

The impact of ventilation patterns on calcite dissolution rates within karst conduits

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Abstract

Erosion rates in streams vary dramatically over time, as differences in streamflow and sediment load enhance or inhibit erosion processes. Within cave streams, and other bedrock channels incising soluble rocks, changes in water chemistry are an important factor in determining how erosion rates will vary in both time and space. Prior studies in surface streams, springs, and caves suggest that variation in dissolved CO₂ is the strongest control on variation in calcite dissolution rates. However, the controls on CO₂ variation remain poorly quantified. Limited data suggest that ventilation of karst systems can substantially influence dissolved CO₂ within karst conduits. However, the interactions among cave ventilation, air-water CO₂ exchange, and dissolution dynamics have not been studied in detail. In this study, three years of time series measurements of dissolved and gaseous CO₂, cave airflow velocity, and specific conductance from Blowing Springs Cave, Arkansas, were analyzed and used to estimate continuous calcite dissolution rates and quantify the correlations between those rates and potential physical and chemical drivers. We find that chimney effect airflow creates temperature-driven switches in airflow direction, and that the resulting seasonal changes in airflow regulate both gaseous and dissolved CO₂ within the cave. As in previous studies, partial pressure of CO₂ (pCO₂) is the strongest chemical control of dissolution rate variability. However, we also show that cave airflow direction, rather than streamflow, is the strongest physical driver of changes in dissolution rate, contrary to the typical situation in surface channel erosion where floods largely determine the timing and extent of geomorphic work. At the study site, chemical erosion is typically active in the summer, during periods

of cave downdraft (airflow from upper to lower entrances), and inactive in the winter, during updraft (airflow from lower to upper entrances). Storms provide only minor perturbations to this overall pattern. We also find that airflow direction modulates dissolution rate variation during storms, with higher storm variability during updraft than during downdraft. Finally, we compare our results with the limited set of other studies that have examined dissolution rate variation within cave streams and draw an initial hypothesis that evolution of cave ventilation patterns strongly impacts how dissolution rate dynamics evolve over the lifetime of karst conduits.

Keywords:

karst, bedrock channel, cave ventilation, carbon dioxide, dissolution

1 Introduction

The variation in geomorphic rates has an important influence on the relationship between erosional processes and the landforms that they produce (Wolman and Miller, 1960). Whereas concepts of the magnitude and frequency of geomorphic work have long been explored in the study of processes on Earth’s surface, fewer studies have examined the variability in rates of cave development or the factors that control this variation (Groves and Meiman, 2005). Cave passages are often developed by subsurface streams incising through bedrock. Since many cave streams carry substantial sediment loads (Farrant and Smart, 2011), mechanical erosion processes, such as occur within surface bedrock channels (Whipple et al., 2000), undoubtedly are active within these streams. However, most caves develop in karst settings, within highly soluble rocks, where chemical dissolution of the rock is an important driver of channel development and evolution (Ford and Williams, 2007; Palmer, 2007a).

A number of studies have measured solute export from basins and used these data to examine how rates of chemical denudation vary with streamflow (Wolman and Miller, 1960; Gunn, 1982; Schmidt, 1985; Goudie and Viles, 1999). These studies conclude that low to moderate flows produce an important percentage of the overall chemical geomorphic work at the basin scale. However, rates of channel incision by dissolution and basin-wide chemical denudation do not in general display the same relationship to streamflow. Groves and Meiman (2005) show that, in the Logsdon River passage of Mammoth Cave, Kentucky, conduit wall dissolution rates are a strong function of

25 discharge, with 87% of the work being done during high discharges that occur
26 less than 5% of the time. In contrast, they find that solute export is important
27 across a range of discharges, with only 38% of the export occurring during the
28 highest discharge class. Palmer (2007b) finds a similar relationship between
29 discharge and calcite dissolution rate in McFail’s Cave, New York. Analysis
30 of water chemistry data from streams across the United States suggests that
31 variability in calcite dissolution rates at most sites is more strongly correlated
32 with variability in dissolved CO₂ than with streamflow, and that in-channel
33 dissolution rates are often much higher than estimates of basin-wide denuda-
34 tion rates (Covington et al., 2015). Covington and Vaughn (2019) show that
35 seasonal variability in CO₂ is the primary driver for variation in calcite dis-
36 solution rates at a pair of karst underflow-overflow springs. Additionally, they
37 hypothesize that, during low flows, ventilation within the conduit feeding
38 the overflow spring drives a reduction in CO₂ and, consequently, dissolution
39 rates. In general, previous investigators have argued that cave ventilation,
40 which often occurs in the later stages of cave development, may reduce the
41 rates of chemical erosion within cave streams as caves becomes more mature
42 (Palmer, 2007b).

43 While it is clear that both floods and CO₂ dynamics are important drivers
44 of dissolution rate variability within cave streams, there are relatively few
45 cave sites where dissolution rate variability has been quantified (Groves and
46 Meiman, 2005; Palmer, 2007b; Covington and Vaughn, 2019). This lack of
47 data limits the ability to generalize about the controls on calcite dissolution
48 rate variability. Here, we analyze water chemistry data in a cave stream in
49 the Ozark Plateaus physiographic region of Arkansas to explore potential
50 controls of dissolution rate variability. We use recently developed techniques
51 for direct, high temporal resolution measurements of dissolved CO₂ (John-
52 son et al., 2010), combined with time series of specific conductance (SpC), to
53 estimate calcite dissolution rates over a three-year period. These measure-
54 ments are complemented by simultaneous measurements of cave air CO₂ and
55 cave airflow velocity. This enables us to explore interactions between cave
56 atmosphere dynamics and cave stream chemistry.

57 **2. Description of field site**

58 Blowing Springs Cave is located in Bella Vista, Arkansas, within the
59 Springfield Plateau region of the Ozark Plateaus (Figure 1). The cave is
60 developed in the cherty Mississippian Boone Limestone primarily within the

61 St. Joe Limestone Member (McFarland, 1998). Because of a high concentra-
62 tion of chert and clay impurities, the Boone Limestone develops a mantled
63 karst, where a thick regolith composed of chert and clay covers the karst
64 surface (Brahana, 2011). Within the region surrounding the cave, the Boone
65 Limestone spans the topography from valley floors to ridges. The cave, there-
66 fore, is autogenically recharged through a regolith cover. Because of the re-
67 golith cover, the region contains few obvious surface karst features, such as
68 sinkholes, though most of the valleys located in the recharge area are dry ex-
69 cept during periods of intense precipitation. The land cover in the recharge
70 area is a mixture of deciduous forest and low intensity residential. The region
71 has a temperate continental climate with a mean annual air temperature of
72 approximately 15°C (Adamski et al., 1995) and a mean annual precipitation
73 of about 114 centimeters per year (cm/yr) (Pugh and Westerman, 2014).

74 Blowing Springs Cave contains 2,397 meters (m) of mapped cave passage
75 and consists of a dendritic stream network (Figure 1). Water enters the
76 main cave stream through a number of small infeeding channels and through
77 many percolating fractures within the cave ceiling, though the largest source
78 of discharge to the cave stream is the upstream sump. Many of the cave
79 passages are oriented along an orthogonal set of NE-SW and NW-SE trending
80 fractures. The only known entrance of the cave is at the spring, and, as
81 the name suggests, the cave exhibits strong airflow, with air blowing out of
82 the spring entrance during times of warm outside temperatures. The spring
83 emerges near the elevation of the local base-level stream, which is Little Sugar
84 Creek.

85 3. Methods

86 3.1. Collection of time series data

87 Time series data of water quality and cave atmospheric parameters were
88 collected at a site located approximately 150 m inside the cave entrance,
89 which is labeled as Cave Measurement Station in Figure 1. Cave airflow ve-
90 locity, cave air barometric pressure, and CO₂ concentrations in the cave air
91 and water were logged on a Campbell ScientificTMCR850 datalogger. Cave
92 airflow velocity and direction were measured using a Campbell Scientific
93 WINDSONIC1 2D ultrasonic anemometer (resolution 0.01 m/s; accuracy
94 $\pm 2\%$ at 12 m/s; directional accuracy $\pm 3^\circ$). Barometric pressure was mea-
95 sured using a Campbell ScientificTMCS100 sensor (accuracy ± 0.5 hPa). CO₂
96 concentrations in the air and water were measured using Vaisala GMM220

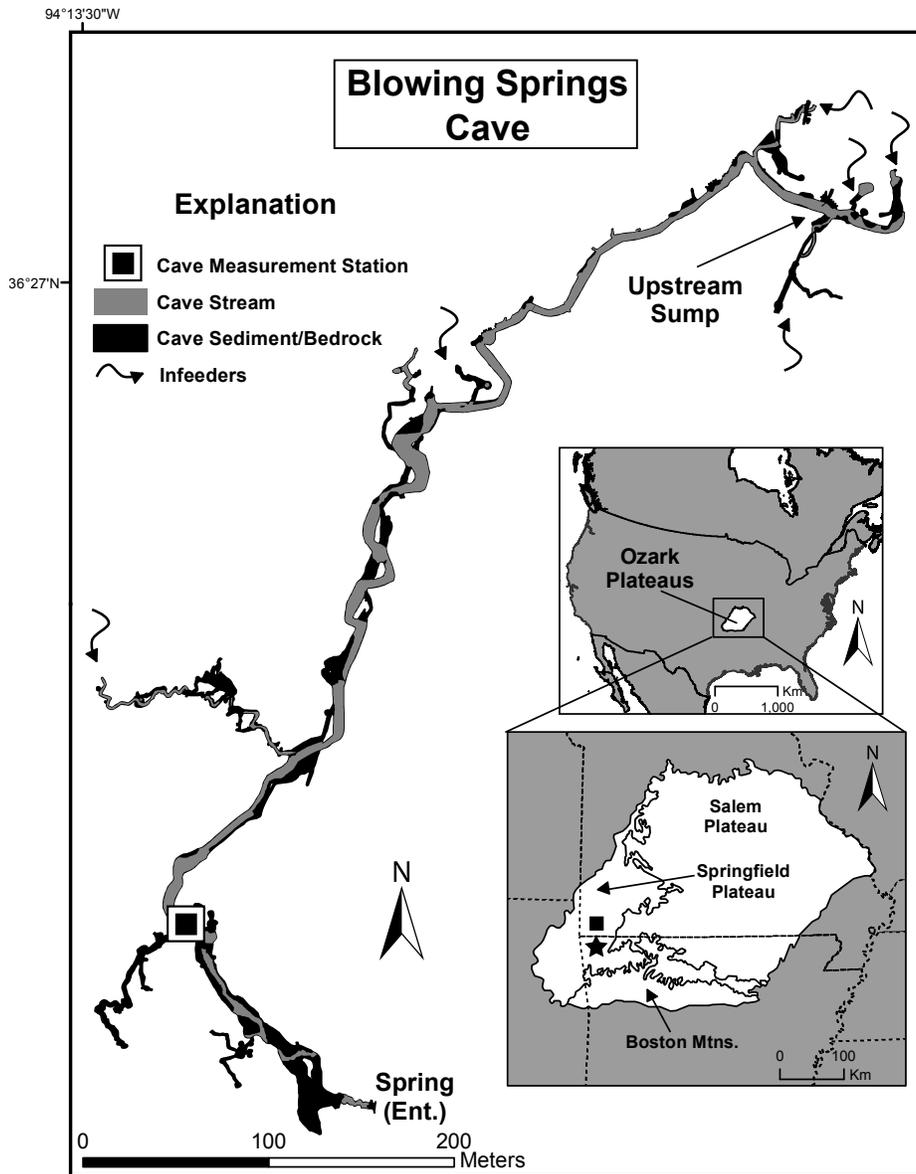


Figure 1: Location of Blowing Springs Cave (star) and map of the cave depicting the entrance, measurement location, stream, and upstream sump. Within the inset, the square indicates the location of the USGS stream gauge on Little Sugar Creek (USGS station no. 07188838). Modified from Knierim et al. (2017). Original cave map and survey data from Covington (2007).

97 CO₂ transmitters with a range of 0 to 5,000 parts per million (ppm) (accu-
98 racy $\pm 1.5\%$ of range $\pm 2\%$ of reading). The CO₂ sensors were protected from
99 moisture with a waterproof breathable membrane (PTFE) as described in
100 Johnson et al. (2010). One sensor was exposed to the cave air, and the other
101 was submerged in the cave stream to enable direct measurement of the par-
102 tial pressure of CO₂ (pCO₂) of the water. This setup provides a more reliable
103 means of recording pCO₂ than continuous measurement of pH, as the CO₂
104 sensors exhibit much less drift over time than pH electrodes (Johnson et al.,
105 2010; Covington and Vaughn, 2019). For the CO₂ sensors, a warm-up pe-
106 riod of 15 minutes was used before a measurement was taken. This warm-up
107 period allowed thermal equilibration of the sensor and helped to drive any
108 moisture out of the sensor optics. We found that this warm-up period was
109 crucial, as substantial instrument drift occurred within the first few minutes
110 of power-up. Measured values of pCO₂ were not adjusted with water depth as
111 described by (Johnson et al., 2010), because subsequent theoretical analysis
112 and experiments have shown that this adjustment is incorrect (Blackstock
113 et al., 2019). CO₂ readings were taken once an hour to conserve power,
114 whereas other parameters recorded on the CR850 were read on a one-minute
115 interval.

