



$1 \qquad \text{Sensitivity of a leaf gas-exchange model for estimating paleoatmospheric CO_2}$

- 2 concentration
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11 **Abstract.** Leaf gas-exchange models show considerable promise as paleo-CO₂ proxies. They are largely 12 mechanistic in nature, provide well-constrained estimates even when CO_2 is high, and can be applied to 13 most subaerial, stomata-bearing leaves from C_3 taxa, regardless of age or taxonomy. Here we place 14 additional observational and theoretical constraints on one of these models, the "Franks" model. In 15 order to gauge the model's general accuracy in a way that is appropriate for fossil studies, we estimated 16 CO₂ from 40 species of extant angiosperms, conifers, and ferns based only on measurements that can be 17 made directly from fossils (leaf δ^{13} C and stomatal density and size) and a limited sample size (1-3 leaves per species). The mean error rate is 28%, which is similar to or better than the accuracy of other leading 18 19 paleo-CO₂ proxies. We find that leaf temperature and photorespiration do not strongly affect estimated 20 CO₂, although more work is warranted on the possible influence of O₂ concentration on 21 photorespiration. Leaves from the lowermost 1-2 m of closed-canopy forests should not be used 22 because the local air δ^{13} C value is lower than the global well-mixed value. Such leaves are not common 23 in the fossil record, but can be identified by morphological and isotopic means.

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26 1 Introduction

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28 Leaves on terrestrial plants are well poised to record information about the concentration of 29 atmospheric CO2. They are in direct contact with the atmosphere and have large surface-area-to-volume 30 ratios, so the leaf internal CO₂ concentration is tightly coupled to atmospheric CO₂ concentration. Also, 31 leaves are specifically built for the purpose of fixing atmospheric carbon into structural tissue, and face 32 constant selection pressure to optimize their carbon uptake relative to water loss. As a result, many 33 components of the leaf system are sensitive to atmospheric CO₂, and these components feedback on 34 one another to reach a new equilibrium when atmospheric CO₂ changes. In terms of carbon assimilation, 35 Farquhar and Sharkey (1982) modeled this system in its simplest form as:

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$$37 A_n = g_{c(tot)} \times (c_a - c_i), (1)$$

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where A_n is the leaf CO₂ assimilation rate (µmol m⁻² s⁻¹), $g_{c(tot)}$ is the total operational conductance to CO₂ diffusion from the atmosphere to site of photosynthesis (mol m⁻² s⁻¹), c_a is atmospheric CO₂

41 concentration (μ mol mol⁻¹ or ppm), and c_i is leaf intercellular CO₂ concentration (μ mol mol⁻¹ or ppm)

42 (see also Von Caemmerer, 2000).
 43 Rearranging Eq. (1) for atr

Rearranging Eq. (1) for atmospheric CO₂ yields:

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$$c_a = \frac{A_n}{g_{c(tot)} \times (1 - \frac{c_i}{c_a})}.$$
 (2)





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$$A_n = A_0 \frac{\left[\left(\frac{c_i}{c_a}\right)c_a - \Gamma^*\right]\left[\left(\frac{c_{i0}}{c_{a0}}\right)c_{a0} + 2\Gamma^*\right]}{\left[\left(\frac{c_{i0}}{c_a}\right)c_a + 2\Gamma^*\right]\left[\left(\frac{c_{i0}}{c_{a0}}\right)c_{a0} - \Gamma^*\right]},$$

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where Γ^* is the CO₂ compensation point in the absence of dark respiration (ppm) and the subscript "0" refers to conditions at a known CO₂ concentration (typically present-day). Equations (2) and (3) are then solved iteratively until the solution for c_a converges.

These gas-exchange approaches grew out of a group of paleo- CO_2 proxies based on the CO_2 sensitivity of stomatal density (*D*) or the similar metric stomatal index (Woodward, 1987; Royer, 2001). Here, the *D*-*c*_a sensitivity is calibrated in an extant species, allowing paleo- CO_2 inference from the same (or very similar) fossil species. These empirical relationships typically follow a power-law function (Wynn, 2003; Franks et al., 2014; Konrad et al., 2017):

 $c_a = \frac{1}{kD^{\alpha'}}$

6667 where *k* and α are species-specific constants.

68 The related stomatal ratio proxy is simplified: *D* is measured in an extant species (D_0 , at present-69 day c_{a0}) and then the ratio of D_0 to *D* in a related fossil species is assumed to be linearly related to the 70 ratio of paleo- c_a to present-day c_{a0} (Chaloner and McElwain, 1997; McElwain, 1998):

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$$\frac{c_a}{c_{a0}} = k \frac{D_0}{D}.$$
(5)
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74 Equation (5) can be rearranged to match Eq. (4) but with α fixed at 1. Thus, paleo-CO₂ estimates using 75 the stomatal ratio proxy are based on a one-point calibration and an assumption that $\alpha = 1$; 76 observations do not always support this assumption (e.g., $\alpha = 0.43$ for *Ginkgo biloba*; Barclay and Wing, 77 2016). The scalar k was originally set at 2 for Paleozoic and Mesozoic reconstructions so that paleo- CO_2 78 estimates during the Carboniferous matched that from long-term carbon cycle models (Chaloner and 79 McElwain, 1997). For younger reconstructions, k is probably closer to 1 (by definition, k = 1 for present-80 day plants). We note that the stomatal ratio proxy was originally conceived as providing qualitative 81 information, only, about paleo-CO₂ (McElwain and Chaloner, 1995, 1996; Chaloner and McElwain, 1997; 82 McElwain, 1998) and has not been tested with dated herbaria materials or with CO₂ manipulation 83 experiments. At high CO_2 , the *D*- c_a sensitivity saturates, leading to uncertain paleo- CO_2 estimates, often with

