



1 Using Respiration Quotients to Track Changing Sources of Soil 2 Respiration Seasonally and with Experimental Warming

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11 **Abstract.** Developing a more mechanistic understanding of soil respiration is hampered by the difficulty in
12 determining the contribution of different organic substrates to respiration and in disentangling autotrophic versus
13 heterotrophic and aerobic versus anaerobic processes. Here, we present a relatively novel tool for better understanding
14 soil respiration: the apparent respiration quotient (ARQ). ARQ is the amount of CO₂ produced in the soil divided by
15 the amount of O₂ consumed and it changes according to which organic substrates are being consumed and whether
16 oxygen is being used as an electron acceptor. We investigated how the ARQ of soil gas varied seasonally, by soil
17 depth, and by experimental warming *in situ* in a coniferous forest whole-soil-profile warming experiment over two
18 years. We then compared the patterns in ARQ to those of soil δ¹³CO₂. Our measurements showed strong seasonal
19 variations in ARQ from ≈0.9 during the late spring and summer to ≈0.7 during the winter. This pattern likely reflected
20 a shift from respiration being fueled by oxidized substrates like sugars and organic acids derived from root and root
21 respiration during the growing season to more reduced substrates such as lipids and proteins derived from microbial
22 necromass during the winter. This interpretation was supported by δ¹³CO₂ values, which were relatively depleted, like
23 lipids, in the winter and more enriched, like sugars, in the summer. Furthermore, wintertime ARQ was higher in
24 warmed (+4°C) than in control plots, probably due to an increase in the use of more oxidized carbon substrates with
25 warming. Our results demonstrate that soil ARQ shows strong seasonal patterns in line the phenology of carbon inputs
26 and patterns in soil δ¹³CO₂, verifying ARQ as a tool for disentangling the biological sources contributing to soil
27 respiration.



28 **1 Introduction**

29 Despite making extensive measurements of soil respiration (Bond-Lamberty and Thomson, 2010), scientists lack methods to
30 disentangle the processes underlying, and substrates contributing to, soil respiration, which hampers predictions of terrestrial
31 carbon cycle responses to global change (Phillips et al., 2017). Mechanistic uncertainty surrounding soil respiration is partly
32 responsible for the 1000 Pg spread in model predictions of end-of-century terrestrial carbon-climate feedbacks (Friedlingstein
33 et al., 2013). Soil respiration is the sum of autotrophic respiration by plant roots and heterotrophic respiration by soil microbes.
34 Heterotrophic respiration, which has increased globally over the past three decades (Bond-Lamberty et al., 2018), is itself the
35 sum of various processes using different sources of energy. For example, microbes consume different organic substrates
36 depending on what molecules are accessible and whether the microbes are living in the rhizosphere or bulk soil, and microbes
37 utilize different terminal electron acceptors depending on O₂ availability in the microsites in which they reside (Keiluweit et
38 al., 2016; Liptzin et al., 2011). The electron donors (the organic substrates) and the electron acceptors used by soil microbes
39 during respiration cannot be resolved by measuring the CO₂ flux alone. Previous studies have used measurements of δ¹³C to
40 partition respiration into autotrophic and heterotrophic components (e.g., Dorrepaal et al., 2009), radiocarbon to partition
41 respiration sources by age (e.g., Trumbore, 2000), or both isotopes in combination to more finely separate respiration among
42 sources (e.g., Hicks Pries et al., 2013; Hopkins et al., 2012). However, isotopes are not the only way to disentangle soil
43 respiration's various components.

