shift, or
678 it could represent an unrealistic value for the given system, meaning that the predicted shift in
679 geochemistry is also unrealistic (assuming the initial geochemistry is accurate). As another
680 example, $\delta^{13}\text{C}_{\text{DIC}}$ values in lakes can vary seasonally, with higher values in the summer
681 (commonly due to some combination of enhanced productivity and/or evaporation) than in other
682 seasons; this pattern has been observed in Green Lake (Leapaldt et al., 2024) and a variety of
683 other lakes (Mybro and Shapley, 2006; Bade et al., 2004; Hodell et al., 2003). [DIC] values and
684 other aspects of lake water chemistry (i.e., major ion concentrations, which in turn reflect
685 differences in salinity and alkalinity) can also vary seasonally; such variation has been
686 documented in both Great Salt Lake and Green Lake (Ingalls et al., 2020; Leapaldt et al., 2024).
687 Here, we chose the same initial conditions for our different tests of seasonality and
688 phytoplankton community to most clearly show the impacts of these different variables on the
689 magnitude of diurnal changes. Neglecting such seasonal changes (which are as yet not



adequately constrained in any of the example lakes to model appropriately) likely contributes to some of the predictions of baseline $\delta^{13}\text{C}_{\text{DIC}}$ values, which might be resolved with incorporation of better field data for a particular lake.

Second, it is important to note that there are biogeochemical processes not included in this model that can have an impact on $\delta^{13}\text{C}$ values, and that this model is, by design, not in perfect equilibrium (this is also true for all existing models of the diurnal engine; Trower et al., 2024; Geyman and Maloof, 2019). Evaporation, for example, could have a measurable impact on $\delta^{13}\text{C}$ values in these systems by further increasing $\delta^{13}\text{C}_{\text{DIC}}$ values during the daytime (e.g., Horton et al., 2016; Stiller et al., 1985), but it is not incorporated into this diurnal carbon cycling model. Similarly, mixing, including wind-driven mixing of waters between areas of differing productivity and/or water chemistry or seasonal mixing of stratified waters, would also influence the pattern of $\delta^{13}\text{C}$ values and is not included in the model scenarios presented here. Assuming processes like mixing and evaporation are sufficiently well constrained in a given lake system, their effects could easily be integrated with the diurnal carbon cycling model presented here in future implementations. While these caveats are important, we believe that the model results presented here still provide important and realistic constraints on lacustrine diurnal cycling, and specifically its impact on $\delta^{13}\text{C}_{\text{carb}}$ values, in most instances.

5.4 Implications for $\delta^{13}\text{C}_{\text{carb}}$ interpretations

Our findings predict $\delta^{13}\text{C}_{\text{carb}}$ offsets driven by the diurnal carbon engine, which differ according to productivity, lake geochemistry, and phytoplankton community (**Figure 8**). Our model results predict ~0.10‰ of variability in $\delta^{13}\text{C}_{\text{carb}}$ in GSL due to seasonality (**Figure 7**), and



712 we found $\Delta^{13}\text{C}$ offsets of 0.03‰ and 0.27‰ for GSL and GL, respectively (**Figure 8**). Therefore,
713 many different variables (productivity, seasonality, water geochemistry, and phytoplankton
714 community) conspire together to contribute variability to bulk $\delta^{13}\text{C}_{\text{carb}}$ signals, which essentially
715 time-averages this variability. Figure 8 demonstrates that large diurnal variability does not
716 necessarily produce isotopic bias; rather, bias emerges only where carbonate precipitation is
717 sufficiently active and temporally structured. Thus, along with seasonal variability (**Figure 7**),
718 both temporal forcing and system properties jointly control the magnitude of the diurnal carbon
719 engine. This complicates point-to-point $\delta^{13}\text{C}_{\text{carb}}$ correlations (e.g., Hay et al., 2019; Hagen et al.,
720 2024), as the absolute $\delta^{13}\text{C}_{\text{carb}}$ values may not be correlative. While normalized $\delta^{13}\text{C}_{\text{carb}}$
721 correlation may help address this problem, a new $\delta^{13}\text{C}_{\text{carb}}$ correlation scheme that can incorporate
722 uncertainty in individual $\delta^{13}\text{C}_{\text{carb}}$ values would significantly improve correlation efforts with both
723 marine and lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records (see discussion in Hagen and Creveling (2024)). In some
724 cases, it may be possible to tailor stratigraphic sampling resolution to account for suspected
725 productivity, geochemistry, and phytoplankton in hopes of better capturing $\delta^{13}\text{C}_{\text{carb}}$ variability.