At high CO₂, the *D*-*c*_a sensitivity saturates, leading to uncertain paleo-CO₂ estimates, often with unbounded upper limits (e.g., Smith et al., 2010; Doria et al., 2011). Stomatal density does not respond to CO₂ in all species (Woodward and Kelly, 1995; Royer, 2001), and because *D*-*c*_a relationships can be species-specific (that is, different species in the same genus with different responses; Beerling, 2005; Haworth et al., 2010), only fossil taxa that are still alive today should be used. The gas-exchange proxies partly address these limitations: 1) CO₂ estimates remain well-bounded—even at high CO₂—and their precision is similar to or better than other leading paleo-CO₂ proxies (~+35/-25% at 95% confidence; Franks et al., 2014); 2) the models are mostly mechanistic; that is, they are explicitly driven by plant





physiological principles, not just empirical relationships measured on living plants; 3) because the
 models retain sensitivity at high CO₂ and do not require that a fossil species still be alive today, much of
 the paleobotanical record is open for CO₂ inference, regardless of age or taxonomy; and 4) because the
 models are based on multiple inputs linked by feedbacks, they can still perform adequately even if one

96 or more of the inputs in a particular taxon is not sensitive to CO₂, for example stomatal density (Milligan
 97 et al., in review).

98 We note that the published uncertainties (= precision) associated with the stomatal density 99 proxies are probably too small because they usually only reflect uncertainty in the calibration regression 100 or in the measured values of fossil stomatal density, but not both; when this is done, errors often 101 exceed ±30% at 95% confidence (Beerling et al., 2009). Also, error rates in estimates from extant taxa 102 where CO₂ is known (= accuracy) are usually smaller with the stomatal density proxies (e.g., Barclay and 103 Wing, 2016), but this is expected because the same taxa have been calibrated in present-day (or near 104 present-day) conditions. Because the gas-exchange proxies are largely built from physiological 105 principles, they have less "recency" bias; that is, the gas-exchange proxies estimate present-day and 106 paleo-CO₂ with similar certainty when the same methods are used to determine the inputs.

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109 2 Study Aims and Methods

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111 Leaf gas-exchange proxies for paleo-CO₂ are becoming popular (Konrad et al., 2008; Grein et al., 112 2011b; Grein et al., 2011a; Erdei et al., 2012; Roth-Nebelsick et al., 2012; Grein et al., 2013; Franks et al., 113 2014; Maxbauer et al., 2014; Roth-Nebelsick et al., 2014; Montañez et al., 2016; Reichgelt et al., 2016; 114 Konrad et al., 2017; Tesfamichael et al., 2017; Kowalczyk et al., 2018; Lei et al., 2018; Londoño et al., 115 2018; Richey et al., 2018; Milligan et al., in review). However, many elements of these models remain 116 understudied. Here we investigate four such elements for the Franks et al. (2014) model: how does the 117 model perform across a large number of phylogenetically diverse taxa; and how is the model affected by 118 temperature, photorespiration, and proximity to the forest floor? We describe next the motivation and 119 details of the study design.

- 120
- 121 2.1 General testing in living plants

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Franks et al. (2014) tested the model on four species of field-grown trees (three gymnosperms and one angiosperm) and one conifer grown in chambers at 480 and 1270 ppm CO₂. The average error rate
(absolute value of estimated CO₂ minus measured CO₂, divided by measured CO₂) was 5%. Follow-up
work with three field-grown tree species (Maxbauer et al., 2014; Kowalczyk et al., 2018), CO₂

- experiments on seven tropical trees species (Londoño et al., 2018), and experiments on two fern and
- 128 one conifer species (Milligan et al., in review) indicate somewhat higher error rates (Fig. 1). Combined,
- the average error rate is 19% (median = 13%).
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Figure 1. Published CO₂ estimates using the Franks model for extant plants where the physiological inputs A_0 (assimilation rate at a known CO₂ concentration) and/or $g_{c(op)}/g_{c(max)}$ (ratio of operational to maximum leaf conductance to CO₂) were measured directly. Horizontal lines are the correct CO₂ concentrations. Uncertainties in the estimates correspond to the 16th-84th percentile range. Circles are from Londoño et al. (2018), squares from Milligan et al. (in review), large triangle from Maxbauer et al. (2014), small triangles from Kowalczyk et al. (2018), and diamonds from Franks et al. (2014).

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141 In these studies, two of the key physiological inputs were measured directly with an infrared gas 142 analyzer: the assimilation rate at a known CO_2 concentration (A_0) and/or the ratio of operational to maximum stomatal conductance to CO₂ ($g_{c(op)}/g_{c(max)}$, or ζ), the latter of which is important for 143 144 calculating the total leaf conductance $(q_{c(tot)})$. These two inputs cannot be directly measured on fossils; 145 thus, the error rates associated with Figure 1 may not be representative for fossil studies. Franks et al. 146 (2014) argue that within plant functional types growing in their natural environment, mean A_0 is fairly 147 conservative, leading to the recommended mean A_0 values in Franks et al. (2014) (12 µmol m⁻² s⁻¹ for 148 angiosperms, 10 for conifers, and 6 for ferns and ginkgos). Along similar lines, the mean ratio $g_{c(op)}/g_{c(max)}$ 149 tends to be conserved across plant functional types; Franks et al. (2014) recommend a value of 0.2, 150 which may correspond to the most efficient setpoint for stomata to control conductance (Franks et al., 151 2012). This conservation of physiological function is one of the underlying principles in the Franks 152 model.