44

45 Our ability to understand soil respiration is limited by measuring only one half of the respiration equation, the CO₂ produced.
46 Simultaneously measuring the O₂ consumed can provide a more mechanistic understanding of the processes and substrates
47 contributing to soil respiration (Phillips et al., 2017). The paired measurements of CO₂ and O₂ can be used to calculate a
48 respiration quotient (ARQ; Angert and Sherer, 2011). All organic matter has an oxidative ratio (OR=1/RQ), which can be
49 calculated based on an elemental analysis of its C, H, O, and N (Masiello et al., 2008). The oxidation state of carbon in
50 carbohydrates is 0 with a corresponding RQ of 1. More reduced energy sources such as lipids have lower RQ values (≈0.73)
51 and the RQ of proteins range from 0.67 to 1; more oxidized sources such as organic acids have RQ ranges from 1 to 4 (Masiello
52 et al., 2008; Table 1). The RQ of aerobic respiration therefore changes based on what substrates are being consumed (Dilly,
53 2001; Theenhaus et al., 1997). Anaerobic respiration increases RQ to values greater than one, as electron acceptors like Fe(III)
54 and NO₃⁻ replace O₂. Thus, RQ can help differentiate between the electron donors (organic substrates) and terminal electron
55 acceptors used during soil respiration. We will refer to 'apparent' RQ because not all ecosystem CO₂ or O₂ fluxes are due to
56 respiratory processes (Angert and Sherer, 2011). For example, fluctuating redox conditions can lead to consumption of O₂
57 during metal oxidation and drive ARQ below the value of the most reduced organic matter (Angert et al., 2015).

58



59 Thus far, CO₂:O₂ ratios have been primarily used to understand large scale earth system processes and only few studies have
60 examined processes within ecosystems. This ratio in atmospheric samples has been used to estimate a) the magnitude of the
61 terrestrial carbon sink, because carbon uptake by terrestrial ecosystems is balanced by O₂ production whereas ocean CO₂
62 uptake is decoupled from O₂, (Keeling, 1988; Keeling et al., 1996; Randerson et al., 2006; Worrall et al., 2013) and b)
63 anthropogenic impacts on the carbon cycle, based on the principle that burning of reduced fossil fuels results in a different
64 oxidative ratio than does photosynthesis and subsequent respiration of carbohydrates (Keeling, 1988). The CO₂:O₂ ratio of
65 ecosystem-atmosphere exchanges is an essential quantity in these carbon cycle calculations. CO₂:O₂ ratios have been estimated
66 from measurements of net ecosystem exchange of CO₂ and O₂ (e.g., Seibt et al., 2004) and from elemental analysis of biomass
67 (Hockaday William C. et al., 2015; e.g., Masiello et al., 2008), both of which are assumed to be similar over multiyear
68 timescales. In early carbon sink calculations, the oxidative ratio of ecosystem fluxes was assumed to be 1.1 (ARQ=0.9) based
69 on a single study of temperate soils (Severinghaus, 1995). However, the few subsequent studies examining the CO₂:O₂ ratio
70 of soil respiration fluxes have shown soil fluxes can deviate widely from that value.

71

72 Soil ARQ from incubations shift as a result of temperature changes, substrate additions, and soil management. For example,
73 the ARQ of peat soils decreased from about 1.1 to about 0.6 when temperatures increased from 0 to 20°C, attributed to changing
74 substrate use (Chapman and Thurlow, 1998). Glucose additions to German forest soils increased soil ARQ to 0.95-1.0 from a
75 basal value around 0.7 (Dilly, 2001; Theenhaus et al., 1997). Soils under organic agriculture were found to have a greater ARQ
76 (1.19) than soils under conventional agriculture (0.72; Theenhaus et al., 1997). Soil ARQ in mesocosms containing pine
77 seedlings changed seasonally and when the pine seedlings were cut, indicating the ratio is responsive to changes in vegetation
78 (Andersen and Scagel, 1997; Scagel and Andersen, 1997). Lastly, in one of the only studies using *in situ* measurements, soil
79 ARQ taken from gas wells across multiple forested ecosystems ranged widely from 0.14 to 1.23 indicating the influence of
80 abiotic processes that consume O₂ (Angert et al. 2015). The wide range in soil ARQ values associated with different
81 biochemical conditions indicates the ratio has the potential to provide insight into the substrates contributing to respiration as
82 well as into abiotic O₂ consumption. Finer scale research is needed, however, to explore ARQ values in the same soils under
83 different conditions to learn what these values indicate about the processes and substrates contributing to soil respired CO₂.