726 In terms of the predicted $\delta^{13}\text{C}_{\text{carb}}$ offsets, several studies have reported $\delta^{13}\text{C}_{\text{carb}}$ values in
727 modern lacustrine carbonate microbialites that are up to 4‰ higher than expected for carbonate
728 minerals forming in isotopic equilibrium with lake water DIC (Warden et al., 2016; White et al.,
729 2021). This is consistent with the idea that the diurnal carbon engine is driving $\delta^{13}\text{C}_{\text{carb}}$ offsets in
730 carbonate-producing lakes. Comparison of previously published $\delta^{13}\text{C}_{\text{carb}}$ data from GSL and GL
731 reveal a similar observation: in both lakes, observed $\delta^{13}\text{C}_{\text{carb}}$ values of surface sediment are
732 higher than predicted for carbonate minerals forming in isotopic equilibrium with modern lake
733 water DIC (e.g., Ingalls et al., 2020; Leapaltd et al., 2024). While it is possible that this offset
734 could be partially related to a time-averaging effect (i.e., surface sediments are old and



735 precipitated from lake water DIC that was substantially different from modern), we hypothesize
736 that these data reveal the diurnal engine effect at play in both lakes. Additional data from ML
737 would be very valuable, as it likely has a very small reservoir age correction (being shallow and
738 well-mixed), eliminating the time-averaging issue.

739

740 **6 Conclusions**

741 In this contribution, we presented a diurnal carbon engine model framework to evaluate
742 how diurnal variability and the timing of carbonate precipitation influence lacustrine $\delta^{13}\text{C}_{\text{carb}}$
743 values. In doing so, we demonstrated a large amount of variability in the diurnal carbon engine
744 across lake systems and over the course of a year. Importantly, we show that isotopic offsets are
745 governed not simply by the magnitude of diurnal variability, but by the timing of carbonate
746 precipitation relative to that timing. Differences in lake geochemistry and dominant
747 phytoplankton community appear to play a role in controlling the diurnal carbon engine and its
748 impact on $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ values. Both environmental and ecological changes can result in
749 cryptic $\delta^{13}\text{C}_{\text{carb}}$ variability that is not directly linked to changes in long-term $\delta^{13}\text{C}_{\text{DIC}}$. We estimate
750 that seasonality accounts for $\sim 0.28\text{‰}$ and $\sim 0.10\text{‰}$ of variability in $\delta^{13}\text{C}_{\text{DIC}}$ and $\delta^{13}\text{C}_{\text{carb}}$ in GSL,
751 and that $\Delta^{13}\text{C}$ offsets are likely in excess of 0.03‰ and 0.27‰ in $\delta^{13}\text{C}_{\text{carb}}$ for GSL and GL,
752 respectively. This variability may complicate $\delta^{13}\text{C}_{\text{carb}}$ correlations and the construction
753 chronostratigraphic frameworks (see discussion in Hagen, 2024). Previously published $\delta^{13}\text{C}_{\text{carb}}$
754 data from carbonate-producing lakes are consistent with these findings, with comparable offsets
755 between microbialites, surface sediment, and lake water DIC. These findings demonstrate that
756 the interpretation of lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records requires explicit consideration of both
757 environmental variability and the timing of carbonate formation. While we provided a set of end-



member model scenarios, more lake environments should be incorporated into this modeling framework to provide more modern analogs for lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records. Ground truthing our example lakes (GSL, GL, and ML) and additional lake environments will require extensive field campaigns collecting high temporal resolution water chemistry datasets over ≥ 24 -hour periods. Furthermore, better constraints on κ_p are critical, as this parameter exerts primary control on diurnal carbon engine strength but has thus far only been estimated in marine systems. While a difficult parameter to constrain, it is necessary to estimate κ_p for each lake system with observational data. Ultimately, this framework provides a foundation for developing proxy system models capable of predicting $\delta^{13}\text{C}_{\text{carb}}$ records in paleolake environments and improving interpretation of these key paleoclimate archives.

768

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780

781 **Author contributions**

782 C.J.H. and E.J.T. conceived the study, obtained funding, and developed the model; C.J.H.
783 conducted model experiments with input from E.J.T, S.J.H., and A.H.; All authors contributed to
784 discussions and manuscript review; C.J.H. wrote and edited the manuscript with input from
785 E.J.T., S.J.H., A.H., S.J.H., S.W.S., M.I., and K.E.S. S.W.S. provided necessary data from
786 Morrison Lake.

787



788 **Supplementary Information**

789 A Supplementary Information document includes two additional model output figures (Figures
790 S1–S2), and a pH dataset from Morrison Lake, MT (Data Set S1).

791

792 **Code and Data Availability**

793 The model code used in this contribution is free and open-source, available in Zenodo: Hagen
794 and Trower (2026; doi:10.5281/zenodo.20128480).

795

796 **Competing Interests**

797 The authors declare no conflicts of interest.

798



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1094 Tables

1095 **Table 1. Major ion and carbonate chemistry for the different model scenarios**

Model scenario	Ca (mmol/kg)	Mg (mmol/kg)	Na (mmol/kg)	K (mmol/kg)	Cl (mmol/kg)	SO ₄ (mmol/kg)	DIC (mmol/kg)	pH	Alkalinity (mmol/kg)	$\delta^{13}\text{C}_{\text{DIC}}$ (‰)	$\delta^{13}\text{C}_{\text{org}}$ (‰)
Turks and Caicos Platform	10.2	52.6	464.5	10.1	540.5	27.9	2.1	N/a	2.4	0.7	-8.0
Great Salt Lake (South Arm)	4.5	156.0	1433.0	58.0	1659.0	78.0	3.7	N/a	4.6	-0.7	-8.0
Green Lake	9.8	2.5	1.2	0.1	0.6	9.8	2.5	N/a	2.4	-7.0	-14.3
Morrison Lake	9.8	2.5	1.2	0.1	0.6	9.8	N/a	7.6 (W)–8.7 (S)	3.5	-7.0	-14.3

1096

1097 Note that major ion chemistry values for Morrison Lake were adopted from Green Lake as these
 1098 data have not been measured or published. For the Morrison Lake pH data, W refers to winter
 1099 values and S refers to summer values. Morrison Lake pH data are available in Data Set S1.