116 In this study, all CO₂ concentrations are reported in ppm, which are nom-
117 inally values of ppm volume (ppmV) within a gas phase, where the concen-
118 trations for our dissolved measurements are within a headspace equilibrated
119 to the dissolved phase. Air and water values are reported in the same units
120 to enable direct comparison, and use ppm as these are the units typically
121 used in cave atmospheric studies. While the instrument reports ppmV, it ac-
122 tually directly measures molecular density. Concentrations in ppmV must be
123 corrected both for pressure and temperature. We are specifically interested
124 in pCO₂, which can be more directly calculated from molecular density than
125 can ppmV, only requiring a temperature correction and the factory calibra-
126 tion pressure (Blackstock et al., 2019), which we assume to be 1 atm. We
127 apply the temperature correction, and then 1 ppm is equivalent to a pCO₂
128 of 10^{-6} atm. Therefore, our reported ppm values are directly proportional
129 to pCO₂ and can be thought of as a type of pCO₂ value, albeit with unusual
130 units.

131 The specific conductance (SpC) and temperature of the cave stream were
132 measured at the same site as CO₂ on a time interval of 5 minutes using an
133 Onset HOBOTMU24-001 freshwater conductivity logger with an accuracy of
134 3% or 5 microsiemens per centimeter ($\mu\text{S}/\text{cm}$). Outside of the cave, air tem-

135 perature and relative humidity were measured at a 5-minute interval using an
 136 Onset HOBOTMU23-001 temperature (accuracy $\pm 0.21^{\circ}\text{C}$) and relative hu-
 137 midity (accuracy $\pm 2.5\%$) logger that was mounted onto a tree. To provide a
 138 proxy for stable cave temperature, we deployed an Onset HOBOTMU20L-04
 139 pressure and temperature logger in the cave air approximately 400 m inside
 140 the cave (accuracy $\pm 0.44^{\circ}\text{C}$). Unless specified otherwise, all data presented
 141 here are hourly averages, to align with the frequency of CO_2 measurements.

142 The site within the cave was visited roughly every four weeks, which
 143 was the approximate duration of the battery power supply (two 12-volt,
 144 20 Amp-hour lithium-ion batteries). During each site visit, batteries were
 145 changed, data downloaded, and quality control on-site measurements were
 146 made of specific conductance, water temperature, and CO_2 concentrations in
 147 the air and water. To make on-site measurements of CO_2 a portable Vaisala
 148 GMM220 was used that was connected to a battery and data logger. Spot
 149 measurements of CO_2 in the water required a roughly 30-minute equilibration
 150 period for gas concentrations to exchange across the PTFE membrane.

151 Whereas a partial record of stage and estimated discharge is available
 152 at the spring, frequent human disturbances of the stream channel near the
 153 weir (the site is located in a park) reduced the quality of the available
 154 dataset. Rather than using this corrupted record, we employ a modifica-
 155 tion of an estimation of discharge at Blowing Springs Cave developed by
 156 Knierim et al. (2015b) using the nearby United States Geological Survey
 157 (USGS) streamflow-gaging station on Little Sugar Creek (USGS Station no.
 158 07188838 Little Sugar Creek near Pineville, Missouri) with data available
 159 from the USGS National Water Information System (U.S. Geological Sur-
 160 vey, 2020).

161 Knierim et al. (2015b) found that a linear regression provided a reason-
 162 able approximation of discharge at Blowing Springs. However, they also
 163 noted that Blowing Springs exhibited a minimum baseflow discharge, and
 164 therefore they introduced a threshold minimum value. To remove the sharp
 165 threshold and reduce errors on moderate discharge values, we introduce a
 166 new regression to daily averaged discharges using a sigmoidal function in log
 167 space (Figure 2),

$$\log(Q_{\text{BSC}}) = \frac{A}{1 + \exp(-D[\log(Q_{\text{LSC}}) - B])} + C, \quad (1)$$

168 where Q_{BSC} is the discharge at Blowing Springs in m^3/s and Q_{LSC} is the
 169 streamflow in Little Sugar Creek, and the best fit values of the fitting pa-

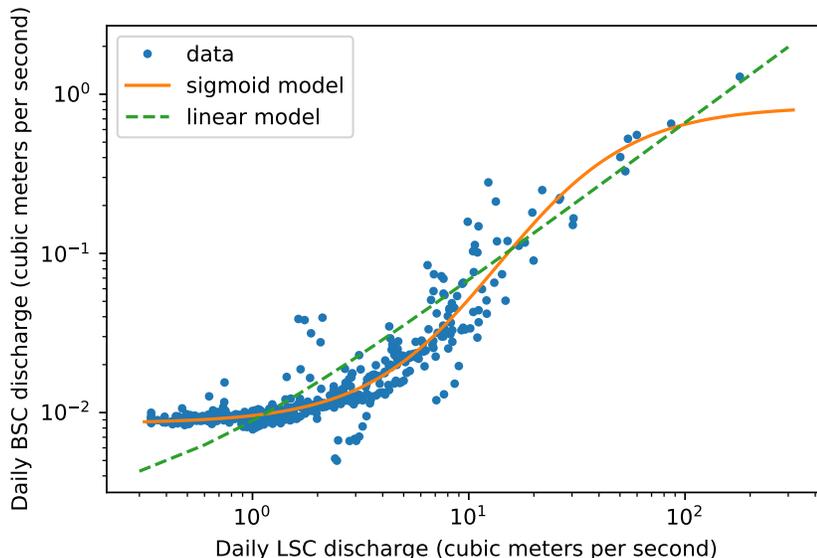


Figure 2: Regression used in the estimation of Blowing Springs discharge from the stream gauge at Little Sugar Creek. The sigmoid model is used in this work, and the linear model is from Knierim et al. (2015b).

170 rameters are $A = 1.99$, $B = 1.14$, $C = -2.07$, and $D = 3.24$. The root mean
 171 squared percentage error (RMSPE) in this model is 33%. The sigmoidal
 172 shape nicely captures the transition to baseflow and also captures a possible
 173 rolloff at high discharges ($\sim 1 - 2 \text{ m}^3/\text{s}$) that is evident in the 15-minute
 174 data. Many karst springs do exhibit maximum discharges (Bonacci, 2001),
 175 though this upper rolloff also has limited impact on our estimates due to
 176 the small number of data points that fall at these highest discharges. Most
 177 of our results are independent of this discharge estimate, which is primarily
 178 used to indicate the occurrence and frequency of high and low flow periods
 179 to examine how dissolution rate varies with flow. The precise magnitudes of
 180 discharge are not necessary for interpreting our data.

181 3.2. Calculation of dissolution rates

182 In order to calculate dissolution rates from available kinetic rate equa-
 183 tions, we need values for the dissolved Ca concentration and the pCO_2 . While
 184 we measure pCO_2 directly, Ca concentrations were estimated from SpC time

185 series collected at the site, as has also been done in prior studies of dissolution
186 rate dynamics within karst systems (Groves and Meiman, 2005; Covington
187 and Vaughn, 2019) and is appropriate where water is predominantly Ca–
188 HCO₃ type. To estimate Ca, we use a linear regression on available SpC
189 and Ca data ($n = 109$) from a prior study at Blowing Springs Cave (Knierim
190 et al., 2017). We find that $[Ca] = 0.175 \times SpC - 2.51$, where $[Ca]$ is concentra-
191 tion in mg/L. The RMSPE of this relationship is 10%. However, uncertainties
192 are larger for lower SpC values. Therefore, for error propagation purposes we
193 divide the data into quartiles and use separate uncertainty values for each
194 quartile, these uncertainties range from 4.5% to 13.7%. These data and fur-
195 ther details on error estimation can be found in the data repository described
196 in the Data Availability section of this paper.

197 To calculate dissolution rates from $[Ca]$ and pCO_2 time series, we use two
198 available calcite kinetic equations, which we refer to as the PWP (Plummer
199 et al., 1978) and Palmer equations (Palmer, 1991). Both equations are de-
200 rived from the same experimental dataset, but the Palmer equation is a direct
201 fit to the data, whereas the PWP equation incorporates a more mechanistic
202 approach to parameter estimation. The Palmer equation generates a closer fit
203 to the observed dissolution rates near saturation and also provides parameter
204 values for impure calcite, which is more appropriate for limestone. Coving-
205 ton and Vaughn (2019) found that the Palmer equation provided much closer
206 estimates of mass loss rates of limestone tablets deployed in the field, and
207 therefore, it is likely that these rates are more accurate in natural settings.

208 The PWP equation sometimes produces negative rates, which might
209 suggest calcite precipitation. However, calcite precipitation normally does
210 not occur until waters are highly supersaturated, therefore negative PWP
211 rates are not necessarily indicative of precipitation. In time series of dis-
212 solution rates, we show both the PWP and Palmer equations. However,
213 when studying sensitivity of rates to various potential controls, it is help-
214 ful to have a broader range of values, including a range of negative val-
215 ues that represent different extents of supersaturation. Therefore, we use
216 PWP rates to explore controls on variability, even though the magnitude
217 of the rates is likely too high (Covington and Vaughn, 2019). Rates pre-
218 dicted by both equations typically vary monotonically with one another.
219 Both dissolution rate equations were calculated using algorithms in the *Olm*
220 Python package v0.36 (Covington et al., 2015), which is available on *GitHub*
221 (<https://doi.org/10.5281/zenodo.4030690>). Uncertainties in dissolution
222 rates were estimated using Monte Carlo error propagation from uncertainties

223 in $[Ca]$ and pCO_2 .

224 4. Results

225 4.1. Large-scale patterns in the time series data

226 The water chemistry and cave atmosphere parameters were recorded over
227 a period of approximately three years (Oct. 2014-Jan. 2018). CO_2 concentra-
228 tions in air and water range between near atmospheric concentration (≈ 500
229 ppm) to above 5,000 ppm (Figure 3a). For a few short periods, pCO_2 in the
230 water exceeded the measurement range of the sensor deployed ($\approx 5,500$ ppm).
231 Concentrations in the water almost always exceeded those in the air. Both
232 dissolved and gaseous CO_2 concentrations within the cave showed seasonal
233 patterns, with higher concentrations in summer and lower concentrations
234 during winter. Gaseous CO_2 within the cave dropped near atmospheric val-
235 ues for much of the winter. Dissolved CO_2 often exhibited spikes to higher
236 values associated with high discharge events. Gaseous CO_2 displayed strong
237 diurnal variability during certain periods, particularly during the spring and
238 fall. These periods of variability are associated with times when outside air
239 temperatures are near those of the cave air temperature, which is approxi-
240 mately the mean surface air temperature (Badino, 2010).

241 Cave airflow velocity also had a seasonal pattern with outward (positive)
242 airflow during warm periods and inward (negative) airflow during cold peri-
243 ods (Figure 3b). There was diurnal variability in airflow velocity, particularly
244 during spring and fall periods, with some days exhibiting both inward and
245 outward airflow at different times of day. Again, these periods of high vari-
246 ability are times when outside temperatures are near the temperature of the
247 cave atmosphere.