84 Here we investigated how the ARQ of soil gas *in situ* varied seasonally, by soil depth, and by experimental warming in a
85 whole-soil-profile warming experiment in a well-drained, oxygenated coniferous forest soil (Hicks Pries et al., 2017). We
86 characterized soil ARQ at 30 and 90 cm depths in the winter and growing season over two years and compared the patterns in
87 ARQ to monthly patterns in soil profile δ¹³C. We hypothesized that ARQ values would change seasonally and with warming
88 reflecting the values of the organic carbon substrates being consumed by microbes. Like ARQ, the δ¹³C of soil CO₂ is
89 influenced by the use of different organic substrates since more reduced substrates tend to also be depleted in δ¹³C (Bowling
90 et al., 2008). By comparing ARQ values to other indicators of respiration sources, such as δ¹³C, augmented by what we
91 understand about plant allocation of carbon substrates belowground, we aim to advance the utility of ARQ as a tracer of
92 respiration processes.



93 2 Methods

94 2.1 Warming Experiment

95 The whole soil profile warming experiment is located at the University of California Blodgett Forest Research Station, in the
96 Sierra Nevada foothills near Georgetown, CA at 1370 m above sea level. Mean annual precipitation is 1774 mm with most of
97 it occurring from November through April and mean annual temperature is about 12.5°C (Bird and Torn, 2006). The
98 experiment is in a thinned 80-year-old stand of mixed conifers including ponderosa pine (*Pinus ponderosa*), sugar pine (*Pinus*
99 *lambertiana*), incense cedar (*Calocedrus decurrens*), white fir (*Abies concolor*), and douglas fir (*Pseudotsuga menziesii*). The
100 soils are Holland series: fine-loamy, mixed, superactive, mesic Ultic Haploxeralfs of granitic origin with thick, >5 cm O
101 horizons, minimal carbonates (Rasmussen et al., 2005), and a pH that ranges from 5.6 to 6.5 (Hicks Pries et al., 2018). The
102 warming treatment warmed the soil +4°C to 1 m depth while maintaining the natural temperature gradient with depth and
103 temporal variations in soil temperature as described in Hicks Pries et al. (2017). Briefly, there were three pairs of control and
104 heated 3 m diameter circular plots heated by 22 vertical resistance heater cables in metal conduit (BriskHeat, Ohio, USA) that
105 surrounded them. To compensate for surface heat loss, two concentric rings of heater cable at 1 and 2 m in diameter were
106 installed 5 cm below the soil surface in heated plots. Unheated cables were installed similarly in control plots. Heating
107 throughout the plot volume was generally even, ranging from 3.5 to 4.5°C except at 5 cm depth where the heated plots were
108 on average only $2.4 \pm 1.2^\circ\text{C}$ warmer than the control due to a lack of aboveground heating. Soil moisture was slightly decreased
109 in the warmed plots by an average of 1.5-3.5% volumetric water content (Hicks Pries et al., 2017).

110 2.2 Sample Collection and Analysis

111 Dataloggers (CR1000, Campbell Scientific, Utah, USA) continuously recorded soil temperature and moisture at 30 min
112 intervals. Temperature was monitored at 5, 15, 30, 50, 75, and 100 cm depths at a radial distance of 0.75 m from the center of
113 each plot. Temperature probes consisted of thermistors (Omega 44005) epoxied to PVC rods, placed inside thin-walled steel
114 conduit. To monitor soil moisture, we used an enviroSCAN (Sentek, Australia) probe fitted with capacitance sensors at 10, 30,
115 50, and 90 cm at a radial distance of 0.75 m from the center of each plot. We calibrated the soil moisture measurements by
116 comparing the sensor values at each depth to the volumetric water content measured in nearby (within 0.5 m) soil cores that
117 were sampled five times over two years.