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1102 **Figures**

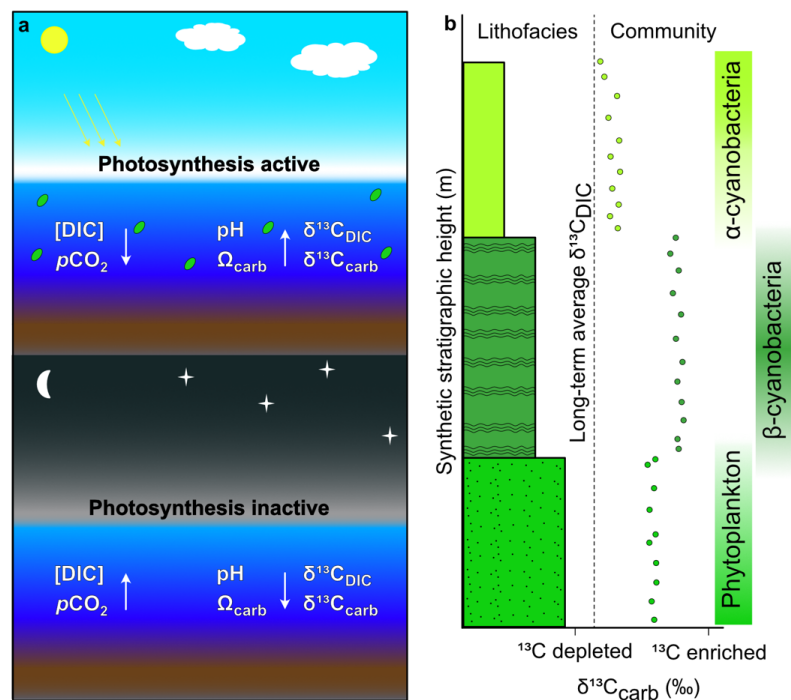


Figure 1. Theoretical impact of the diurnal carbon engine on $\delta^{13}\text{C}_{\text{carb}}$ records. **a**, a schematic depicting how the diurnal engine modulates carbonate geochemistry based on time of day. **b**, a theoretical stratigraphic column and corresponding precipitated $\delta^{13}\text{C}_{\text{carb}}$ record indicating how sedimentary lithofacies and the dominant phytoplankton community could drive fluctuations in $\delta^{13}\text{C}_{\text{carb}}$ values (modifying the local diurnal carbon engine) that do not reflect changes in long-term average $\delta^{13}\text{C}_{\text{DIC}}$ values but instead environmental/ecological change. Note that in **b** the shade of green indicates both sedimentary lithofacies and dominant phytoplankton community.