248 Specific conductance displayed a range from 65 to 265 $\mu S/cm$. Variability
249 in SpC was more strongly related to discharge than to season (Figures 3c,d).
250 SpC was high during periods of low flow and low during periods of high flow,
251 particularly flood events. All records display gaps that are associated with
252 sensor or power failures. However, the SpC record is the least complete, with
253 a large number of gaps resulting from sensor failure, damage during storms,
254 or download failure, which lead to memory filling on the datalogger before the
255 next opportunity to download. As can be seen visually in Figures 3c,d, there
256 is a strong correlation between specific conductance and stream discharge
257 (Q). The relationship between these parameters is explicitly displayed in

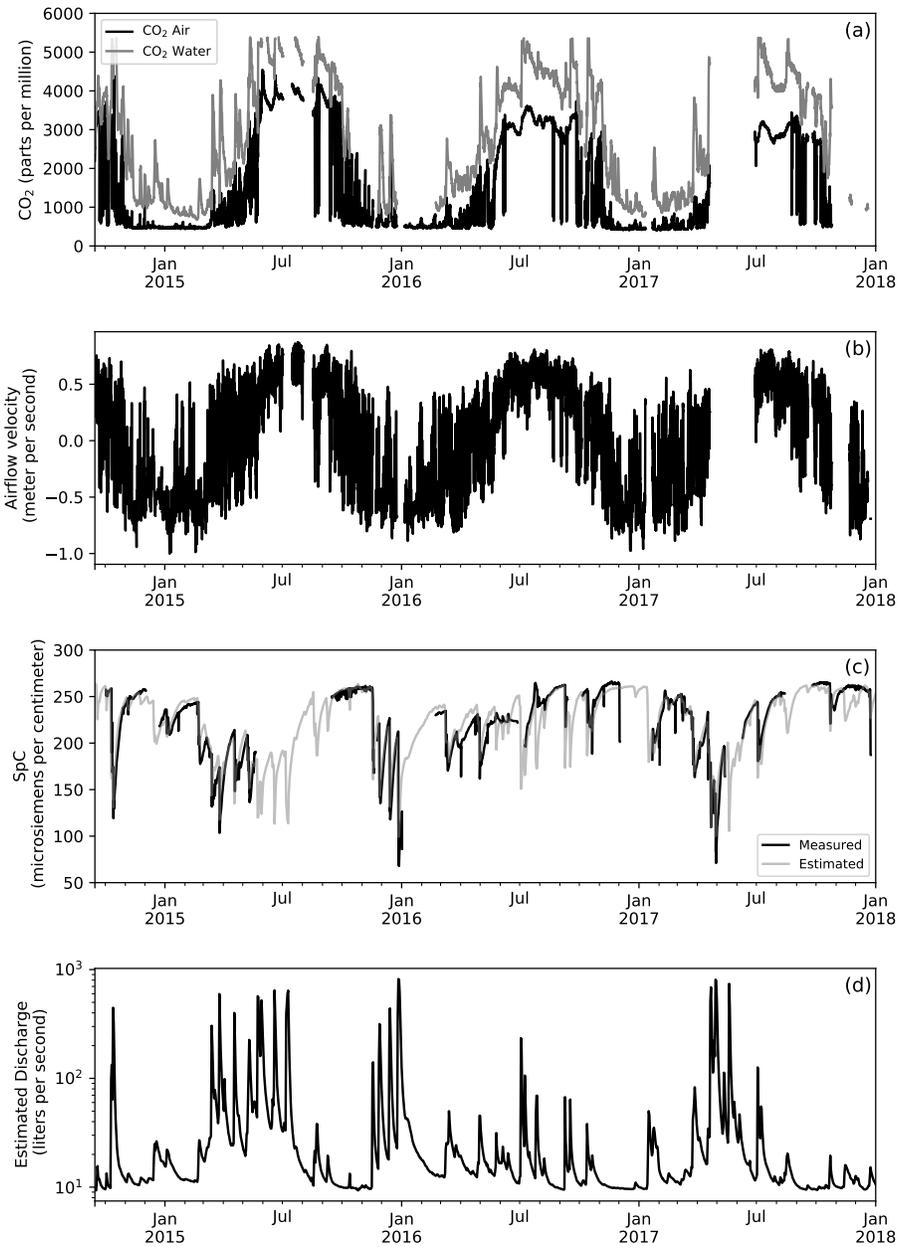


Figure 3: Time series data for the entire study period: (a) CO₂ concentrations in the air and water, (b) cave airflow velocity (positive values indicate the cave is blowing out), (c) specific conductance in the cave stream (black) and an estimated daily specific conductance using a regression to discharge (gray), and (d) estimated discharge of the cave calculated from streamflow at Little Sugar Creek using Equation 1.

258 Figure 4 along with a 4th-order polynomial regression between $\log(Q)$ and
259 specific conductance given by

$$\text{SpC} = Ax^4 + Bx^3 + Cx^2 + D, \quad (2)$$

260 where $x = \log(Q)$, $A = -48.6$, $B = 296.2$, $C = -639.4$, and $D = 649.2$.
261 The coefficients were determined using the *polyfit()* function from the *NumPy*
262 package in Python. Because of the large number of gaps, and the strong rela-
263 tionship between discharge and SpC, the daily estimated SpC using Equation
264 2 is shown in Figure 3c. The root-mean-squared error (RMSE) in estimated
265 SpC is $14.9 \mu\text{S}/\text{cm}$. The valid discharge range for the fit is from approxi-
266 mately 10 L/s to 1,000 L/s. During periods where SpC data were lacking, we
267 estimate [Ca] from these modeled values. To account for additional uncer-
268 tainty introduced by this step, we divide these data into 10 percentile-based
269 bins of discharge and calculate the RMSE for each of these bins. We use
270 Gaussian Error Propagation to convert these uncertainties in SpC into un-
271 certainty in [Ca].

272 4.2. Relationship between cave airflow and external air temperature

273 The seasonal and diurnal patterns in cave airflow velocity suggest a re-
274 lationship between outside air temperature and cave airflow, as would be
275 expected in the case of chimney effect airflow (Wigley and Brown, 1976;
276 Luetscher et al., 2008; Badino, 2010; Covington and Perne, 2015). Chimney
277 effect airflow is an airflow mechanism driven by density contrasts between
278 the cave air and outside air and occurs within cave systems with more than
279 one opening to the outside. During periods of warm outside temperatures,
280 cave air is more dense than outside air and it is therefore pushed from upper
281 entrances to lower entrances. During cold outside temperatures, cave air is
282 less dense than outside air and rises from lower entrances to upper entrances.
283 Note that such airflow does not require human-sized entrances or large el-
284 evation differences. Millimeter-scale fracture apertures and decimeters of
285 elevation difference are sufficient (Covington, 2016).

286 The pressure difference, ΔP , that drives chimney effect airflow can be
287 approximated using (cf. Badino (2010))

$$\Delta P = \rho_{\text{in}} g h \frac{\Delta T}{T_{\text{ext}}}, \quad (3)$$

288 where ρ_{in} is the density of the air inside the cave, g is Earth's gravitational
289 acceleration, h is the height difference between the two entrances, ΔT is the

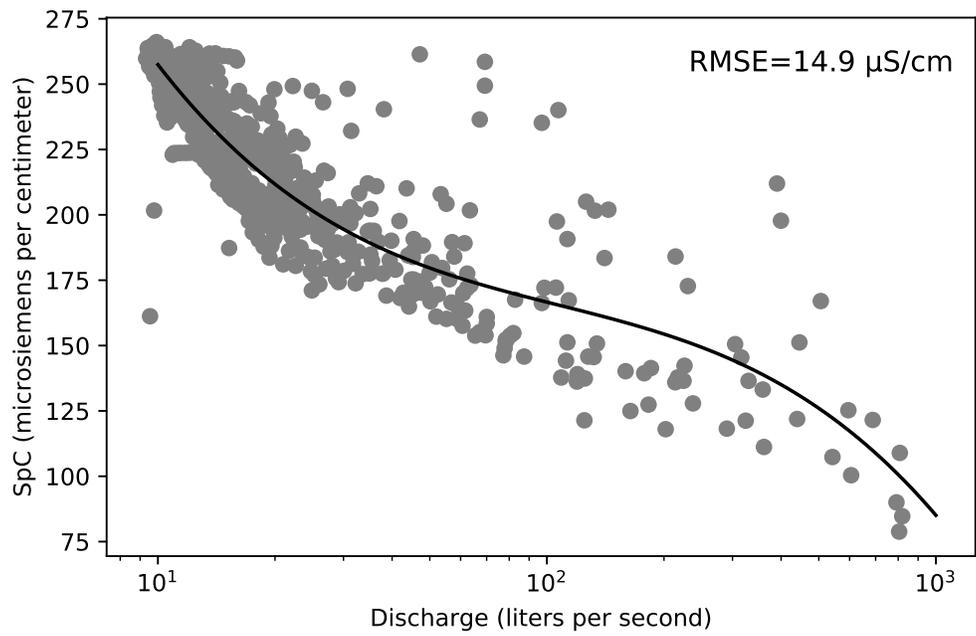


Figure 4: Relationship between specific conductance and discharge shown along with the polynomial regression (Equation 2) used for estimating specific conductance during periods of missing record (RMSE=root-mean-squared error).

290 difference between cave and external temperature, and T_{ext} is the external
 291 air temperature in Kelvin. Chimney effect airflow is typically turbulent, and
 292 therefore the Darcy–Weisbach equation for flow of fluid in a pipe provides a
 293 reasonable approximation (Luetscher and Jeannin, 2004) for airflow velocity,
 294 V , with

$$V = \sqrt{\frac{2D_{\text{H}}\Delta P}{\rho_{\text{in}}fL}}, \quad (4)$$

295 where D_{H} is the hydraulic diameter of the flow path, f is the Darcy-Weisbach
 296 friction factor, and L is the length of the flow path. Combining these two
 297 equations, leads to

$$V = \sqrt{\frac{2D_{\text{H}}gh\Delta T}{fLT_{\text{ext}}}}, \quad (5)$$

298 where one can see that the airflow velocity is predicted to scale with the
 299 square root of the temperature difference between outside and cave air. To
 300 test the plausibility of chimney effect airflow as the primary mechanism be-
 301 hind the observed airflow in Blowing Springs Cave, airflow velocity is plotted
 302 against the temperature difference between inside and outside air, and a
 303 square root relationship is fit to the data (Figure 5). Not only is there a
 304 strong relationship between temperature difference and cave airflow velocity,
 305 but the shape of the relationship is closely matched by a square root function,
 306 $V = R\Delta T^{1/2}$, where the best fit value of the resistance factor $R = 0.18$.

307 There are no known human-sized upper entrances to Blowing Springs
 308 Cave. However, we can use knowledge of the cave system to estimate ap-
 309 propriate values of unknown parameters in Equation 5, using $L = 1,000$ m,
 310 which is the approximate distance from the entrance to the upstream sump,
 311 and $h = 25$ m, which is the approximate elevation difference between the
 312 spring entrance and the valleys feeding the cave that are likely to hold upper
 313 entrances. Using values of $g = 9.8\text{ m/s}^2$ and $f = 0.05$, which is a typical
 314 value for a rough pipe and high Reynolds Number (Larock et al., 2000), we
 315 can estimate that the hydraulic diameter would have to be approximately
 316 equal to 1 m in order to produce the observed value of R . Though diameters
 317 of the mapped portion of the cave are highly variable (Figure 1), with values
 318 reaching up to 5–10 m within larger rooms, a diameter of 1 m is roughly
 319 consistent with observed diameters in much of the cave. The untraversable
 320 upper portions of the flow paths must also be substantially smaller, because
 321 they are too small for a human to enter.

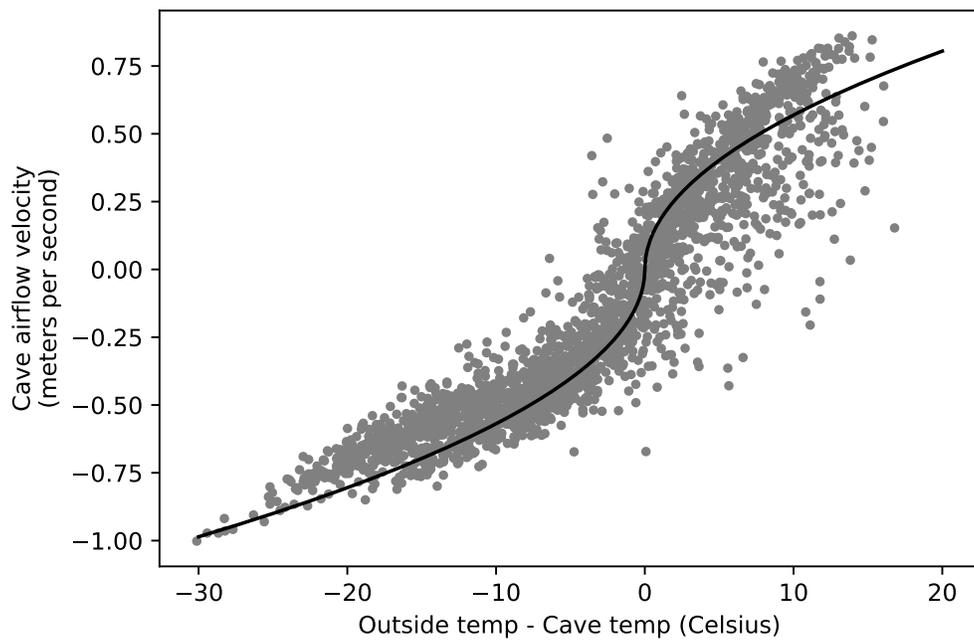


Figure 5: Relationship between airflow velocity and temperature difference between outside air and cave air, with a fitting function $V = R\Delta T^{1/2}$, with the resistance factor set to $R = 0.18$.

322 To make the link between airflow direction and the chimney effect mech-
323 anism explicit in our further discussion, from this point on we will refer to
324 cave airflow direction as either “updraft” or “downdraft” (Figure 6). Updraft
325 occurs during periods when the cave air is less dense than outside air (e.g.
326 winter) and air flows from lower to upper entrances (inward). Downdraft
327 occurs when the cave air is denser than outside air (e.g. summer) and air
328 flows from upper to lower entrances (outward). Because the airflow velocity
329 was measured near a lower entrance, updraft corresponds to inward airflow
330 (negative velocity), and downdraft corresponds to outward airflow (positive
331 velocity). For the study period, the cave exhibited nearly continuous down-
332 draft from June 1 to October 1, and relatively consistent updraft from De-
333 cember 1 to March 1, with occasional winter reversals during warm periods.
334 The spring and fall seasons exhibit frequent switching between updraft and
335 downdraft, often diurnally.