118

119 Each of the six plots has a set of gas wells at 15, 30, 50, 75, and 90 cm. The gas wells were 6.35 mm diameter stainless steel
120 tubes inserted into the soil at a 45° angle to the desired depth and topped with straight swage pipefittings (Swagelok Ohio,
121 USA) with septa. For CO₂ and δ¹³CO₂ measurements, samples were collected from the wells with a syringe on a nearly monthly
122 basis from March 2014 through June 2017 (32 months total) and always during morning hours. After clearing the headspace
123 in each well, a 25 ml gas sample was transferred to an evacuated 20 ml septum-topped glass vial. For analysis, 5 ml samples
124 were injected into the small sample isotope module of a cavity ring down spectrometer (CRDS, Picarro, Santa Clara,



125 California) where they were diluted with ultra zero air (without CO₂). A four-point calibration curve ranging from 2,000 to
126 20,000 ppm ($\delta^{13}\text{C}=-26.7\text{‰}$) was used to calculate the CO₂ concentration from the CRDS data and to correct for mass
127 dependency of the $\delta^{13}\text{C}$ measurement.

128

129 In July 2015, February 2016, April 2016, August 2016, March 2017, and June 2017, we collected additional samples from the
130 30 (except July 2015) and 90 cm gas wells into 13 ml flasks equipped with O-ring valves (LouwersHanique, Hapert,
131 Netherlands) to simultaneously measure CO₂ and O₂ concentrations in order to calculate ARQ. The flasks were analyzed in
132 the laboratory at the Hebrew University by a closed system (The Hampadah; Hilman and Angert, 2016). This fully automated
133 system uses an infra-red gas analyzer (IRGA) for CO₂ measurement (LI 840A LI-COR; Lincoln, NE, USA) and a fuel-cell
134 based analyzer (FC-10; Sable Systems International, Las Vegas, NV, USA) for measuring O₂. The flasks were analyzed within
135 2-3 weeks of collection.

136

137 In June 2017, we also ran a set of short (3 hour) incubations of root-free soil and of excised roots collected adjacent to the
138 experimental plots. We collected four mineral soil cores with a 5 cm diameter hammer corer, separated the cores into 0-20 and
139 20-40 cm depths, and removed roots >1 mm diameter. Roots were collected from four 25 cm x 25 cm x 25 cm soil pits. We
140 rinsed roots with water to remove soil and blotted them dry before placing them into mason jars. The root-free soil was also
141 placed into mason jars, and both sets of mason jars were flushed with ambient, outside air. After a three-hour incubation of the
142 root samples and a 21-hour incubation of the soil samples, the headspace was sampled for CO₂ and O₂ and analyzed as
143 described above. Incubations were run at room temperature, which was similar to the field temperature at the time of collection.

144 2.3 Sample Calculations and Statistics

145 To calculate ARQ, we used the following equation from Angert et al. (2015):

146
$$ARQ = -0.76 \frac{\Delta CO_2}{\Delta O_2}$$

147 Where ARQ is the apparent respiratory quotient, ΔCO_2 (ppmv) is the difference between CO₂ concentrations in the soil pore
148 space gas and ambient (i.e., near-surface aboveground) samples, ΔO_2 (ppmv) is the difference of the soil pore space O₂
149 concentration and ambient O₂ concentration, and 0.76 is the ratio of CO₂ to O₂ diffusivity in air (Massman, 1998). The negative
150 sign is for convenience so the ARQ value will typically be positive, because the difference in O₂ concentration is always
151 negative. For the jar incubations we used the same equation without the 0.76 factor. Ambient CO₂ concentrations were
152 measured in the field at the time of sampling with the CRDS, while the ambient O₂ concentration was assumed to be 20.95%.
153 To relate the $\delta^{13}\text{C}$ value of soil pore space CO₂ to the $\delta^{13}\text{C}$ of CO₂ production, we corrected the pore-space $\delta^{13}\text{C}$ value for
154 diffusion since ¹³C diffuses slower in air than ¹²C and thus the measured value does not accurately represent the value of
155 production. For the correction, we used the following equation from Bowling et al. (2015):



156
$$\delta_{production} = \frac{C_s(\delta_s - 4.4) - C_a(\delta_a - 4.4)}{1.0044(C_s - C_a)}$$

157 Where C_s is the soil pore space CO₂ concentration (ppmv), δ_s (‰) is the isotopic composition of soil pore space CO₂ and C_a
158 and δ_a are the CO₂ concentration and isotopic composition of ambient air, respectively. The ambient CO₂ concentrations and
159 $\delta^{13}\text{C}$ values needed for these corrections were measured in the field at the time of sampling with the CRDS.