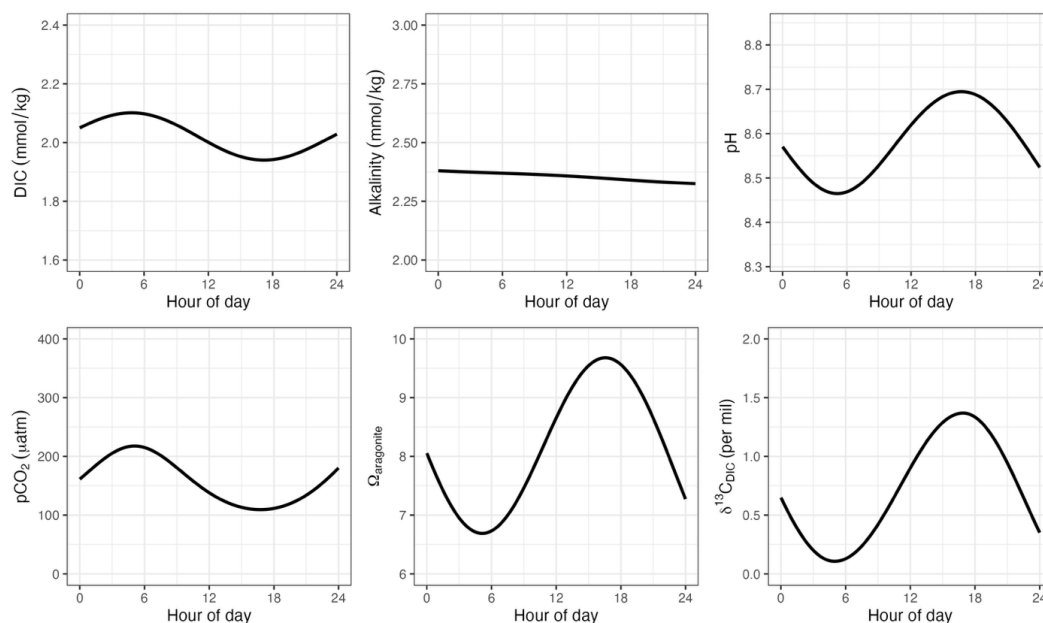
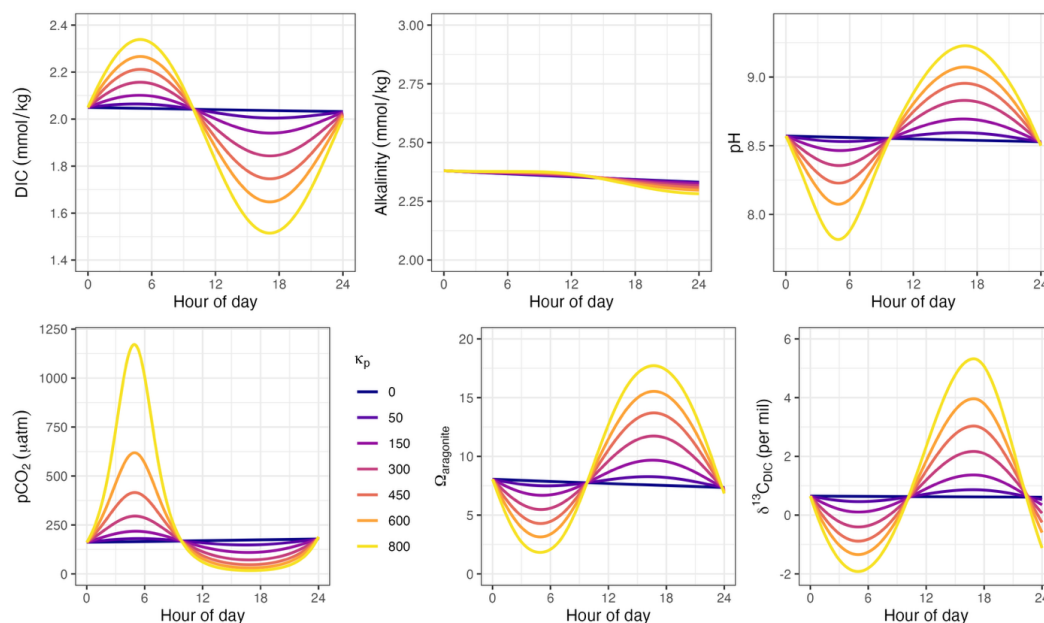


Figure 2. An open-source diurnal carbon engine model adapted to handle lake geochemistry. The diurnal carbon engine model (adapted from Trower et al. (2024)), with modifications to how geochemistry calculations are made to allow for lacustrine datasets. Diurnal variability in dissolved inorganic carbon concentration ([DIC]), alkalinity (Alk), pH, $p\text{CO}_2$, aragonite saturation state (Ω_{ar}), and the carbon isotopic value for DIC ($\delta^{13}\text{C}_{\text{DIC}}$) are plotted here for the Caicos Platform. These model outputs are identical to those detailed in Trower et al. (2024). Hour 0 represents 00:00 (midnight).



1123

1124 **Figure 3. A sensitivity test exploring the impact of productivity on the diurnal carbon engine.**

1125 K_p (reflecting photosynthetic forcing strength) is a key parameter in shaping the photosynthetic
 1126 carbon flux in the diurnal carbon engine model yet has only been constrained in marine settings
 1127 (see Trower et al., 2024). We vary the value of K_p from 0–800 (see pCO_2 panel for color-coded
 1128 legend) for the Caicos Platform and find that greater values of K_p (and therefore a stronger diurnal
 1129 carbon engine) lead to a greater response in carbonate geochemistry parameters including
 1130 dissolved inorganic carbon concentration ([DIC]), pH, pCO_2 , aragonite saturation state (Ω_{ar}), and
 1131 the carbon isotopic value for DIC ($\delta^{13}C_{DIC}$), with the exception of alkalinity (alk). Hour 0
 1132 represents 00:00 (midnight).

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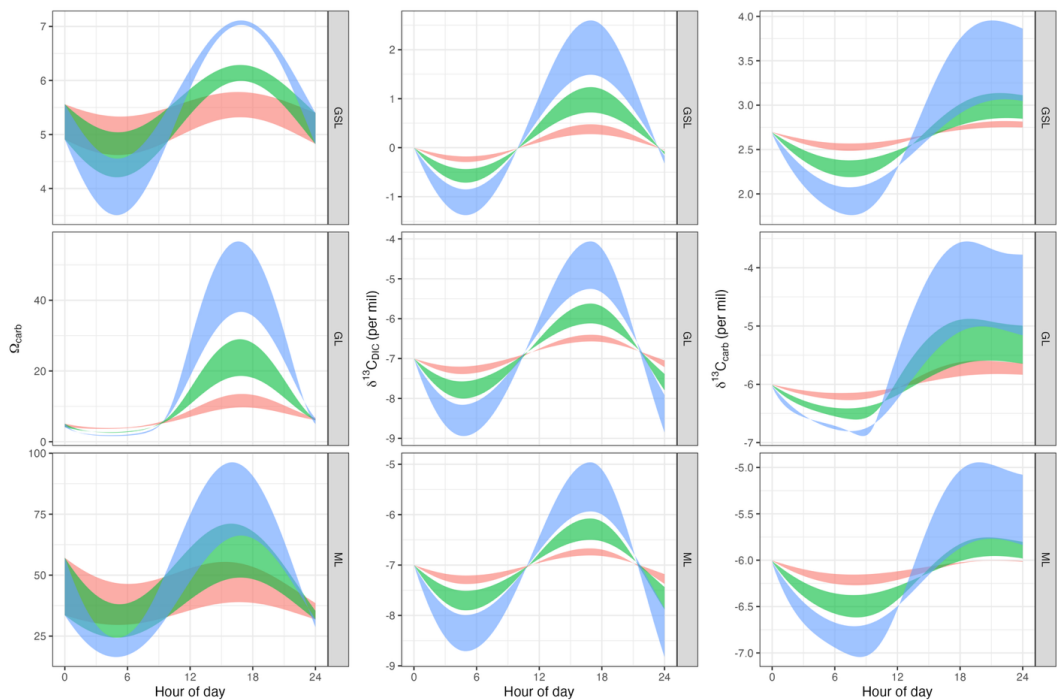