153 Here we test this assumption by estimating CO₂ from 40 phylogenetically diverse species of 154 field-grown trees. In making these estimates, we use the recommended mean values of A_0 and 155 $g_{c(op)}/g_{c(max)}$ from Franks et al. (2014) instead of measuring them directly. Thus, this dataset should be a 156 more faithful gauge for model accuracy as applied to fossils. Of the 40 species, 21 were previously 157 published in Londoño et al. (2018), who collected sun-adapted canopy leaves of angiosperms using a 158 crane in Parque Nacional San Lorenzo, Panama. To test the method in temperate forests, we collected 159 leaves from eleven angiosperm and seven conifer species from Dinosaur State Park (Rocky Hill, Connecticut), Wesleyan University (Middletown, Connecticut), and Connecticut College (New London, 160





161 Connecticut) during the summer of 2015. Here, all trees grew in open, park-like settings; one to three 162 sun leaves were sampled from the lower outside crown of each tree. In January of 2015, we also

- sampled sun-exposed leaves from the tree fern *Cyathea arborea* in El Yunque National Forest, Puerto
- 164 Rico (near the Yokahú Tower).

165 Stomatal size and density were measured either on untreated leaves using epifluorescence 166 microscopy with a 420-490 nm filter, or on cleared leaves (using 50% household bleach or 5% NaOH) 167 using transmitted-light microscopy. For most species, whole-leaf δ^{13} C comes from Royer and Hren 168 (2017); the same leaves were measured for δ^{13} C and stomatal morphology. The UC Davis Stable Isotope 169 Facility measured some additional leaf samples. Table S1 summarizes for these 40 species all of the 170 inputs needed to run the Franks model, along with the estimated CO₂ concentrations. Uncertainties in 171 the estimates are based on error propagation using Monte Carlo simulations (Franks et al., 2014).

172173 2.2 Temperature

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175 The Franks model can be configured for any temperature. Franks et al. (2014) recommend that the photosynthesis parameters A_0 and Γ^* , and the air physical properties affecting diffusion of CO₂ into the 176 177 leaf (the ratio of CO₂ diffusivity in air to the molar volume of air, or d/v) correspond with the mean 178 daytime growing-season leaf temperature (more precisely, assimilation-weighted leaf temperature). The 179 reasoning behind this is that (i) the assimilation-weighted leaf temperature will correspond with the 180 mean c_i/c_a derived from fossil leaf δ^{13} C; and (ii) both theory (Michaletz et al., 2015; Michaletz et al., 181 2016) and observations (Helliker and Richter, 2008; Song et al., 2011) indicate that the control of leaf 182 gas exchange leads to relatively stable assimilation-weighted leaf temperatures (~19-25 °C from 183 temperate to tropical regions) despite large differences in air temperature. This is mostly due to the 184 effects of transpiration on leaf energy balance. Franks et al. (2014) chose a fixed temperature of 25 °C 185 because much of the Mesozoic and Cenozoic correspond to climates warmer than the present-day. 186 When applying the Franks model to known cooler paleoenvironments, improved accuracy may be 187 achieved with leaf-temperature-appropriate values for A_0 , Γ^* , and d/v.

188 Bernacchi et al. (2003) proposed the following temperature sensitivity for Γ^* based on 189 experiments:

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$$\Gamma^* = e^{(19.02 - \frac{37.83}{RT})},$$
 (6)
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where *R* is the molar gas constant $(8.31446 \times 10^{-3} \text{ kJ K}^{-1} \text{ mol}^{-1})$ and *T* is leaf temperature (K). Marrero and Mason (1972) describe the sensitivity of water vapor diffusivity to temperature as:

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$$d = 1.87 \times 10^{-10} \left(\frac{T^{2.072}}{P}\right),$$
 (7)
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where *P* is atmospheric pressure, which we fix at 1 atmosphere. Lastly, the temperature sensitivity of
 the molar volume of air follows ideal gas principles:

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$$v = v_{STP}(\frac{T}{T_{STP}})(\frac{P}{P_{STP}}),$$
(8)

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where T_{STP} is 273.15 K, P_{STP} is 1 atmosphere, and v_{STP} is the air volume at T_{STP} and P_{STP} (0.022414 m³ mol⁻ 1).

Using Eqs. (6-8), we can describe how, conceptually, the sensitivities of Γ^* and d/v to leaf temperature affect estimates of CO₂ from the Franks model. We apply these relationships to a suite of





409 fossil and extant leaves from 62 species of angiosperms, gymnosperms, and ferns. These data come
from the current study (see Sect. 2.1 and 2.4) and Londoño et al. (2018), Kowalczyk et al. (2018), and
Milligan et al. (in review).

210 To experimentally test more generally how the Franks model is influenced by temperature, we 211 grew six species of plants inside two growth chambers with contrasting temperatures (Conviron E7/2; 212 Winnipeg, Canada). Air temperature was 28 °C and 20 °C during the day, and 19 °C and 11 °C during the 213 night. We note that the difference in leaf temperature was probably smaller than that in air 214 temperature during the day (8 °C; see earlier discussion). We held fixed the day length (17 hours with a 215 30 minute simulated dawn and dusk) and CO₂ concentration (500 ppm). Humidity differed moderately 216 between chambers (76.5 \pm 1.8% 1 σ and 90.0 \pm 3.6%). To minimize any chamber effects, we alternated 217 plants between chambers every two weeks.

Four of the species started as saplings purchased from commercial nurseries: bare-root, onefoot tall saplings of *Acer negundo* and *Carpinus caroliniana*, one-foot tall saplings of *Ostrya virginiana* with a soil ball, and bare-root, four-inch tall saplings of *Ilex opaca*. We grew the other two species from seed: *Betula lenta* from a commercial source, and *Quercus rubra* from a single tree on Wesleyan University's campus. All seeds were soaked in water for 24 hours and then cold stratified in a refrigerator for 30 and 60 days, respectively.