160

161 To investigate the effects of season, warming treatment, and soil depth on ARQ and $\delta^{13}\text{C}$, we ran multiple regressions in R (R
162 Development Core Team, 2017). Because ARQ was not sampled from both depths on all dates, we ran separate regressions
163 for each depth (30 and 90 cm) and then ran a regression that included a depth effect while dropping the first sampling date. In
164 all regressions, treatment and sampling date (as a factor) were fixed effects. Following Zuur et al. (2009), we used a full model
165 with all fixed effects and their interactions to optimize the random effects and autocorrelation structure based on AIC. For both
166 versions, we used the individual gas well as a random effect and a temporal autocorrelation did not improve the model, nor
167 did an autocorrelation function graph indicate one was needed. We chose the significant fixed effects by performing a series
168 of pairwise model comparisons using AIC and the F test, dropping the least significant variables each time until only variables
169 that improved the model fit remained. The p-values reported are those from the t-tests of the summary.lme function of best fit
170 model. To investigate seasonal patterns in $\delta^{13}\text{CO}_2$, we had more data in terms of both length of time and temporal density of
171 sampling and were thus able to treat month as a continuous variable. We fit a sine function and tested models including the
172 first and second harmonics of the month effect as well as linear fixed effects of depth, treatment, and a depth by treatment
173 interaction. Graphical exploration indicated the sinusoidal pattern differed slightly by year, so we also added a year effect to
174 the second harmonic of the month effect. As above, we used the full fixed effect model to test the best random and
175 autocorrelation structure. Individual gas well depth was used as a random effect and an autocorrelation-moving average
176 (corARMA, p=2, q=2) was the best correlation structure. To test relationships between ARQ and $\delta^{13}\text{CO}_2$, and both ARQ and
177 $\delta^{13}\text{CO}_2$ individually versus soil temperature and volumetric water content, we ran mixed-model regressions with individual
178 gas well as a random effect. We used sensor data from the depths sampled by gas wells, which limited our analyses to the 30
179 and 90 cm sensor depths only. We tested the need for autocorrelation structures based off of AIC and none improved the
180 model. For these regressions, we report pseudo R^2 values calculated from a linear regression of the actual data versus the model
181 predicted data. For all models, we graphically checked the residuals for violations of normality and heterogeneity of variance.
182 For $\delta^{13}\text{CO}_2$ analyses, we dropped the 15 cm depths due to their unusually low $\delta^{13}\text{C}$ value (<-32‰) after correction (Eq. 1),
183 which indicated potential intrusion of atmospheric air during sampling that led to an overcorrection. We used one-way
184 ANOVA's to compare the ARQ of soil and root incubations and the ARQ of two soil depths we incubated. All statistics were
185 performed in R v 3.4.1 and regressions were done using the lme function (R Development Core Team, 2017).



186 3 Results

187 Both ARQ and $\delta^{13}\text{CO}_2$ had similar, strong seasonal patterns (Fig. 1a and 1b). ARQ values were higher during the growing
188 season (0.89 ± 0.01) and lower during the winter (0.70 ± 0.02). In ARQ regression analyses for both depths, there was a
189 significant effect of date ($p < 0.0001$) with February 2016 and March 2017 differing significantly from July 2015 (90 cm only),
190 April 2016, August 2016, and June 2017. Similarly, $\delta^{13}\text{C}$ was more enriched during the summer (June through October, -28.43
191 ± 0.08) and more depleted during the winter and spring (November through May, -29.26 ± 0.05). While individual dates were
192 not compared statistically for $\delta^{13}\text{CO}_2$, the vast improvement in model fit using month as a sine function instead of a linear
193 function ($\Delta\text{AIC}=112$) is strong statistical evidence for a seasonal effect (Fig. 2b). ARQ and $\delta^{13}\text{CO}_2$ were significantly related
194 according to the mixed effect regression model (Fig. 1c, $p < 0.0001$, $n=64$, pseudo $R^2=0.20$). However, the patterns in ARQ and
195 $\delta^{13}\text{CO}_2$ did not match during April.