Figure 4. Lake chemistry, seasonality, and phytoplankton community affect the strength of the diurnal carbon engine. The impact of the diurnal carbon engine on carbonate saturation state (Ω_{carb} ; left column), the carbon isotope value of dissolved inorganic carbon ($\delta^{13}\text{C}_{\text{DIC}}$; center column), and the cumulative carbon isotope value of carbonate ($\delta^{13}\text{C}_{\text{carb}}$; right column) according to lake chemistry, seasonality, and phytoplankton community is shown in nine panels. Panels are positioned according to lake, with those in the top row Great Salt Lake (South Arm; model scenario 2), those in the middle row representing Green Lake (model scenario 3), and those in the bottom row representing Morrison Lake (model scenario 4). Each panel depicts three model envelopes: $K_p = 150$ in red, $K_p = 375$ in green, and $K_p = 750$ in blue. The upper bound of each model envelope corresponds to summer conditions, while the lower bound corresponds to winter conditions with reduced temperature (5°C versus 25°C) and photosynthetic forcing (60% of summer forcing). In every scenario, $\delta^{13}\text{C}_{\text{carb}}$ values (right column) do not return to their initial values, indicating the change in the isotopic value of the accumulated carbonate over the course of the day. Hour 0 represents 00:00 (midnight).

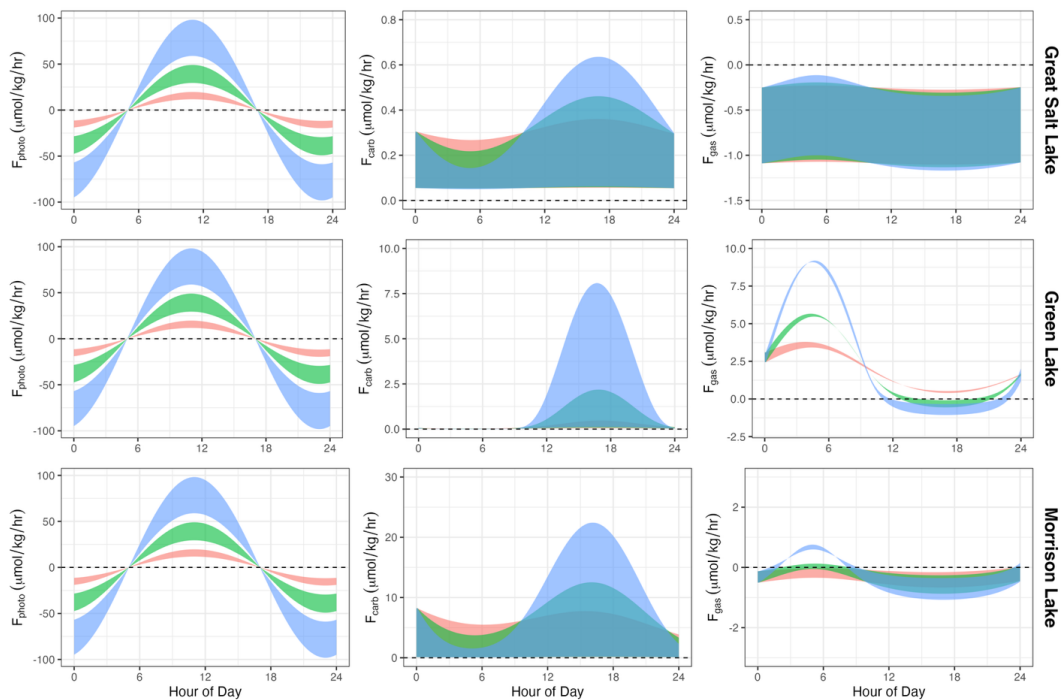
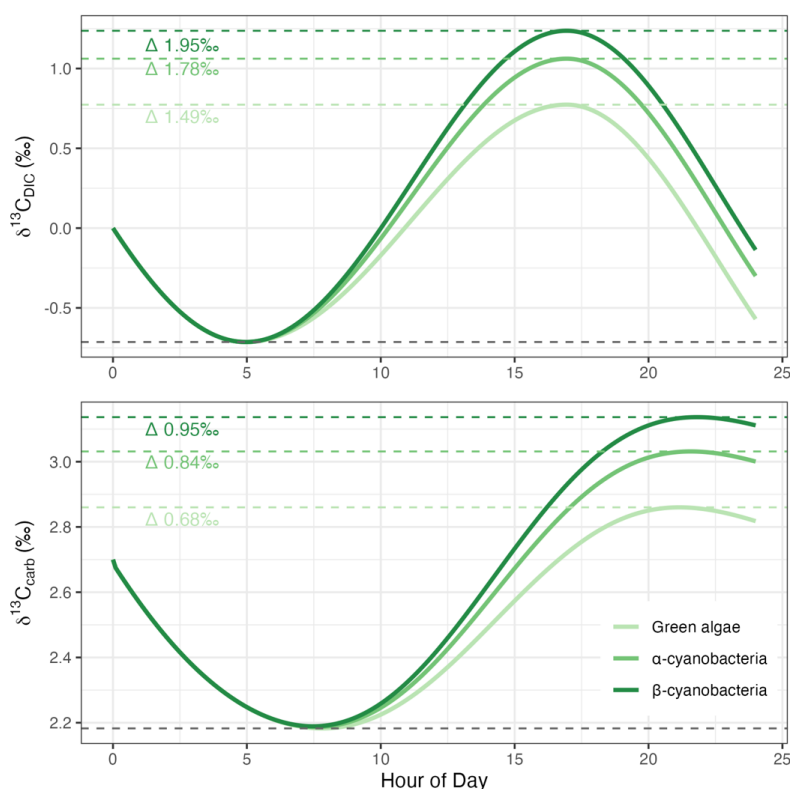