224 All seeds and saplings grew in the same potting soil (Promix Bx with Mycorise; Premier 225 Horticulture; Quakertown, Pennsylvania, USA) and fertilizer (Scotts all-purpose flower and vegetable 226 fertilizer; Maryville, Ohio, USA). They were watered to field capacity every other day, and we discarded 227 any excess water passing through the pots. After three months of growth in the chambers, for each 228 species-chamber pair we harvested the three newest fully expanded leaves whose buds developed 229 during the experiment. In most cases, we harvested five plants per species-chamber pair; the one 230 exception was I. opaca, where we were limited to three plants in the warm treatment and two in the 231 cool treatment.

232 We measured stomatal size and density on cleared leaves (using 50% household bleach) with transmitted-light microscopy. Whole-leaf δ^{13} C comes from the UC Davis Stable Isotope Facility and the 233 234 Light Stable Isotope Mass Spec Lab at the University of Florida; the same leaves were measured for δ^{13} C 235 and stomatal morphology. Because we used the same CO₂ gas cylinder as Milligan et al. (in review), we used their two-end-member mixing model to calculate the δ^{13} C of the chamber CO₂ at 500 ppm (-10.6 236 237 ‰). We used the recommended values from Franks et al. (2014) for the physiological inputs A_0 and $g_{c(op)}/g_{c(max)}$. Table S1 summarizes all of the inputs from this experiment needed to run the Franks model, 238 239 along with the estimated CO₂ concentrations. The standard errors for the inputs are based on plant 240 means.

241 To test if leaf δ^{13} C and stomatal morphology (stomatal density, stomatal pore length, and single 242 guard cell width) differed between temperature treatments across species, we implemented a mixed 243 model in R (R Core Team, 2016) using the Ime4 (Bates et al., 2015) and ImerTest (Kuznetsova et al., 244 2017) packages, with temperature and species as the two fixed factors. To test if there was a significant 245 difference between CO₂ estimates from the two temperature treatments, we ran a Kolmogorov– 246 Smirnov (KS) test in R. For each species, we first estimated CO₂ for each plant in the warm and cool 247 treatments based on simulated inputs constrained by their means and variances. In the typical case with 248 five plants per chamber, this produced five CO₂ estimates for the warm chamber and the same for the 249 cool chamber. A KS test was then used to test for a significant temperature effect. We repeated this 250 procedure 10,000 times, with 10,000 associated KS tests. The fraction of tests with a p-value < 0.05 was 251 taken as the overall p value. An advantage of this approach is that it incorporates both within- and 252 across-plant variation.



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255 2.3 Photorespiration

257 c_i/c_a is estimated in the Franks model following Farquhar et al. (1982):

258 259 $\Delta_{leaf} = a + (b - a) \times \frac{c_i}{c_a}$, (9) 260

261 where *a* is the carbon isotope fractionation due to diffusion of CO₂ in air (4.4%; Farquhar et al., 1982), *b* 262 is the fractionation associated with RuBP carboxylase (30%; Roeske and O'Leary, 1984), and Δ_{leaf} is the 263 net fractionation between air and assimilated carbon ([$\delta^{13}C_{air} - \delta^{13}C_{leaf}$]/[1+ $\delta^{13}C_{leaf}$ /1000]).

264 Equation (9) can be expanded to include other effects, including photorespiration (Farquhar et 265 al., 1982):

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$$\Delta_{leaf} = a + (b - a) \times \frac{c_i}{c_a} - \frac{f\Gamma^*}{c_a},$$
 (10)
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269 where *f* is the carbon isotope fractionation due to photorespiration. Photorespiration occurs when the 270 enzyme rubisco fixes O₂, not CO₂ (i.e., RuBP oxygenase). One product of photorespiration is CO₂ (Jones, 271 1992), whose δ^{13} C is lower than the source substrate glycine. If this respired CO₂ escapes to the 272 atmosphere, the δ^{13} C of the leaf carbon becomes more positive. Thus, if *c_i/c_a* is calculated using Eq. (9), 273 as is common practice, the calculation may be falsely low, leading to an underprediction of atmospheric 274 CO₂.

275 Measured values for *f* vary from ~9-15‰ (see compilation in Schubert and Jahren, 2018), which 276 is in line with theoretical predictions (Tcherkez, 2006). At a 400 ppm atmospheric CO₂ and Γ^* of 40 ppm, 277 Eq. (10) implies that ~1‰ of Δ_{leaf} is due to photorespiration, meaning that c_i/c_a should be ~0.04 higher 278 relative to Eq. (9). Here, using the suite of fossil and extant leaves described in Sect. 2.2, we explore how 279 the carbon isotopic fractionation associated with photorespiration affects CO₂ estimates with the Franks 280 model. Because c_i/c_a is present in both of the fundamental equations (Eqs. 2 and 3), we solve them 281 iteratively until c_i/c_a converges.

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283 2.4 Leaves that grow close to the forest floor

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285 The composition of air close to the forest floor can differ considerably from the well-mixed atmosphere. 286 Of relevance to the Franks model, soil respiration can lead to a locally higher CO₂ concentration and lower $\delta^{13}C_{air}$ (Table 1). This effect is strongest at night, when the forest boundary layer is thickest (e.g., 287 288 Munger and Hadley, 2017), but we focus here on daylight hours because that is when most plants take 289 up CO₂. In wet tropical forests, which can have very high soil respiration rates, CO₂ during the day near the forest floor can be elevated by tens-of-ppm, and the $\delta^{13}C_{air}$ can be 2-3‰ lower; in temperate forests, 290 291 the deviations are smaller (Table 1). Above ~2 m, CO₂ concentrations and air δ^{13} C during the daytime 292 largely match the well-mixed atmosphere. 293

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295	Table 1. Deviations in the δ^{13} C and concentration of CO ₂ close to a forest floor relative to well-mixed air
296	above the canopy. All measurements were made close to mid-day.