336 *4.3. Relationship between airflow velocity and CO₂*

337 The seasonal patterns in cave airflow and CO₂ in the air and water are well
338 aligned (Figure 3). Additionally, there are strong relationships between CO₂
339 and cave airflow on short timescales (Figure 7). During periods of diurnal
340 airflow reversals, CO₂ in the cave air also shows daily peaks and troughs.
341 When airflow direction switches from downdraft to updraft, cave air CO₂
342 drops suddenly to concentrations near atmospheric (~ 500 ppm), as outside
343 air is quickly brought to the location of the sensor. When airflow switches
344 from updraft to downdraft, cave air CO₂ rises somewhat more slowly, likely
345 as a result of mixing of high and low CO₂ air within the cave atmosphere.
346 Dissolved CO₂ within the cave stream does not respond as rapidly to airflow
347 reversals as the cave air. However, the cave stream CO₂ does have a muted
348 response that has a lag of a few days (Figure 7).

349 Dissolved and gaseous CO₂ both display statistically significant corre-
350 lations (p-value<0.0001) with airflow velocity when averaged over daily or
351 weekly timescales (Figure 8). Here we quantify correlation using Spearman’s
352 rank correlation coefficient because the relationships are non-linear (Helsel
353 and Hirsch, 2002). Correlations are stronger over the weekly timescales than
354 the daily timescales, particularly for dissolved CO₂. The gaseous CO₂ con-
355 centrations display a clear threshold near zero airflow velocity (Figure 8),
356 which divides time periods with updraft and downdraft. During periods of
357 updraft, the cave air CO₂ is typically near outside atmospheric concentra-
358 tions, whereas during downdraft, concentrations substantially increase above

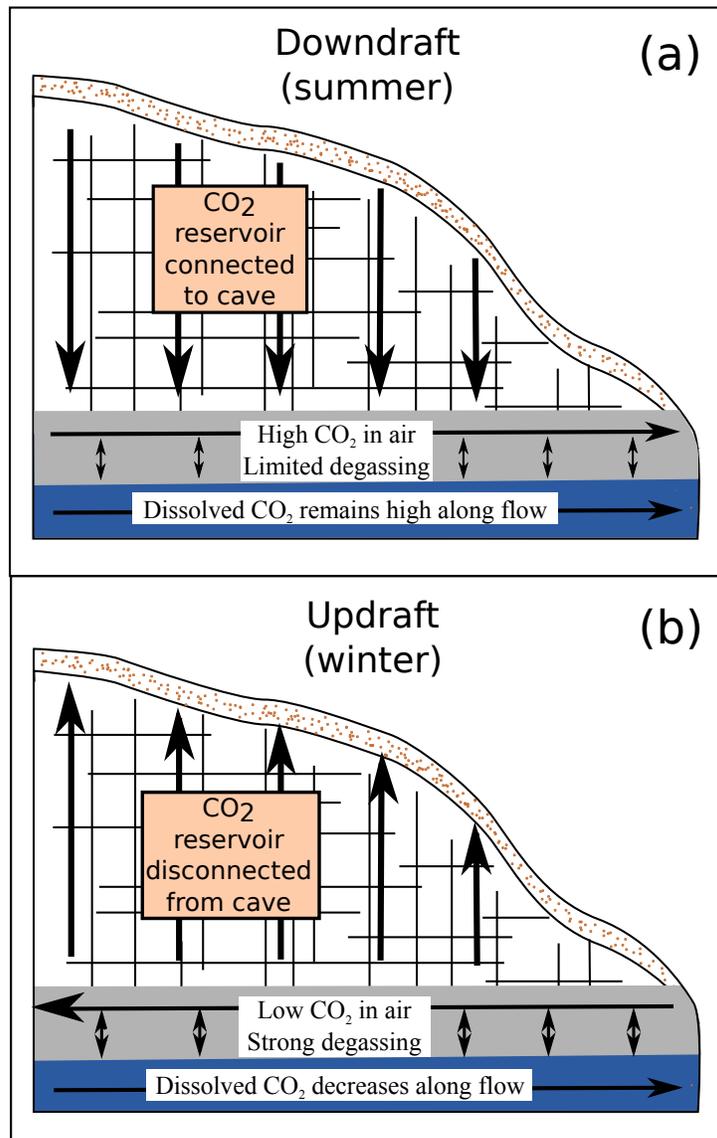


Figure 6: Conceptual model of how ventilation direction impacts dissolution rates in the cave stream: (a) During downdraft (summer conditions), air flows vertically downward through the soil and unsaturated zone, obtaining high CO₂. Cave air pCO₂ is high and therefore degassing of CO₂ from the cave stream is limited. Consequently, dissolved CO₂ and dissolution rates remain high along the main conduit. (b) During updraft (winter conditions), atmospheric air enters the cave through the large lower entrance and then flows upward through the high-CO₂ unsaturated zone. The cave air is disconnected from this high CO₂ zone and strong degassing of CO₂ occurs along the stream, reducing pCO₂ and dissolution rates. During winter storms, vertical flow of water can transport CO₂ through the unsaturated zone and effectively reconnect the cave stream to the CO₂ reservoir.

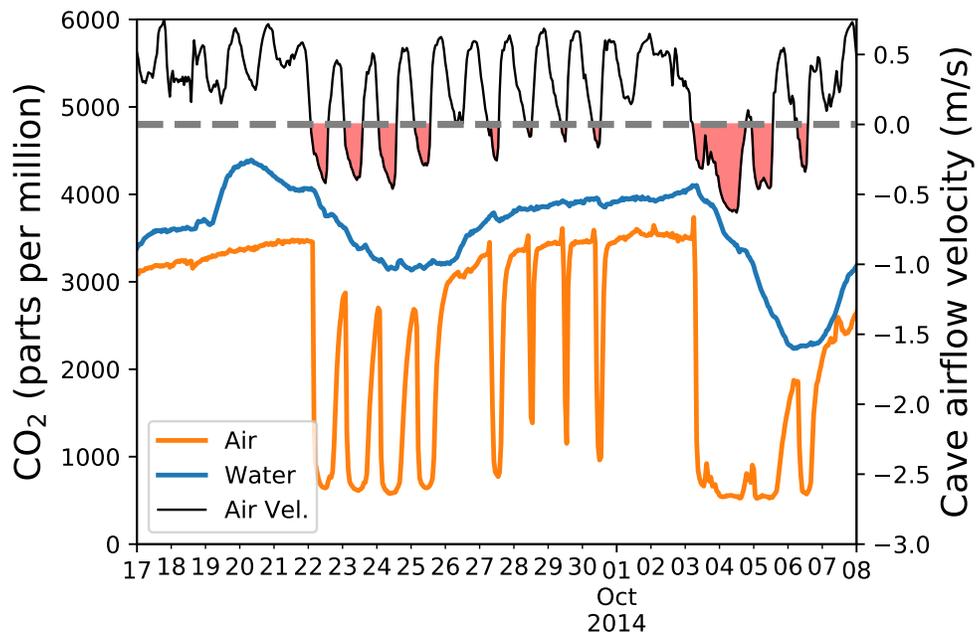


Figure 7: Time series of airflow velocity (top, black), and CO₂ concentrations in air (orange, bottom) and water (blue, middle), where the gray dashed line demarcates zero airflow velocity and the shaded red portions of the curve are periods of updraft (inward airflow). During updraft, gaseous CO₂ concentrations decrease sharply to near atmospheric concentrations. During extended periods of updraft, dissolved CO₂ also decreases. During downdraft CO₂ in the air and water increase.

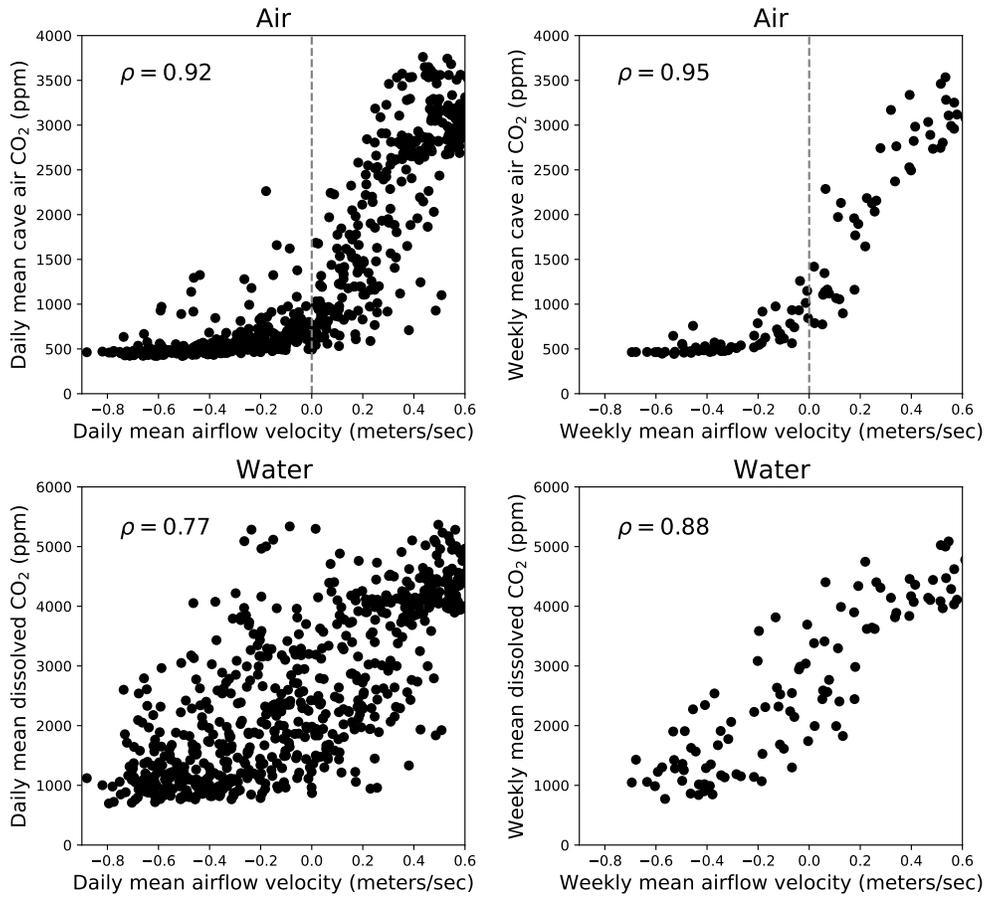


Figure 8: Relationships between airflow velocity and CO₂ concentrations in air and water averaged on daily and weekly timescales. Correlations are quantified using Spearman's rank correlation coefficient ρ . Dashed vertical lines indicate the threshold change in CO₂ concentrations that corresponds to airflow reversals (ppm = parts per million).

359 atmospheric values. The relationship between dissolved CO₂ and cave airflow
360 does not display a clear threshold at zero cave airflow but still has a clear pat-
361 tern of lower concentrations during updraft and higher concentrations during
362 downdraft (Figure 8).

363 4.4. Dissolution rate dynamics in the cave stream

364 The dissolution rates show a strong seasonal signal that is in-phase with
365 the seasonal CO₂ variation (Figure 9). That is, there are higher rates of
366 dissolution during the summer months, when pCO₂ is also high and the water
367 is undersaturated with respect to calcite. Lower rates of dissolution occur
368 during the winter months (frequently negative PWP rates), when pCO₂ is
369 low and the water is typically supersaturated. The average of this seasonal
370 signal is near calcite saturation (or zero dissolution rate), but the stream
371 spends slightly more time in the undersaturated condition, when dissolution
372 is active. In addition to the seasonal signal, there is clear variability on daily
373 to weekly timescales.

374 To study the chemical controls on dissolution rate variation, dissolution
375 rates averaged over daily timescales are plotted versus the two primary chem-
376 ical drivers (Figure 10): dissolved CO₂ and a proxy for dissolved load (SpC).
377 To quantify the correlations between the chemical drivers and dissolution
378 rate, we calculated Spearman's rank correlation coefficients. Both chemical
379 drivers correlate with dissolution rates (p-value<0.0001), but CO₂ is more
380 strongly correlated ($\rho = 0.80$) than SpC ($\rho = -0.44$). The cloud of points in
381 the dissolution rate-CO₂ plot (Figure 10a) shows a relatively sharp edge at
382 low dissolution rate. This edge is a result of baseflow conditions, where SpC
383 displays a typical value of about 220 $\mu\text{S}/\text{cm}$.

384 In addition to direct chemical drivers, dissolution rates vary as a func-
385 tion of external physical controls that produce variations in those chemical
386 drivers. The two most important physical controls on chemical variation at
387 the site are cave airflow velocity and stream discharge. Cave airflow, and
388 particularly its direction, is an important driver of dissolved CO₂ (Figures
389 6 and 8). Discharge may produce variation in both dissolved load and dis-
390 solved CO₂, either through dilution during storm-event runoff or alteration
391 of water sources and flowpaths. Figure 11 shows the relationships between
392 dissolution rate and these two physical drivers over a variety of timescales
393 from daily (a,b), to weekly (c,d), to monthly (e,f). Generally, when airflow
394 velocity is positive (downdraft) and discharge is high, dissolution rate is high.