Figure 5. Resulting carbon fluxes when modeling the impact of lake chemistry, seasonality, and phytoplankton community on the diurnal carbon engine. Modeled photosynthetic flux (F_{photo} ; left column), carbonate precipitation flux (F_{carb} ; center column), and gas exchange flux (F_{gas} ; right column) according to lake chemistry, seasonality, and phytoplankton community is shown in nine panels. Panels are positioned according to lake chemistry as in Figure 4, with those in the top row representing Great Salt Lake (South Arm; model scenario 2), those in the middle row representing Green Lake (model scenario 3), and those in the bottom row representing Morrison Lake (model scenario 4). As in Figure 4, each panel depicts three model envelopes: $K_p = 150$ in red, $K_p = 375$ in green, and $K_p = 750$ in blue. The upper bound of each model envelope corresponds to summer conditions, while the lower bound corresponds to winter conditions with reduced temperature (5°C versus 25°C) and photosynthetic forcing (60% of summer forcing). Hour 0 represents 00:00 (midnight).



1167

1168 **Figure 6. Impact of dominant phytoplankton community on the isotopic response to the**
 1169 **diurnal carbon engine.** Using Great Salt Lake geochemistry, we varied the ϵ_p according to
 1170 phytoplankton community to simulate the impact of different communities (β -cyanobacteria, α -
 1171 cyanobacteria, and green algae; see legend) on the diurnal carbon engine. Due to the associated ϵ_p ,
 1172 α -cyanobacteria resulted in the highest $\delta^{13}\text{C}_{\text{DIC}}$ (above) and $\delta^{13}\text{C}_{\text{carb}}$ values (below) and the
 1173 strongest diurnal carbon engine effect (with green algae and β -cyanobacteria resulting in a
 1174 progressively smaller impact, respectively). The impact of the phytoplankton community on
 1175 $\delta^{13}\text{C}_{\text{DIC}}$ values is larger than on $\delta^{13}\text{C}_{\text{carb}}$ values. Hour 0 represents 00:00 (midnight).