Study	δ ¹³ C _{air} relative to well-mixed air (‰)	CO ₂ relative to well-mixed air (ppm)	Height above forest floor (m)	Forest location
Tropical forest				
Broadmeadow et al. (1992)	-2	+20	0.15-1	Trinidad during dry season
Buchmann et al. (1997)	-2	+30	0.70-0.75	French Guiana during wet and dry seasons
Holtum and Winter (2001)	NA	+50	0.10	Panama during wet and dry seasons
Lloyd et al. (1996)	-3	+70	1	Brazil (Amazon Basin)
Quay et al. (1989)	-3	+20	2	Brazil (Amazon Basin)
Sternberg et al. (1989)	-2	+25	1	Panama during wet and dry seasons
Temperate forest				
Francey et al. (1985)	-1	+20	1	Tasmania
Munger and Hadley (2017)	NA	+15	1	Massachusetts (Harvard Forest)

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As a result, leaves that grow close to the forest floor may cause the Franks model to produce CO₂ estimates higher than that of the mixed atmosphere for at least two reasons. First, the concentration of CO₂ near the forest floor is elevated; that is, the model may correctly estimate a CO₂ concentration that the user is not interested in. Second, because the $\delta^{13}C_{air}$ that a forest-floor plant experiences is lower than the global well-mixed value, if the user chooses the well-mixed value for model input (inferred, for example, from the $\delta^{13}C$ of marine carbonate; Tipple et al., 2010), c_i/c_a and thus atmospheric CO₂ will be overestimated (see Eq. 2).

We sought to test how the Franks model is affected by the forest-floor microenvironment for 306 307 five tropical angiosperm species and fifteen temperate angiosperm and fern species. The tropical leaves 308 were sampled at ~1-2 m height from Parque Nacional San Lorenzo, Panama. In contrast to the canopy 309 data set from San Lorenzo (Sect. 2.1), these CO₂ estimates have not been previously reported. In the 310 summer of 2015, seven fern species were sampled at ~0.5 m height from Connecticut College and 311 Wesleyan University. Also, we used leaf vouchers from Royer et al. (2010), who sampled eight 312 herbaceous angiosperm species at ~0.1-0.2 m height from Reed Gap, Connecticut. For all 20 species, 313 stomatal and carbon isotopic measurements follow the methods described in Sect. 2.1. Table S1 314 contains all of the inputs needed to run the Franks model, along with the estimated CO₂ concentrations. 315 We also investigated if we could include the forest-floor $\delta^{13}C_{air}$ effect in our estimates of 316 atmospheric CO₂. If the only CO₂ inputs close to the forest floor are from the soil and well-mixed 317 atmosphere, the system can be modeled as a two-endmember mixing model where $\delta^{13}C_{air}$ has a positive, linear relationship with 1/CO2 (Keeling, 1958). If the CO2 concentration and δ^{13} C of both 318

endmembers are known, the forest-floor microenvironment should fall somewhere on the modelled

320 line. Importantly, the Franks model provides a second constraint on the system. Here, $\delta^{13}C_{air}$ has a

negative, nonlinear relationship with $1/CO_2$ because $\delta^{13}C_{air}$ is positively related to c_i/c_a and CO_2 . The Franks model thus provides a second calculation for the relationship between $\delta^{13}C_{air}$ and estimated CO_2





323 concentration. The intersection between the two curves should be the correct $\delta^{13}C_{air}$ and CO_2

324 concentration for the forest-floor microenvironment.

To estimate the soil CO₂ endmember, we measured the δ^{13} C of soil organic matter collected 325 326 from the A horizons of 13 soil sites at San Lorenzo, and of five each at Reed Gap and Connecticut 327 College. For all soils, we assume a 5000 ppm CO₂ concentration for a depth that is below the zone of CO₂ 328 diffusion from the atmosphere (~0.3 m; Cerling, 1999; Breecker et al., 2009). The true value for wet 329 temperate and tropical forest soils may be somewhat less or substantially more than 5000 ppm (Medina 330 et al., 1986; Cerling, 1999; Hirano et al., 2003; Hashimoto et al., 2004; Sotta et al., 2004). Because the 331 mixing model uses 1/CO₂, a much higher CO₂ concentration (e.g., 10000 ppm) has little impact on our 332 results.

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335 3 Results and Discussion

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3.1 General testing in living plants

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Estimates of CO₂ across the 40 tree species sampled in the field range from 275 to 850 ppm, with a mean of 478 ppm and median of 472 ppm (Fig. 2). There are no strong differences across taxonomic orders, nor between leaves from tropical and temperate forests. The mean error rate across the estimates is 28% (median = 24%), which is higher than estimates that include direct measurements of the physiological inputs A_0 and $g_{c(op)}/g_{c(max)}$ (mean = 19%; median = 13%; Fig. 1). Along similar lines, if the estimates presented in Fig. 1 are re-estimated using the values for A_0 and $g_{c(op)}/g_{c(max)}$ recommended by Franks et al. (2014), the mean error rate increases to 31% (median = 21%).