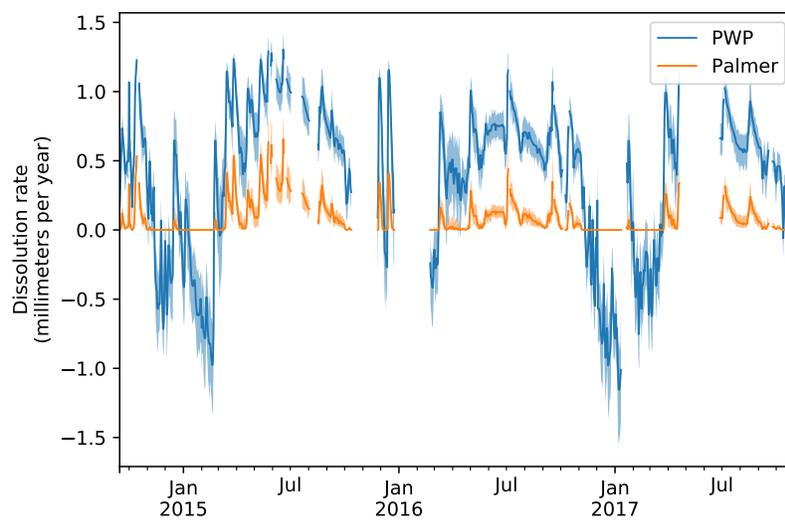


Figure 9: Calcite dissolution rates calculated from $p\text{CO}_2$ and SpC time series, where the blue line depicts rates calculated using the PWP equation (which includes negative values) and the yellow line indicates rates calculated using the Palmer equation. The shaded bands indicate a 90% confidence interval based on the Monte Carlo error propagation. The data indicate a regular seasonal pattern in dissolution rate variability, with undersaturated conditions typical in the summer and supersaturated conditions typical in the winter.

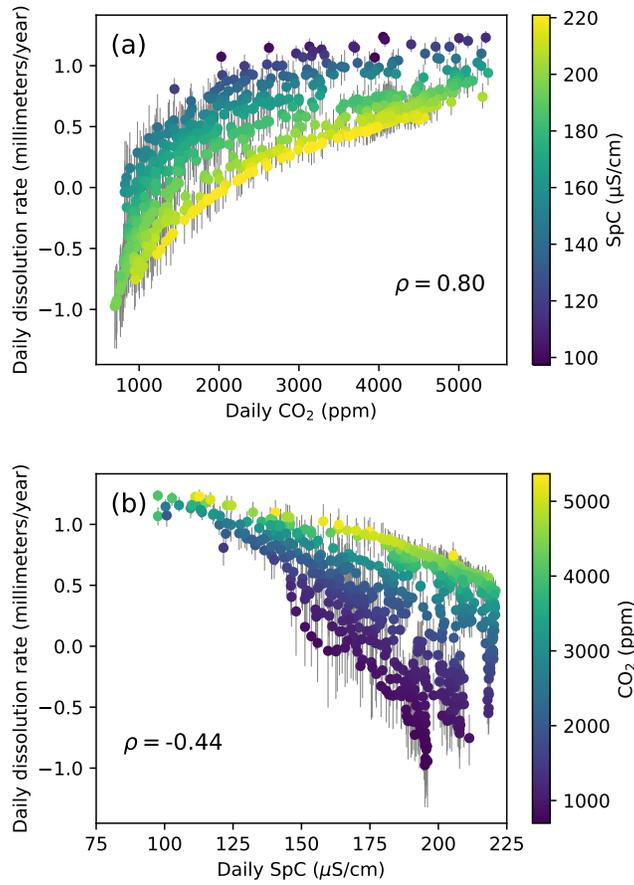


Figure 10: Relationships between daily averaged dissolution rates and (a) pCO₂ (parts per million) or (b) SpC (microsiemens per centimeter), where correlations are quantified using Spearman's rank correlation coefficient, ρ . Error bars depict a 90% confidence interval based on the Monte Carlo error propagation.

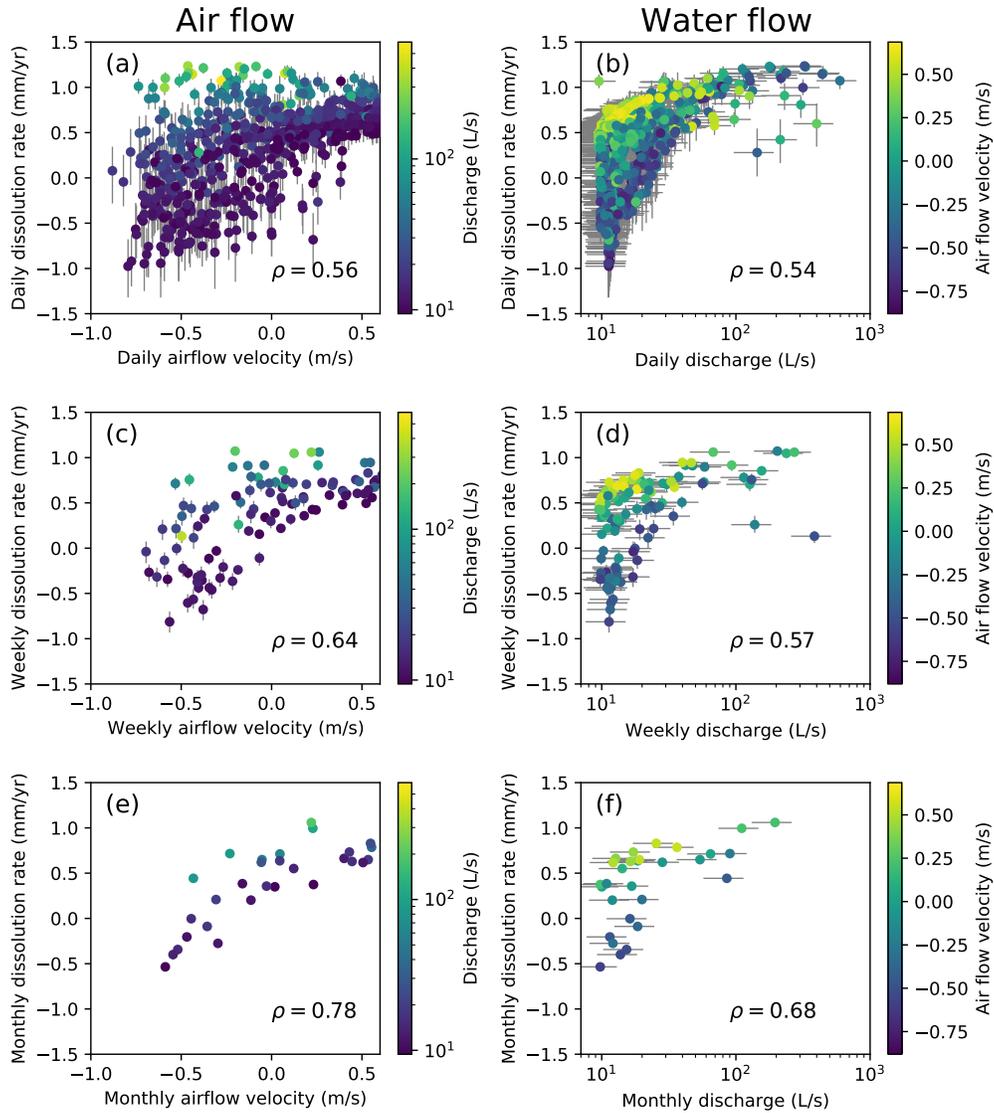


Figure 11: Relationships between dissolution rates and airflow velocity (left column: a,c,e) or stream discharge (right column: b,d,f). Each row represents rates averaged over a different time period, from daily (a,b), to weekly (c,d), to monthly (e,f). Color/shading in the left column indicates discharge and in the right column indicates air flow velocity, and correlations are quantified using Spearman's rank correlation coefficient, ρ . Error bars on dissolution rates depict a 90% confidence interval based on the Monte Carlo error propagation. Error bars on discharge indicate RMSE. Units are: mm/yr = millimeters per year; m/s = meters per second; L/s = liters per second.

395 At all timescales, cave airflow displays a stronger correlation with dissolu-
396 tion rate than discharge, though at the daily timescale the correlations are
397 effectively the same. The strength of the correlation between airflow velocity
398 and dissolution rate increases with the duration of the averaging. Correla-
399 tion with discharge is similar for daily and weekly timescales and somewhat
400 higher over the monthly timescale. All correlations have p-values < 0.0001.
401 Error is known to attenuate correlation (Spearman, 1904), such that large
402 errors will result in bias toward lower correlation. As a result, true correla-
403 tions are likely modestly higher than the estimates here, particularly for the
404 daily timescale where uncertainties are largest. However, even for the daily
405 timescale, the measured natural variations are substantially larger than the
406 estimated uncertainties.

407 Since cave airflow velocity emerges as the strongest external driver of
408 dissolution rate variability, and because airflow direction is likely to be the
409 most important factor in determining CO₂ concentrations, we divide the
410 record into days when airflow is on average updraft (winter regime) and
411 downdraft (summer regime). Dissolution rates are higher during periods of
412 downdraft, when the cave air has higher CO₂ (Figure 12). Interestingly, there
413 is also a strong contrast in the variability of dissolution rates during the two
414 airflow regimes, with periods of updraft having much larger variability in
415 rates. This effect is further considered below as we examine how dissolution
416 rates vary during storms.

417 *4.5. Dissolution rate variation during storms*

418 To explore how dissolution rates vary during storms, we first examine rela-
419 tionships between dissolved CO₂ and discharge, because CO₂ is the chemical
420 parameter most strongly correlated with changes in dissolution rate (Figure
421 10). Since dissolution rates show more variability during upward airflow,
422 one hypothesis might be that airflow direction somehow modulates the vari-
423 ability caused by changes in discharge. As an initial test of this hypothesis,
424 we examine the relationship between daily averaged values of dissolved CO₂
425 and discharge, separated into groups of downdraft and updraft conditions
426 (Figure 13). Dissolved CO₂ is correlated with discharge during periods of
427 updraft ($\rho = 0.31$, p-value < 0.0001), whereas there is no statistically sig-
428 nificant correlation between dissolved CO₂ and discharge during periods of
429 downdraft airflow ($\rho = 0.05$, p-value = 0.38). Again, due to uncertainties in
430 discharge and resulting attenuation of correlation, the real correlations are
431 likely modestly larger than these estimates. However, since the separation in

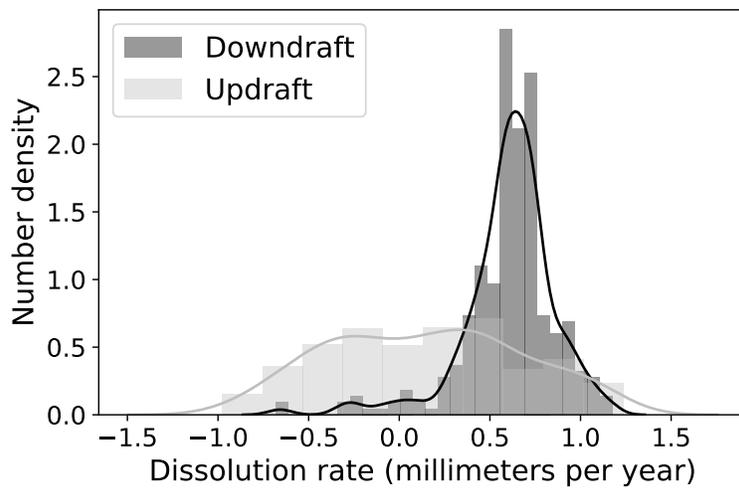


Figure 12: Distribution of dissolution rates under different airflow regimes for daily averaged dissolution rates under both downdraft (dark gray) and updraft (light gray) conditions. During downdraft, dissolution rates are typically high. During updraft, dissolution rates are typically lower; however, they are also much more variable. Kernel density estimates are shown (solid lines) to aid visual distinction of the two overlapping distributions.