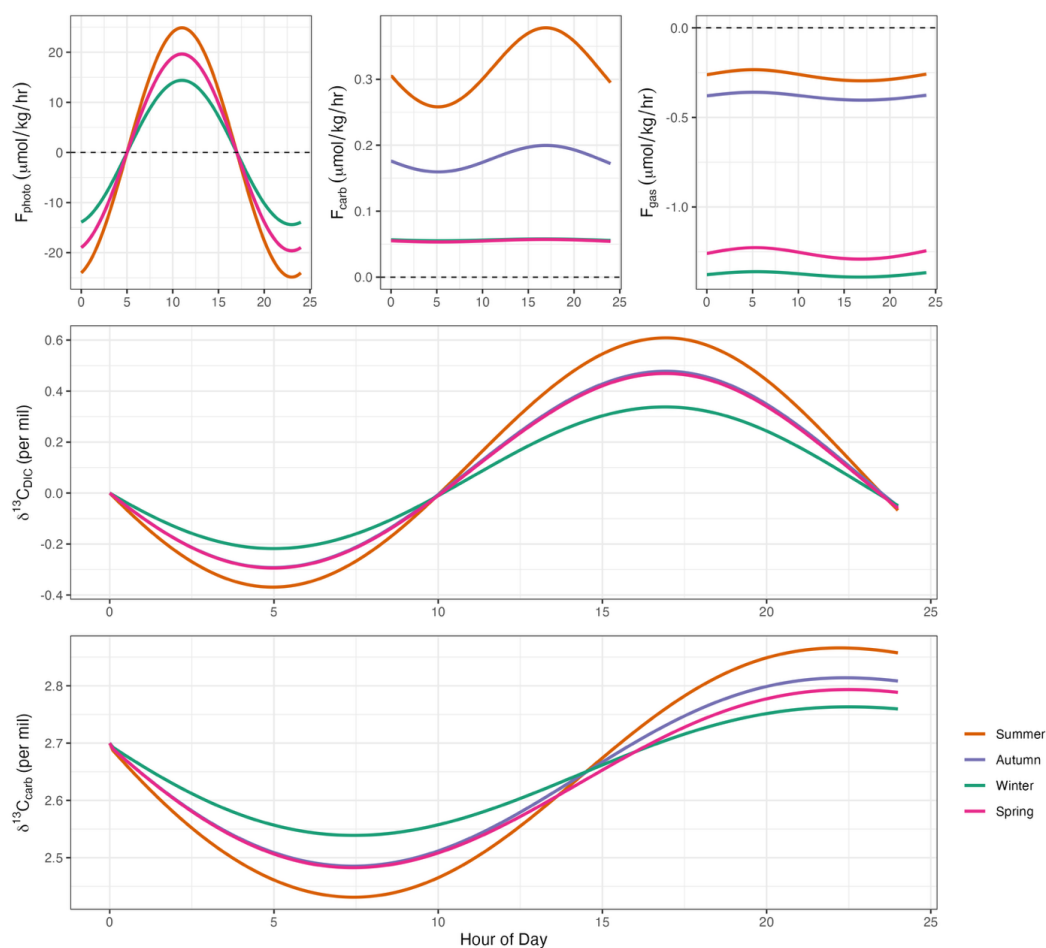


Figure 7. Isolating the impact of seasonality on the diurnal carbon engine in Great Salt Lake.

Because we have the most extensive data for Great Salt Lake, we model Great Salt Lake during summer (orange lines), autumn (purple lines), winter (green lines), and spring (magenta lines), modifying temperature ($T = 25^\circ\text{C}$, 18°C , -3°C , and 5°C , respectively), average wind speed ($u = 4.1$ m/s, 3.8 m/s, 3.3 m/s, and 4.3 m/s, respectively), and estimated productivity ($\kappa_p = 190$, 150 , 110 , and 150 , respectively). Resulting carbon fluxes (photosynthetic flux, F_{photo} ; carbonate precipitation flux, F_{carb} ; and gas exchange flux, F_{gas}) for these seasonal model simulations are shown in the top row. Note that in the F_{photo} panel the spring (green) and autumn (red) simulations result in identical F_{photo} , with the autumn simulation obscuring the spring simulation. In the upper center panel, the estimated F_{carb} during spring (green) and winter (blue) overlap, making the individual lines difficult to see. The center panel shows the modeled carbon isotope value of dissolved inorganic carbon ($\delta^{13}\text{C}_{\text{DIC}}$) for each seasonal simulation. As was the case in the F_{photo} panel, the autumn simulation (red) obscures the spring simulation (green) in this panel. The bottom panel shows the modeled cumulative carbon isotope value of carbonate ($\delta^{13}\text{C}_{\text{carb}}$) for each seasonal simulation. Hour 0 represents 00:00 (midnight).