These results indicate that CO₂ accuracy is generally improved when A_0 and/or $g_{c(op)}/g_{c(max)}$ is measured. These measurements require expensive gas-exchange equipment and are not always easy or practical to make. Moreover, A_0 and $g_{c(op)}/g_{c(max)}$ cannot be measured on fossils. Some gains in accuracy are possible by measuring A_0 and $g_{c(op)}/g_{c(max)}$ on extant relatives of the fossil species (e.g., the same genus). Absent of this, our analysis using the recommended mean values of Franks et al. (2014) indicates an error rate, on average, of approximately 28%. This is comparable to or better than other leading paleo-CO₂ proxies (Franks et al., 2014).

One reliable way to improve accuracy is to estimate CO₂ with multiple species because the
 falsely-high and falsely-low estimates partially cancel each other out. The grand mean of estimates
 presented in Fig. 2 (478 ppm) is 20% from the 400 ppm target, which is less than the 28% mean error
 rate of individual estimates.







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Figure 2. Estimates of CO₂ based on canopy leaves from 40 tree species. Uncertainties in the estimates correspond to the 16th-84th percentile range. Vertical line is the correct concentration (400 ppm). On the

362 left is an order-level vascular plant phylogeny (APW v.13; Stevens, 2001 onwards).





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364 365 Dow et al. (2014) observed that $g_{c(op)}/g_{c(max)}$ inversely varies with CO₂ in Arabidopsis thaliana, but 366 primarily at subambient concentrations (yellow triangles in Fig. 3). At elevated CO₂, g_{c(op)}/g_{c(max)} is close 367 to 0.2, which is the value recommended by Franks et al. (2014). Data from eleven species of 368 angiosperms, conifers, and ferns at present-day (or near present-day) and elevated CO₂ concentrations 369 support the view of a limited effect at high CO₂ (Fig. 3; Franks et al., 2014; Londoño et al., 2018; Milligan 370 et al., in review). More data at subambient CO₂ are needed, but for CO₂ concentrations similar to or 371 higher than the present-day, we see no strong reason to include a CO₂ sensitivity in $g_{c(op)}/g_{c(max)}$. The 372 rather low values for Cedrus deodara and many of the tropical angiosperms (<0.1) are likely due to stress imposed by their growth chamber environment; these $g_{c(op)}/g_{c(max)}$ values are probably not 373 374 representative of field-grown trees, which tend to be closer to 0.2 (Franks et al., 2014). 375 376





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382 3.2 Temperature

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The temperature sensitivities of the ratio of diffusivity of CO_2 in air to the molar volume of air (d/v) and the CO_2 compensation point in the absence of dark respiration (Γ^*) have little effect on estimated CO_2 in the Franks model (Fig. 4). Given that assimilation-weighted leaf temperature only varies about 7 °C across plants today, the differences shown in Fig. 4—which are based on leaf temperatures of 25 °C and 15 °C—are likely a maximum effect. As such, we consider the use of a fixed leaf temperature (e.g., 25 °C) in the model to be a defensible simplification.

390







392

Figure 4. Estimates of CO_2 at leaf temperatures of 25 °C and 15 °C. Each symbol is an extant or fossil leaf. The difference in estimated CO_2 for any leaf is due to the theoretical effects of temperature on gas diffusion (d/v) and the CO_2 compensation point in the absence of dark respiration (Γ^*) (Eqs. 6-8).

396 397

398 Other inputs in the model may respond to temperature, though. In our growth chamber 399 experiments where daytime air temperatures were 28 °C and 20 °C, the effect on estimated CO₂ was 400 mixed (Fig. 5). In five out of six species, estimated CO₂ was higher in the warm treatment, but for all 401 species these differences were not statistically significant (P > 0.05 based on a KS test; see Methods). 402 Incorporating the temperature sensitivities in d/v and Γ^* had little effect ("adj" estimates in Fig. 5), as 403 expected from Fig. 4.

404 None of the measured inputs—stomatal density, stomatal pore length, single guard cell width, 405 and leaf δ^{13} C—were significantly affected by temperature across all species (P > 0.05 for each of the four 406 inputs based on a mixed model; see Methods). These small differences probably cannot account for the 407 differences in estimated CO_2 between temperatures. It is more likely that some of the inputs that we did 408 not directly measure, such as assimilation rate (A_0), the $g_{c(op)}/g_{c(max)}$ ratio, or mesophyll conductance (g_m), 409 differ from the true mean value. In the cases for the five species where estimated CO₂ is higher in the warm treatment, our mean A_0 for the warm plants must be falsely high, or $g_{c(op)}/g_{c(max)}$ or g_m falsely low. 410 411 In summary, we see no strong reason to expand the parameterization of temperature in the 412 model, though more growth-chamber experiments may be warranted. We note that the across-species

413 means of estimated CO_2 for the warm and cool treatments are reasonably close to the 500 ppm target 414 (590 and 502 ppm, respectively) and overall have a mean error rate of 25%. This level of uncertainty is 415 similar to our field estimates where we did not measure A_0 or $g_{c(op)}/g_{c(max)}$ (28%; see Fig. 2). This too

- 416 provides support for our recommendation that it is not critical to include a broader treatment of
- 417 temperature in the model.





418 419



420

421Figure 5. Estimates of CO_2 for plants grown inside growth chambers at daytime air temperatures of 28 °C422and 20 °C. Also shown are estimates after taking into account the temperature sensitivity of gas423diffusion (d/v) and the CO_2 compensation point in the absence of dark respiration (Γ^*) ("adj"; see also424Fig. 4). Dashed line is the correct CO_2 concentration (500 ppm). Uncertainties in the estimates425correspond to the 16^{th} -84th percentile range.