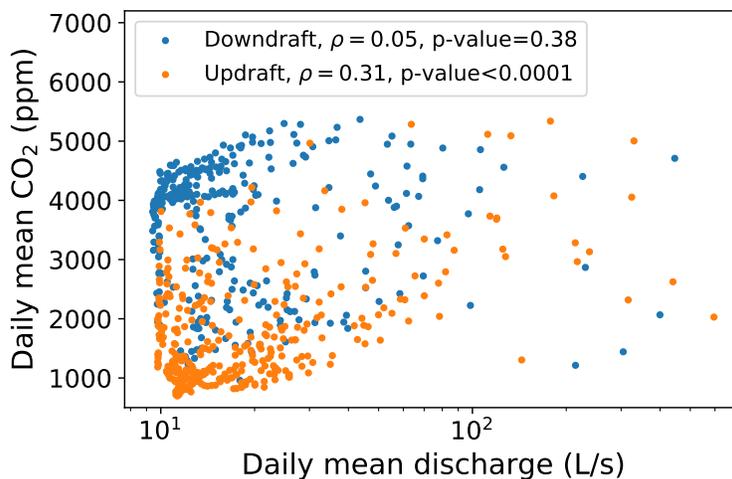


Figure 13: Relationship between discharge and CO_2 under different airflow regimes. Daily averaged values of discharge and CO_2 for periods of either downdraft (blue) or updraft (orange). Spearman's rank correlation coefficients, ρ , and respective p-values indicate that there is a moderate correlation between discharge and CO_2 during periods of updraft airflow but no statistically significant correlation during periods of downdraft airflow. Units are: ppm = parts per million; L/s = liters per second.

432 ρ and p-values is quite large, we do not expect this to influence the general
 433 pattern.

434 To further examine the possibility that airflow modulates discharge-driven
 435 dissolution rate variation during storms, we plot time series of chemistry and
 436 estimated dissolution rates during storm events. Two typical examples are
 437 shown in Figure 14, one during downdraft (summer) conditions and one
 438 during updraft (winter) conditions.

439 The winter storm produces more variation in dissolved CO_2 , which ranges
 440 from about 1,000 ppm to 3,500 ppm. Gaseous CO_2 remains low during most
 441 of the event because of the dominance of updraft conditions, which bring
 442 outside air quickly to the sensor location from the cave entrance. SpC varies
 443 from near maximal values, about 220 $\mu\text{S}/\text{cm}$, to 115 $\mu\text{S}/\text{cm}$. Driven by
 444 both changes in CO_2 and dissolved load, the dissolution rate changes sharply
 445 during the storm, from supersaturated conditions (-0.25 mm/yr) before the
 446 storm to highly undersaturated conditions (1.2 mm/yr) near the peak of the
 447 event. During the winter storm, estimated discharge ranged from 30 L/s to

448 330 L/s.

449 During the summer storm, downdraft conditions prevail, and, conse-
450 quently, CO₂ concentrations in the air remain relatively high at about 3,000 ppm,
451 except for during two brief periods of airflow reversal that follow the storm.
452 Dissolved CO₂ is already high (4,000 ppm) before the start of the event and
453 peaks about 5,000 ppm during the event. Therefore, there is much less vari-
454 ability of dissolved CO₂ during the summer storm than during the winter
455 storm. SpC decreases during the storm from about 255 $\mu\text{S}/\text{cm}$ to about
456 220 $\mu\text{S}/\text{cm}$, displaying less variability than during the winter storm. Disso-
457 lution rate also displays less variability, with rates about 0.5 mm/yr before
458 the storm and 0.8 mm/yr at the peak. During the summer storm, estimated
459 discharge ranged from 8.4 L/s to 80 L/s.

460 In general, winter and spring (periods of mostly updraft) have higher av-
461 erage values of discharge (Figure 3), as might be expected from lower rates
462 of evapotranspiration during these cooler periods. Therefore, one possibility
463 is that the correlation between airflow direction and dissolution rate variabil-
464 ity (Figure 12) is spurious and is actually driven by differences in discharge
465 dynamics during these seasons. Because discharge has a strong negative
466 correlation to dissolved load (Figure 4), storms with greater discharge varia-
467 tion should also have greater variation in dissolved Ca, and this could drive
468 greater variation in dissolution rate.

469 To explore this possibility we identify all storm events and calculate the
470 range of dissolution rate, the average airflow velocity, and the range of dis-
471 charge during each storm over the period of record. The beginnings of storms
472 were defined as increases in discharge of at least a factor of two within a period
473 of less than two days. The end of a storm event was defined to be a return to
474 130% of the pre-storm discharge or one week after the increase in discharge,
475 whichever was shorter. Because most of the chemical variation occurs during
476 the rising limb, the dissolution rate ranges are not particularly sensitive to
477 the criteria for the end of a storm event. However, including these crite-
478 ria enables treatment of multi-peak events as a single storm. We find that
479 change in dissolution rate within a storm was much more strongly correlated
480 to mean cave air velocity during the storm ($\rho = -0.85$, p-value=0.0004) than
481 it was to the magnitude of the change in discharge ($\rho = 0.23$, p-value=0.45),
482 as is shown in Figure 15 for the storm events for which complete chemical
483 records exist. This suggests that cave airflow direction is an important con-
484 trol on dissolution rate variation during storms, and that storm dissolution
485 rate variability is not primarily driven by dilution.

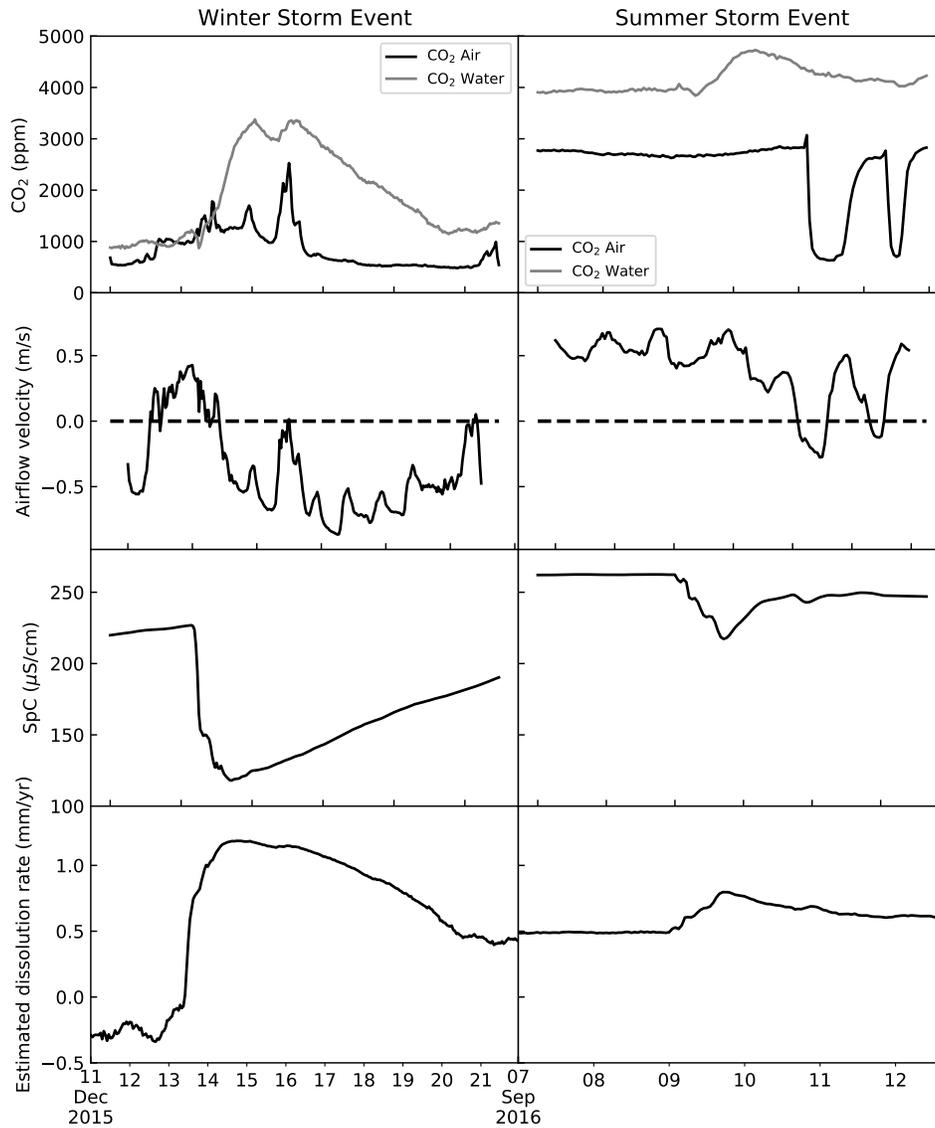


Figure 14: Variability of CO₂, airflow velocity, specific conductance (SpC), and estimated calcite dissolution rate during two example summer and winter storm events. The winter event, when airflow is primarily updraft (negative) exhibits more variation than the summer event in both dissolved CO₂ and SpC, and consequently in dissolution rate. The summer event exhibits low chemical variability. Units are: ppm = parts per million; m/s = meters per second; $\mu\text{S}/\text{cm}$ = microsiemens per centimeter; mm/yr = millimeters per year.

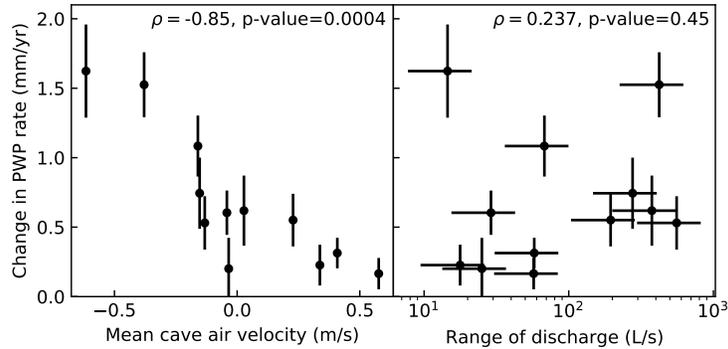


Figure 15: Correlations between storm dissolution rate range and potential controls. Average airflow velocity during a storm event is highly correlated with the range in dissolution rates during the event. The range of discharge within the event is not significantly correlated with the range in dissolution rates. Data are shown for all identified storm events during the study period for which complete chemical records were available. Error bars on dissolution rates depict a 90% confidence interval based on the Monte Carlo error propagation. Error bars on discharge indicate RMSE propagated through the difference used to calculate range.

486 5. Discussion

487 5.1. Controls of dissolution rate variability

488 The concept of the magnitude and frequency of erosional forces is central to understanding how temporal variations in the rates of geomorphic processes influence the long-term rates of landscape evolution and the morphologies that develop (Wolman and Miller, 1960). This concept is most frequently applied in fluvial systems, where frequency relates to the recurrence interval of streamflows of different magnitude. However, magnitude and frequency have also been discussed in the context of weathering processes, such as chemical solution (Goudie and Viles, 1999).

496 A variety of studies have quantified variation in rates of chemical geomorphic work at the basin scale by examining the rate of solute export as a function of river streamflow (Wolman and Miller, 1960; Gunn, 1982; Schmidt, 1985). However, quantifying magnitude and frequency within the context of weathering presents some challenges, and, particularly one must be clear as to the specific process that one is attempting to quantify (Goudie and Viles, 1999). Calculating chemical weathering rates using river solutes provides a

503 quantification of the magnitude and frequency attributes of solute export
504 from a basin. However, since solutes are stored within the basin for some
505 unknown time, these rates are removed from the rates of actual detachment
506 of the ions from mineral surfaces. For example, Covington et al. (2015) show
507 that in-stream calcite dissolution rates may be orders of magnitude higher
508 than basin-wide denudation rates derived from basin solute export. One
509 can imagine, similarly, that the time variability of rates of dissolution on
510 karst landscape surfaces might be quite different than the time variability in
511 basin-wide solute export rates.

512 Relatively few studies have attempted to quantify the time-variation of
513 calcite dissolution rates within karst streams or caves, or to understand
514 the controls of this variability (Groves and Meiman, 2005; Palmer, 2007b;
515 Galdenzi, 2012; Covington et al., 2015; Covington and Vaughn, 2019). The
516 central goal of this study was to examine variability in dissolution rates
517 within a specific cave stream and to develop a mechanistic understanding of
518 the controls on that variability.

519 As in previous studies (Groves and Meiman, 2005; Covington et al., 2015;
520 Covington and Vaughn, 2019), we find that the strongest chemical driver of
521 variation in dissolution rates is variation in dissolved CO_2 , which shows much
522 stronger correlation with dissolution rate at our site than does dissolved load
523 (Figure 10). In turn, dissolved CO_2 displays a strong seasonal pattern. This
524 seasonal pattern is strongly correlated with seasonal changes in the cave air
525 CO_2 that are driven by the direction of cave airflow (Figures 3a-b and 8),
526 which is ultimately controlled by the temperature difference between cave air
527 and outside air (Equation 5). A conceptual sketch of the interactions between
528 these processes is shown in Figure 6. Review of time series data over shorter
529 timescales (days to weeks) provides even stronger evidence for a mechanistic
530 connection between cave airflow and dissolved CO_2 , where switches in airflow
531 direction strongly perturb CO_2 concentrations in the cave atmosphere, and
532 the dissolved CO_2 in the cave stream responds in a lagged and muted fashion,
533 decreasing during periods of low cave air CO_2 and increasing during periods
534 of high cave air CO_2 (Figure 7).