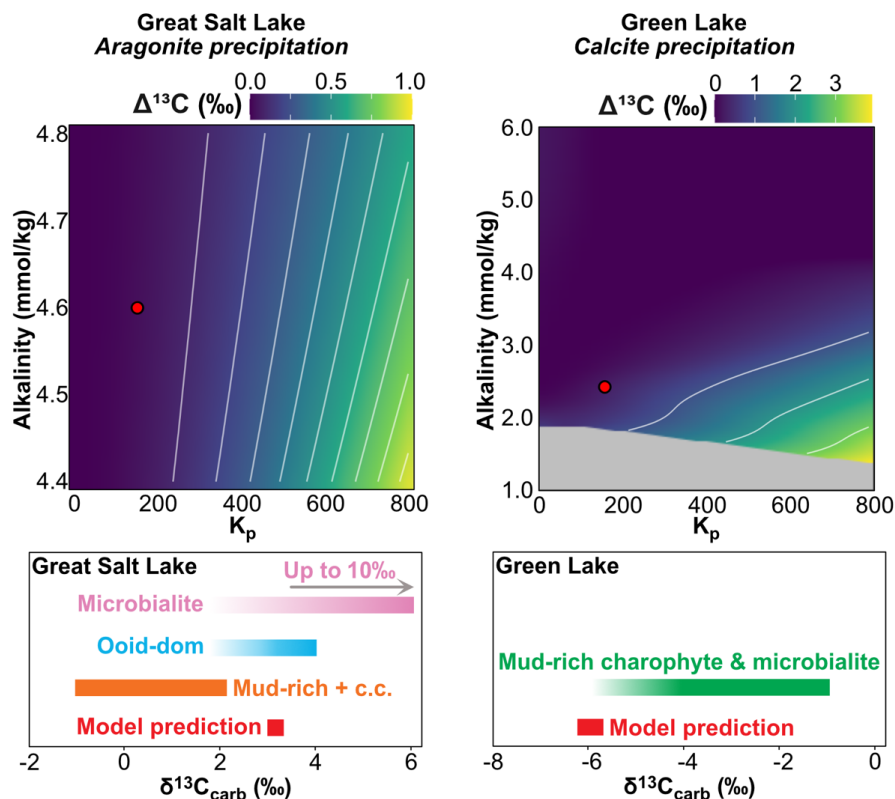
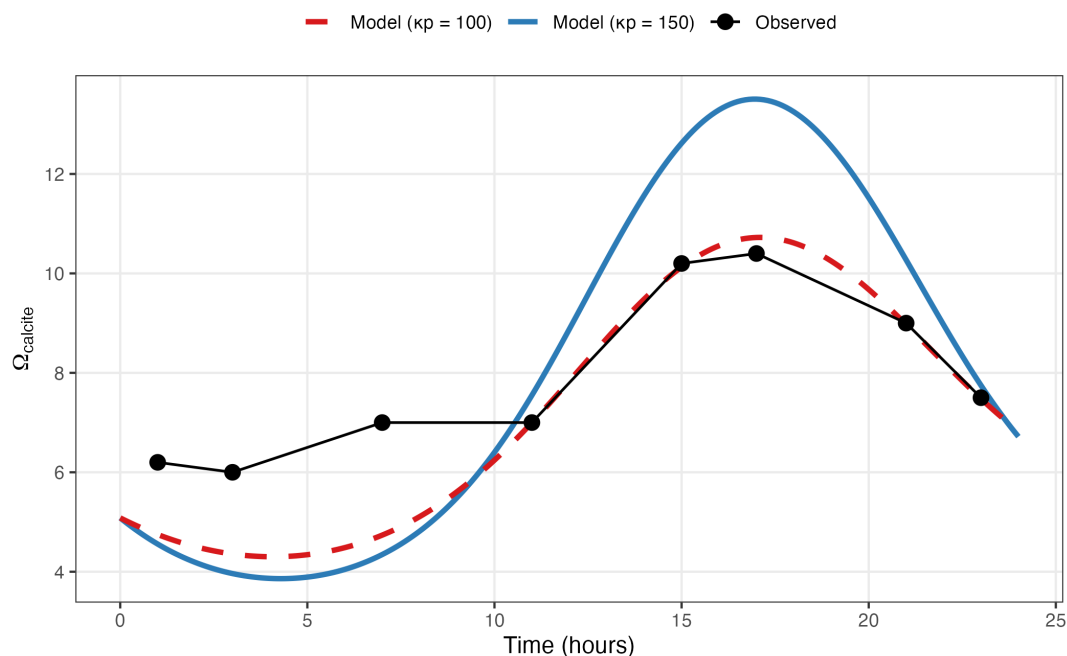


Figure 8. Broader implications of the diurnal carbon engine on lacustrine $\delta^{13}\text{C}_{\text{carb}}$ records. Contour plots show the predicted isotopic offset ($\Delta^{13}\text{C}$) between carbonate precipitated under an active diurnal carbon engine and carbonate precipitated from dissolved inorganic carbon (DIC) in equilibrium with long-term average $\delta^{13}\text{C}_{\text{DIC}}$, for Great Salt Lake (aragonite; left) and Green Lake (calcite; right), as a function of productivity (κ_p) and alkalinity. $\Delta^{13}\text{C}$ is defined as the difference between precipitation-weighted mean $\delta^{13}\text{C}_{\text{carb}}$ and the daily average instantaneous $\delta^{13}\text{C}_{\text{carb}}$ value (expected daily variability). Warmer colors indicate larger positive offsets. Red dots denote representative modern lake conditions used to estimate expected daily variability in $\delta^{13}\text{C}_{\text{carb}}$. In regions of parameter space where carbonate precipitation is limited or absent, $\Delta^{13}\text{C}$ approaches zero, indicating that diurnal variability does not necessarily translate to isotopic bias. Grey regions in the Green Lake panel indicate parameter combinations where calcite saturation state remains below 1 throughout the diurnal cycle, precluding carbonate precipitation and yielding no $\Delta^{13}\text{C}$ signal. Bottom panels compare modeled $\delta^{13}\text{C}_{\text{carb}}$ values with previously reported measurements from different sedimentary components in Great Salt Lake and Green Lake.



1211

1212 **Figure 9. Comparison of modeled and observed diurnal variability in calcite saturation state**
 1213 **(Ω).** Observations from Green Lake (0.25 m depth; July 28–29, 1978; Effler et al., 1981, Fig. 3a)
 1214 are shown as black points, and model results are shown as continuous lines. Model simulations are
 1215 shown for two productivity scenarios ($\kappa p = 100$, dashed red line; $\kappa p = 150$, solid blue line),
 1216 illustrating the sensitivity of diurnal variability to photosynthetic forcing. Observed data were
 1217 temporally aligned to the modeled peak to account for differences in time reference frames. The
 1218 model reproduces the timing and overall structure of the observed diurnal cycle, while the
 1219 magnitude of variability is controlled by productivity, demonstrating that the model captures first-
 1220 order dynamics of biologically mediated carbonate chemistry in lacustrine systems.