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- 427

428 3.3 Photorespiration

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430 The theoretical effects of photorespiration do not strongly impact estimates of CO_2 in the Franks model. 431 The average effect for our 409 extant and fossil leaves is to increase estimated CO₂ by 2.2% plus 38 ppm 432 (Fig. 6). At 1000 ppm, for example, estimates would increase by 60 ppm. This calculation assumes a 433 photorespiration fractionation (f) of 9.1‰, which is the value estimated for Arabidopsis thaliana 434 (Schubert and Jahren, 2018). If a fractionation towards the upper bound of published estimates is used 435 instead (15‰), estimated CO_2 increases on average by 3.8% plus 61 ppm. Across this range in f, the 436 associated uncertainty in estimated CO₂ is well within the method's overall precision (~+35/-25% at 95% 437 confidence; Franks et al., 2014). As such, CO₂ estimates made without these photorespiration effects 438 (i.e. using Eq. 9 instead of Eq. 10) should be reliable, although some improvement is possible using Eq. 439 10 in cases where *f* is accurately known. 440







Figure 6. Estimates of CO₂ with and without a photorespiration effect (f = 9.1%; see Eq. 10). Each symbol is an extant or fossil leaf. Dashed line is y=x.

445

446

447 We note that both f and Γ^* are also affected by atmospheric O₂ concentration. Because O₂ is 448 directly responsible for photorespiration, f should scale with O₂ (or, more precisely, the O₂:CO₂ molar 449 ratio). Unfortunately, this effect is poorly constrained (Beerling et al., 2002; Berner et al., 2003; Porter et 450 al., 2017). In contrast, the theoretical effect of O_2 on Γ^* is known: it is linear with a slope of 0.5 (Farquhar 451 et al., 1982; see their Eq. B13). During the Phanerozoic, O₂ likely ranged from 10-30%, with lows during 452 the early Paleozoic and early Triassic, and highs during the Carboniferous to early Permian and 453 Cretaceous (Berner, 2009; Glasspool and Scott, 2010; Arvidson et al., 2013; Mills et al., 2016; Lenton et 454 al., 2018). Assuming a present-day Γ^* of 40 ppm (at 21% O₂), Γ^* would be 49 ppm at 30% O₂ and 29 ppm 455 at 10% O₂. Running the Franks model on our library of 409 extant and fossil leaves, we find little effect 456 on estimated CO₂: estimates are 3.3% higher on average at 30% O₂ than at 10% O₂.

457

458 3.4 Leaves that grow close to the forest floor

459

460 CO₂ estimates for tropical understory leaves from five species at San Lorenzo, Panama, are very high, 461 ranging from 1903 to 18863 ppm (species mean = 6837 ppm). For two of the species Londoño et al. 462 (2018) also analyzed canopy leaves from trees nearby, and these within-species comparisons highlight 463 the vast discrepancy (*Ocotea* sp.: 541 vs. 5737 ppm; *Tontelea* sp.: 622 vs. 18863 ppm). The primary 464 difference in the inputs between the canopy and understory leaves is the $\delta^{13}C_{leaf}$: Londoño et al. (2018) 465 report a species-mean $\delta^{13}C_{leaf}$ of -30.0% for the 21 canopy species versus -35.6% for the five understory

466 species. This difference leads to very different mean estimates of c_i/c_a : 0.69 for canopy leaves versus a

467 highly unrealistic (Diefendorf et al., 2010) 0.93 for understory leaves.





468 It is likely that the high CO₂ estimates from understory leaves are mostly driven by falsely high 469 $\delta^{13}C_{air}$ inputs. Following the mixing model strategy outlined in Sect. 2.4 (and based on a soil organic 470 matter $\delta^{13}C$ of -28.2% measured at San Lorenzo), we calculate a species-mean $\delta^{13}C_{air}$ of -16.7% (mean 471 of intersection points in Fig. 7). When this $\delta^{13}C_{air}$ is used to estimate CO₂ with the Franks model (instead 472 of -8.5%), the species mean drops to 699 ppm. This is somewhat higher than the species mean from 473 canopy leaves in the same forest (563 ppm; red triangles in Fig. 2; Londoño et al., 2018).

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476

477Figure 7. Sensitivity of estimated CO2 in the Franks model to the δ^{13} C of atmospheric CO2. Estimates478come from leaves of five angiosperm species that grew close to the forest floor in Parque Nacional San479Lorenzo, Panama. The step in δ^{13} Cair between estimates is 0.5‰. The dashed line is a two-endmember480mixing model for CO2 between the soil and well-mixed atmosphere. The intersection between the481mixing model and the Franks model should correspond to the CO2 concentration and δ^{13} Cair of the482forest-floor microenvironment.

483 484

485 Understory leaves from Connecticut temperate forests show similar but less dramatic patterns, 486 which we attribute to a more open canopy with stronger atmospheric mixing. CO_2 estimates for the 15 487 species range from 447 to 1567 ppm (mean = 794 ppm). Our intersection method identifies a mean 488 $\delta^{13}C_{air}$ of -11.2% for the Wesleyan and Connecticut College campuses (based on a soil $\delta^{13}C$ of -27.6% 489 measured at Connecticut College) and -10.3% for Reed Gap (soil $\delta^{13}C = -26.4\%$). Using these adjusted





490 $\delta^{13}C_{air}$, the species mean of estimated CO₂ drops to 566 ppm, which is somewhat higher than the species 491 mean from canopy leaves in the same areas (449 ppm; blue circles in Fig. 2).