535 Review of the entire dataset shows that, perhaps surprisingly, dissolution
536 rate is more strongly correlated with cave airflow velocity than with discharge
537 (Figure 11). The difference in the correlation strength increases moving from
538 daily to monthly timescales, suggesting that cave airflow is most important
539 in impacting the seasonal pattern though still has a strong impact on the
540 timescales of storms. Similar observations have been made on the impact of

541 cave ventilation on the saturation state of drip water and resulting seasonal
542 biases within speleothem records (Spötl et al., 2005; Banner et al., 2007;
543 Wong et al., 2011), and ventilation has also previously been argued to impact
544 spatial or temporal changes in cave stream dissolved CO₂ or dissolved load
545 (Troester and White, 1984; Jeannin et al., 2017). Gulley et al. (2014) also
546 found that seasonal cave ventilation patterns can explain seasonal changes
547 to dissolved load and dissolved CO₂ within a water table cave in Florida.
548 Therefore, the patterns observed here cohere with previous studies, though
549 our data provide much higher time resolution to examine the connections
550 between ventilation and cave stream saturation state in more detail.

551 Though the time series data suggest that cave airflow direction is an
552 important control of the seasonal oscillation of dissolved CO₂, there may be
553 additional drivers. Specifically, CO₂ production in the soil through microbial
554 decay and root respiration is known to vary with surface temperature and
555 solar radiation (Hibbard et al., 2005; Lloyd and Taylor, 1994). At a nearby
556 site with a similar hydrogeological setting, Covington and Vaughn (2019)
557 observed a strong seasonal signal (range \approx 20,000 ppm) in dissolved CO₂ at
558 Langle Spring, Arkansas, which is thought to drain an unventilated portion of
559 the karst aquifer. They hypothesized that this signal derived from seasonal
560 changes in subsurface CO₂ production. It is uncertain how much of the
561 seasonal signal in dissolved CO₂ at Blowing Springs might also be a function
562 of changes in the rate of CO₂ production.

563 In contrast to the seasonal respiration-driven pattern, some karst aquifers
564 that exhibit higher pCO₂ at depth than in the soil have very little seasonal
565 CO₂ variation at depth (Atkinson, 1977a). A prior study at Blowing Springs
566 Cave measured soil CO₂ concentrations, with summer values frequently be-
567 ing above what we observe in the cave stream and winter values frequently
568 being below (Knierim et al., 2017). Additionally, the study used stable car-
569 bon isotopes to quantify the mixture of atmospheric CO₂ versus unsaturated
570 zone CO₂ (produced via respiration/decomposition) in the cave atmosphere.
571 Knierim et al. (2017) found that the proportion varied seasonally and, addi-
572 tionally, that there were different mixing lines for each season, highlighting
573 seasonally variable unsaturated zone CO₂ sources. At the least, these obser-
574 vations suggest that there is some storage of CO₂ in the unsaturated zone
575 that might reduce seasonal variation in the cave. The few available on-site
576 measurements of dissolved CO₂ at the upstream sump in Blowing Springs
577 Cave indicate a range of approximately 1,500 ppm between summer and win-
578 ter measurements (Young, 2018), in contrast to the range of approximately

579 5,000 ppm that we observe near the downstream end of the cave. However,
580 it is also unclear how much ventilation might occur within the portion of the
581 aquifer that is upstream of the sump. Therefore, whereas there is a clear
582 impact of cave ventilation on the annual CO₂ cycle, there may also be a sea-
583 sonal signal driven by production. The magnitude of that production signal
584 is uncertain.

585 The mechanistic link between cave airflow direction and dissolved CO₂ in
586 the stream is generated because the primary CO₂ source for the cave air can
587 either be upwind or downwind of the main cave stream (Figure 6). The pri-
588 mary source of CO₂ to the cave atmosphere is a CO₂ reservoir within the soil
589 and unsaturated zone above the cave. During periods of downdraft (summer
590 regime) ventilation brings gases from this reservoir into the cave, maintain-
591 ing a high pCO₂ within the cave air that limits degassing of CO₂ from the
592 cave stream (Figure 6a). During periods of updraft (winter regime) ventila-
593 tion brings fresh outside air into the cave, reducing the pCO₂ of the cave air
594 and enhancing degassing of CO₂ from the stream (Figure 6b). Though the
595 cave has strong ventilation during both summer and winter conditions, the
596 restricted nature of the airflow pathways through the unsaturated zone must
597 produce a sufficiently high surface area to volume ratio that air transiting
598 this zone obtains a high pCO₂.

599 *5.2. Dissolution rate variability during storms and the role of airflow*

600 Whereas storms play a secondary role in driving variability in dissolu-
601 tion rates, there are still statistically significant correlations ($\alpha=0.05$)
602 between discharge and dissolution rate (Figure 11b). We can observe these
603 variations clearly on the basis of individual storms, and see that they are
604 driven by a combination of dilution and increasing dissolved CO₂ (Figure
605 14). Interestingly, airflow direction also appears to modulate the dissolution
606 rate variability within storms, with greater storm variability during updraft
607 conditions. This is supported by at least three observations:

- 608 1. Variation in dissolution rates is much greater during updraft than
609 downdraft (Figure 12);
- 610 2. Dissolved CO₂ is positively correlated with discharge during updraft
611 but not during downdraft (Figure 13);
- 612 3. Dissolution rate range during individual storms is correlated to the
613 airflow velocity but not to the range of discharge during the storm
614 (Figure 15).

615 It is perhaps counterintuitive that cave airflow direction should have any
616 importance for dissolution rate variation during storms. However, the ob-
617 served pattern can be explained using an existing conceptual model for un-
618 saturated zone CO₂ within karst (Mattey et al., 2016) and a basic mathe-
619 matical framework for transport of CO₂ within the karst unsaturated zone
620 (Covington, 2016). Mattey et al. (2016) argue, based on eight years of field
621 measurements at the Rock of Gibraltar and other observations of deep CO₂
622 within karst systems (Atkinson, 1977a; Wood, 1985; Wood and Petraitis,
623 1984), that karst unsaturated zones contain a body of “ground air,” which is
624 a reservoir of CO₂ produced by the microbial decay of organic matter that
625 has infiltrated to depth. Cave air is considered to be a mixture of surface air
626 with ground air, where the percentages depend largely on the outside temper-
627 ature and the resulting direction of air circulation through the unsaturated
628 zone.

629 Other work has suggested that the CO₂ in cave air is often associated
630 with root respiration of the deepest rooting plants (Breecker et al., 2012),
631 again suggesting production at depth. At Blowing Springs Cave, the carbon
632 isotope ratios of CO₂ are consistent with soil/root respiration (Knierim et al.,
633 2017) (with summer values around -21‰), so it is unclear whether the source
634 of deep unsaturated zone CO₂ might be particulate organic matter or from
635 root respiration. However, to explain the observations, we hypothesize that
636 there is a substantial volume of CO₂ stored at depth in the unsaturated zone.

637 During winter (periods of cave updraft), storms bring water that is charged
638 with CO₂, frequently 2,000–4,000 ppm. These concentrations are substan-
639 tially higher than typically observed in soil at the site during fall/winter
640 (1,500 ppm) in a prior study (Knierim et al., 2017). This observation sup-
641 ports the conception of a reservoir of high pCO₂ within the unsaturated
642 zone (Atkinson, 1977a; Mattey et al., 2016). Additionally, a simple model of
643 CO₂ transport within a vertical fracture suggests that vertical flow of water
644 through karst fractures can efficiently redistribute CO₂ within the subsurface,
645 pushing it to greater depth (Covington, 2016). Observations of hysteresis be-
646 tween discharge and dissolved CO₂, with higher CO₂ during the recession,
647 have also been interpreted as indicating that later arriving diffuse recharge
648 water can transport soil and unsaturated zone CO₂ into karst conduits (Ves-
649 per and White, 2004). Therefore, it is physically plausible that vertical flow
650 of water through the unsaturated zone during a storm could effectively trans-
651 port a pulse of CO₂ to the water table.

652 Downward transport of CO₂ during winter storms produces variation in

653 the cave stream because the winter airflow regime has disconnected the cave
654 stream from the CO₂ source (Figure 6b), reduced the pCO₂ of the cave
655 stream, and the pulse of high CO₂ has a large effect. Conversely, in the
656 summer airflow regime (downdraft) the cave air is already in contact with
657 the ground air (Figure 6a), because the air is entering the cave via the soil
658 and unsaturated zone. Therefore, degassing is reduced and the cave stream
659 is maintained at high pCO₂. Consequently, summer storms produce much
660 less variation in CO₂ within the cave stream and therefore less variation in
661 dissolution rate. This conceptual model is also supported by previous work
662 at Blowing Springs Cave where isotopic disequilibrium between dissolved
663 inorganic carbon (DIC) in the cave stream and CO₂ in the cave air was
664 greater during winter periods, when the cave stream is disconnected from
665 the CO₂ source, but approached equilibrium during summer, when cave air
666 CO₂ was higher (Knierim et al., 2017).

667 5.3. *Dissolution rate variation in the context of similar studies*

668 Since discharge is not the primary driver of variation in dissolution rates
669 at the study site, normal concepts of magnitude and frequency break down, as
670 they are based on flood recurrence intervals. To estimate rates of geomorphic
671 work in the cave stream, we are better off asking, “Which way is it blowing?”
672 rather than, “How much is it flowing?” However, this pattern is seemingly
673 not a universal one, and it is worth putting into the context of the limited
674 set of other studies of dissolution rate variation in karst conduits.

675 First we compare against the nearby study of Langle and Copperhead
676 Springs (Covington and Vaughn, 2019), located in the same limestone layer
677 and climate setting as Blowing Springs Cave. These two springs compose a
678 karst underflow-overflow system, where Langle Spring is completely phreatic
679 and carries most of the flow at low discharge. Langle and Copperhead Spring
680 both exhibit strong seasonal CO₂ variation that is the strongest control on
681 dissolution rate. Data suggest that Langle drains a relatively small phreatic
682 conduit, which has no ability to ventilate. Langle Spring has the highest
683 variation in CO₂ concentration of any of the available studies of dissolution
684 rates within karst conduits, with summer values that exceed 20,000 ppm and
685 winter values about 3,000 ppm. One potential reason for the higher CO₂
686 concentrations and strong production-related signal is that land use in the
687 spring recharge zone is predominantly pasture, and grasslands have higher
688 CO₂ production rates than forested areas (Smith and Johnson, 2004; Knierim
689 et al., 2015a, 2017). Copperhead Spring has peak values in early summer

690 about 15,000 ppm and then late summer values about 5,000–6,000 ppm,
691 which are similar to peak summer values at Blowing Spring. The sudden
692 decrease in CO₂ at Copperhead Spring in the early summer coincides with
693 a discharge threshold. Below this threshold, the cave system feeding this
694 spring begins to ventilate, and CO₂ decreases dramatically as a result of the
695 onset of ventilation (Covington and Vaughn, 2019). Therefore, if we want to
696 estimate dissolution rates at Langle Spring, we need to consider variability in
697 soil CO₂ production, and might ask ourselves, “Is it growing?” Whereas, at
698 Copperhead Spring, which is intermittently ventilated, it is more important
699 to ask, “Is it blowing?”

700 In the two other cave streams where dissolution rate or saturation state
701 has been quantified as a function of discharge (Groves and Meiman, 2005)
702 or recurrence interval (Palmer, 2007b), the cave water was supersaturated
703 during most of the study period, with only short periods of active dissolution
704 occurring at high flow. Groves and Meiman (2005) studied the Logsdon River
705 in Mammoth Cave, Kentucky, and Palmer (2007b) studied McFail’s Cave,
706 New York. Both studies found that the majority of the dissolution occurs in
707 the top 5% flow regime. Therefore, these sites fall more into the standard
708 magnitude and frequency framework, where active dissolution is driven by
709 high flow events.

710 One reason for the tendency toward supersaturation at these two sites
711 may be that they are more highly ventilated than any of the other study sites.
712 Mammoth Cave is the longest cave in the world (Gunn, 2004), has many en-
713 trances, and is, consequently, well-ventilated. This high density of entrances
714 may produce relatively low CO₂ concentrations in the cave air during all sea-
715 sons. Therefore, water flows through the soil and unsaturated zone dissolving
716 calcite under relatively high pCO₂ conditions, then it enters the cave stream,
717 is brought to much lower pCO₂, and becomes supersaturated. Storm events
718 may in part increase dissolution rates by reducing ventilation when portions
719 of the system flood to the ceiling. During the largest flood event in the study,
720 Logsdon River remained under pipefull flow conditions for 114 hours (Groves
721 and Meiman, 2005). An additional factor that may create variability with
722 discharge is the nature of recharge to the system. Approximately 40% of the
723 recharge to Logsdon River is allogenic (from streams flowing off of sandstone
724 caprock). It is plausible that flow from non-carbonates, and changes to the
725 percentage of that flow during floods, could increase the sensitivity of dis-
726 solution rates to discharge (Atkinson, 1977b; Scanlon and Thrailkill, 1987;
727 Worthington et al., 1992). Palmer (2007b) also describes McFail’s Cave as

728 “well-aerated,” and suggests that the cave stream is supersaturated because
729 of ventilation and degassing of CO₂. Therefore, it is plausible that episodic
730 storm-driven dissolution is a common pattern within highly ventilated karst
731 conduit systems, which typically have low concentrations of dissolved CO₂.

732 5.4. *The role of ventilation over the history of cave evolution*

733 Taken within the context of prior studies (Groves and Meiman, 2005;
734 Palmer, 2007b; Covington and Vaughn, 2019), the data presented here elu-
735 cidate how ventilation may drive changes in dissolution rates within karst
736 conduits as they evolve. The observed behaviors can be arranged on an axis
737 of increasing ventilation (Figure 16). Except during periods of base-level
738 aggradation, karst systems will also tend to evolve along this axis over time,
739 from no ventilation at the beginning toward highly ventilated as they mature.

740 During the first stage of karst conduit evolution, the pre-breakthrough
741 stage (Figure 16a), the penetration length of undersaturated water is less
742 than the length of the incipient conduit (Dreybrodt, 1996; Covington et al.,
743 2012). Consequently, calcite dissolution under closed-system conditions within
744 the flowpath leads to the consumption of CO₂, which greatly reduces disso-
745 lution rates at depth.

746 Once breakthrough occurs, and the penetration length exceeds the flow-
747 path length, then water can traverse the conduit without substantially re-
748 duced pCO₂ despite the closed-system conditions (Covington and Vaughn,
749 2019). This is the stage that we observe at Langle Spring (Figure 16b), the
750 pattern that we refer to as, “Is it growing?” This stage shows the highest
751 average dissolution rates among the study sites compared here. These high
752 rates are maintained because the water is at high pCO₂ and has no means
753 of degassing that CO₂. At Langle Spring, there is a strong seasonal signal
754 driven by CO₂ production. However, some karst springs have very low an-
755 nual variation in pCO₂ (Atkinson, 1977a), so this seasonal pattern is not
756 universal. Why some karst systems have a strong production-related signal,
757 and some do not, remains an open question.

758 The third stage, “Is it blowing?” represents the onset of intermittent
759 ventilation (Figure 16c), where the karst system is intermittently ventilated.
760 For Copperhead Spring, this switch is driven by changes in water level. The
761 temporal changes in dissolution rate at this site show a seasonal signal, but
762 superimposed on that seasonal signal is a strong switching behavior where
763 periods of ventilation dramatically reduce the dissolution rates.

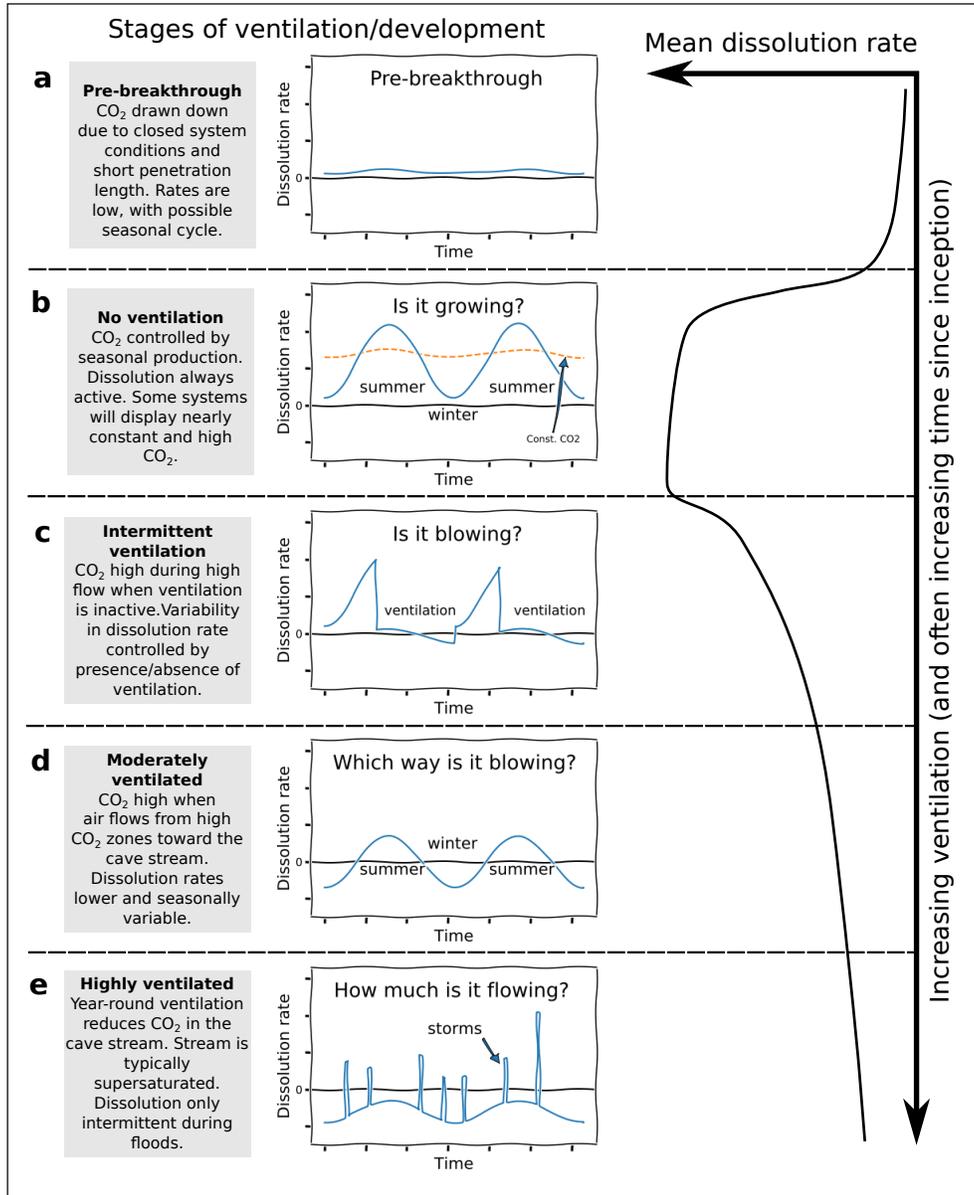


Figure 16: Patterns of observed dissolution rate variation from this and other studies and how they relate to ventilation strength and resulting CO₂ dynamics. Except during periods of base-level aggradation, caves will typically evolve toward being more ventilated over time.

764 The fourth stage, moderately ventilated, is observed at Blowing Springs
765 Cave (Figure 16d). During this stage the conduit undergoes continuous ven-
766 tilation. However, the direction of airflow strongly impacts dissolution rates.
767 We ask, “Which way is it blowing?” The seasonal ventilation patterns that
768 are driven by chimney effect airflow create a seasonal pattern in dissolution
769 rate as the CO₂ source switches between being upwind and downwind of the
770 cave stream. The seasonal pattern is more muted than in the previous two
771 stages, and the average dissolution rates are lower.

772 At Blowing Springs, we see secondary variability driven by storms, which
773 typically increase dissolution rates. However, even this storm variation is
774 modulated by airflow direction. To create the moderately ventilated pattern
775 of dissolution rate variability, the CO₂ of air passing through the zone of CO₂
776 sources must be strongly influenced by those sources. The exact physical
777 requirements for this influence are unclear. However, the rate and spatial
778 distributions of CO₂ production may be important. Furthermore, the air
779 pathways must have a sufficiently high surface area to volume ratio in order
780 to create effective exchange of CO₂. This may be more likely if airflow is
781 divided between many smaller pathways. Clearly, if an air pathway is too
782 open, then it will rapidly bring in outside air that reduces the pCO₂.

783 The final stage is a highly ventilated cave (Figure 16e) and is illustrated
784 by prior studies at Mammoth Cave and McFail’s Cave. Here, we return to
785 the more standard framework for considering variation in geomorphic work
786 within a stream, “How much is it flowing?” Within these systems, venti-
787 lation is sufficiently strong that the stream is normally in supersaturated
788 conditions. There may still be a seasonal variation in CO₂ (Groves and
789 Meiman, 2005), but dissolution primarily occurs during short-term high-flow
790 events. This variation may be driven by dilution, particularly in the case of
791 allogenic recharge, and may also be driven by temporary shutoff or reduction
792 of ventilation as many conduits transition into full pipe conditions during a
793 flood. Pulses of CO₂ brought through the unsaturated zone by water may
794 also impact cave stream CO₂, as observed at Blowing Springs Cave.

795 After initial conduit breakthrough (2nd stage, Figure 16), the overall pat-
796 tern is one of increasing ventilation and, as a result, decreasing pCO₂ and
797 decreasing dissolution rates (Palmer, 2007b). Therefore, we might expect
798 that chemical erosion rates within cave streams gradually reduce over the
799 history of evolution, except perhaps during periods of base level rise, where
800 more conduits would become flooded. This trend toward reduced chemical
801 erosion rates over time also has implications for the importance of mechan-

802 ical erosion within cave streams. Since instantaneous chemical erosion rates
803 are limited to relatively low magnitudes in comparison to mechanical erosion
804 (Covington et al., 2015), this result suggests that mechanical erosion pro-
805 cesses should become much more important once caves are well-ventilated.
806 For the well-ventilated end member, only intermittent dissolution is observed
807 during floods. These same flood events are likely to overcome thresholds for
808 transport of sediment and consequent mechanical erosion. Using the tortoise
809 and the hare analogy (Simms, 2004), chemical erosion processes are most
810 effective when they occur nearly continuously (tortoise). If chemical erosion
811 processes become intermittent, mechanical erosion is likely to dominate.

812 While we have sketched a broad hypothesis about how ventilation controls
813 the rate of calcite dissolution within karst conduits, and how that role
814 might evolve as a karst system matures, the observed patterns come from a
815 relatively limited set of karst systems that are far from spanning the full range
816 of climatic and geological settings of karst. Therefore, there are likely other
817 potential controls on dissolution rate variability and perhaps other ways in
818 which ventilation interacts with CO₂ dynamics. The conceptualization in
819 Figure 16 is relatively simplistic, and it seems likely to grow in complexity as
820 further sites are studied and more dimensions of the problem are understood.
821 Importantly, all of the sites discussed are dominated by autogenic recharge.
822 It seems plausible that sites dominated by allogenic recharge will display
823 somewhat different dynamics. For example, ventilation may not bring wa-
824 ter to a supersaturated state, because dissolved load is always sufficiently
825 low. Dilution may be more important. However, patterns of CO₂ production
826 and degassing have also been shown to control spatial patterns of dissolution
827 within allogenic recharge systems (Covington et al., 2013).

828 Here we have categorized each study site into a single pattern/stage of
829 Figure 16, but most karst systems will contain a range of ventilation con-
830 ditions within them. Therefore, the presented stages may also represent
831 spatial contrasts in dissolution rate dynamics within different portions of a
832 karst system that have different ventilation strengths. Processes such as CO₂
833 production, ventilation, and gas exchange are currently absent from numeri-
834 cal models of speleogenesis. Developing and exploring mathematical models
835 for these processes would aid future understanding of the long-term inter-
836 actions among ventilation, CO₂ dynamics, and calcite dissolution and how
837 they influence the rates and patterns of cave development.

838 **6. Conclusions**

839 We collected time series data from a stream cave in Arkansas to study the
840 temporal variation in calcite dissolution rates and the factors that drive them.
841 Ventilation of the cave atmosphere is driven by external temperature changes
842 through the process of chimney effect airflow. The direction of air flow is
843 the primary control on gaseous CO₂ within the cave atmosphere, with low
844 CO₂ during periods with updraft, when the cave is effectively ventilated by
845 outside air, and high CO₂ during periods of downdraft, when outside air flows
846 through a zone of high CO₂ before entering the main cave passage. In turn,
847 dissolved CO₂ in the cave stream is strongly impacted by the concentration of
848 CO₂ in the cave atmosphere, generating a seasonal variation in dissolved CO₂
849 that emerges as the primary driver of dissolution rate variability within the
850 cave stream. Dissolution rate is more strongly correlated with cave airflow
851 direction than it is with discharge, indicating that the standard framework
852 of geomorphic work partitioned by flood stage is inappropriate for this site.
853 We also find that the variations of dissolution rates during individual storm
854 events are modulated by airflow direction, with more variation occurring
855 during updraft (winter) conditions. We compare the results from this study
856 with prior studies of dissolution rate variability within karst systems and
857 propose a preliminary framework to explain the different observed patterns
858 of dissolution rate variation along an axis of increasing cave ventilation. We
859 suggest that the onset of ventilation reduces the rates of chemical erosion
860 within karst systems, and that as karst systems mature they will generally
861 evolve toward greater ventilation and lower dissolution rates. This effect may
862 accentuate the importance of mechanical erosion during the later stages of
863 cave evolution.

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873 not imply endorsement by the U.S. Government.

874 **Data Availability**

875 The data used in this manuscript and the python code that was used to
876 analyze the data and create the figures are provided in a Github repository
877 that is archived on Zenodo: <https://doi.org/10.5281/zenodo.3839802>.

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