492 We acknowledge that this analysis is too simple: other factors probably contribute to the 493 differences in estimated CO₂ between canopy and understory leaves. Our predicted $\delta^{13}C_{air}$ values are too 494 low (~8‰ and 2‰ lower than the well-mixed atmosphere for the tropical and temperate forests) and 495 our estimated CO₂ too high (~100 ppm higher than that from canopy leaves). In the lowermost 1-2 496 meters of the canopy, previous work suggests up to a -3‰ and +70 ppm deviation in tropical forests and 497 -1‰ / +20 ppm in temperate forests (Table 1). One input that could help to resolve this discrepancy is 498 the assimilation rate (A_0). We assumed a fixed A_0 of 12 µmol m⁻² s⁻¹ for all leaves, regardless of canopy 499 position. Shade leaves often have lower assimilation rates than sun leaves (Givnish, 1988). Substituting 500 lower A₀ values for understory leaves would lower estimated CO₂ roughly in proportion (Eqs. 2-3). Using 501 lower A_0 values for shade leaves in the model is appropriate, but determining the best value is difficult. 502 Typical A_0 values for leaves growing at the top of the canopy in full sun are far more consistent because 503 photosynthesis in these leaves is usually at its maximum capacity (saturated at full sunlight) for the 504 prevailing atmospheric CO₂ concentration. Because the degree of shadiness near the forest floor is 505 highly variable, photosynthesis (A_0) in these leaves will be acclimated to some fraction of the full-sun 506 maximum in a sun exposed leaf, but careful thought must go into determining what this fraction is.

507 We note that our mixing-model strategy cannot be applied to fossils because the global 508 atmospheric CO₂ concentration is needed (one endpoint for dashed line in Fig. 7). Instead, our 509 motivation for the analysis is to demonstrate that: 1) leaves growing in the lowermost 2 m of the canopy 510 should be considered with caution in the context of the Franks model; and 2) the failure of the model is 511 due to faulty inputs (mostly $\delta^{13}C_{air}$), not the model itself.

In most fossil leaf deposits, shade morphotypes are comparatively rare (e.g., Kürschner, 1997; 512 513 Wang et al., in press) because—relative to sun leaves—they are not as tough, do not travel as far by 514 wind, and are produced at a slower rate (Dilcher, 1973; Roth and Dilcher, 1978; Spicer, 1980; Ferguson, 515 1985; Burnham et al., 1992). Our recommendation is to exclude such leaves. There are several ways to 516 differentiate sun vs. shade morphotypes: overall shape (Talbert and Holch, 1957; Givnish, 1978; 517 Kürschner, 1997; Sack et al., 2006), shape of epidermal cells (larger and with a more undulated outline in 518 shade leaves; Kürschner, 1997; Dunn et al., 2015), vein density (lower in shade leaves; Uhl and 519 Mosbrugger, 1999; Sack and Scoffoni, 2013; Crifò et al., 2014; Londoño et al., 2018), and range in $\delta^{13}C_{leaf}$ 520 (high when both sun and shade leaves are present, for example in our study; Graham et al., 2014). Not 521 all shade leaves grow within 2 m of the forest floor, but excluding all such leaves would eliminate the 522 forest-floor bias.

523 524

525 4 Conclusions

526

527 The Franks model is reasonably accurate (~28% error rate) even when the physiological inputs A_0 528 (assimilation rate at a known CO₂ concentration) and $g_{c(op)}/g_{c(max)}$ (ratio of operational to maximum leaf 529 conductance to CO₂) are inferred, not measured. Accuracy does improve when these inputs are 530 measured (~19% error rate), but such measurements are not possible with fossils and may not always 531 be feasible with nearest living relatives. A 28% error rate is broadly in line with (or better than) other 532 leading paleo-CO₂ proxies.

533 Most of the possible confounding factors that we investigated appear minor. The temperature 534 sensitivities of d/v (related to gas diffusion) and Γ^* (CO₂ compensation point in the absence of dark 535 respiration) have a negligible impact on estimated CO₂. Our temperature experiments in growth 536 chambers point to larger differences in some species, which must be related to incorrect values for 537 inputs that were not directly measured, such as A_0 , $g_{c(pp)}/g_{c(max)}$, and g_m (mesophyll conductance).





Overall, though, we find that the differences in estimated CO₂ imparted by temperature are generally
 smaller than the overall 28% error rate.

540Incorporating the covariance between CO_2 concentration and photorespiration leads to only541small changes in estimated CO_2 . O_2 concentration affects photorespiration and thus may confound CO_2 542estimates from the Franks model, but presently the effect is poorly quantified. The effect of O_2 on Γ^* is543better known, and imparts only small changes in estimated CO_2 across a feasible range in Phanerozoic544 O_2 of 10-30%.

545 Leaves from the lowermost 1-2 m of the canopy experience slightly elevated CO₂ concentrations 546 and lower air δ^{13} C during the daytime relative to the well-mixed atmosphere. We find that if we use the 547 well-mixed air δ^{13} C to estimate CO₂ from leaves that grew near the forest floor, estimates are too high, 548 especially in dense tropical canopies. When we use a two-endmember mixing model to calculate the 549 correct local air δ^{13} C, the falsely-high CO₂ estimates largely disappear. For fossil applications, shade 550 leaves from the bottom of the canopy should be avoided. Shade leaves are typically rare in the fossil 551 record (relative to sun leaves), and can be identified by their overall shape, the shape of their epidermal cells, their low leaf δ^{13} C, and their low vein density. 552

553 Conceptually, the Franks model holds considerable promise for quantifying paleo-CO₂: it is 554 mechanistically grounded and can be applied to most fossil leaves. Our tests of the model's accuracy 555 and sensitivity to temperature and photorespiration largely uphold this promise.

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557

Author contribution. DR, KM, MM, and LL designed and conducted the experiments; all authors
 interpreted the data; DR prepared the manuscript with contributions from all co-authors.

560

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

562

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570 571 **References**

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