196

197 Both ARQ and $\delta^{13}\text{CO}_2$ differed by warming treatment (Fig. 2) and by depth (Table 1), although to a lesser magnitude than by
198 seasonal differences. For 30 cm depths, there was a significant treatment-by-date interaction ($p=0.051$, $n=30$) whereby heated
199 plots had greater ARQ values during the winter months (February 2016 and March 2017; Fig. 2a). In contrast, the best fit
200 model for 90 cm did not include a significant treatment effect or treatment-by-date interaction (Fig. 2a). For $\delta^{13}\text{CO}_2$, treatment
201 was a significant effect ($p=0.0065$, $n=758$) with warmed soil on average having a slightly more enriched $\delta^{13}\text{CO}_2$ ($-28.33 \pm$
202 0.05) than the control soil (-28.83 ± 0.06 ; Fig 2b). The treatment-by-depth interaction was not significant for $\delta^{13}\text{CO}_2$ and was
203 not included in the best fit model. Looking at depth only, ARQ at 30 cm was significantly greater than ARQ at 90 cm by 0.03
204 units ($p=0.0495$, $n=59$), while $\delta^{13}\text{CO}_2$ became slightly more enriched with depth going from -28.98 at 30 cm to -28.34 at 90
205 cm ($p=0.0089$, $n=758$).

206

207 Both ARQ and $\delta^{13}\text{CO}_2$ showed strong relationships with soil climate (Fig 3). We tested relationships with soil temperature and
208 soil moisture individually because of the strong negative correlation between temperature and moisture in this Mediterranean
209 climate (pearson's $r=-0.76$ to -0.78). ARQ increased significantly with increasing soil temperatures ($p < 0.0001$, $n=64$, pseudo
210 $R^2=0.46$; Fig 3a). It decreased with increased soil moisture, but the fit ($p < 0.0001$, $n=59$ due to missing VWC values, pseudo
211 $R^2=0.18$) was not as good as for temperature (Fig. 3b). $\delta^{13}\text{CO}_2$ became more enriched with increasing soil temperatures
212 ($p < 0.0001$, $n=375$, pseudo $R^2=0.33$; Fig. 3c) and more depleted with increased soil moisture ($p < 0.0001$, $n=345$ due to missing
213 VWC values, pseudo $R^2=0.30$; Fig. 3d).

214

215 Our incubations of roots and of root-free soil indicated that heterotrophic and autotrophic respiration has significantly different
216 ARQ values, at least during the summer when we performed the incubations. Roots had a greater ARQ (0.87 ± 0.03) than did
217 root-free soil (0.78 ± 0.02 ; one-way ANOVA, $p=0.029$). Furthermore, ARQ of the soil incubations significantly declined with
218 depth from 0.82 ± 0.01 at 0-20 cm to 0.74 ± 0.02 at 20-40 cm (one-way ANOVA, $p=0.0053$).



219 4 Discussion

220 There are many factors that can affect ARQ; however, our evidence indicates the strong seasonal patterns in ARQ and $\delta^{13}\text{CO}_2$
221 were likely driven by changes in the amount of root-derived organic substrates providing energy for heterotrophic microbial
222 respiration and changes in the contributions of autotrophic root respiration. This interpretation is supported by previous soil
223 ARQ studies, our incubations, and the scientific understanding of how plant carbon inputs change seasonally. The seasonal
224 range in ARQ from ≈ 0.9 during the growing season to ≈ 0.7 during the winter may reflect a shift in the molecules fueling
225 respiration from more oxidized substrates like sugars and organic acids derived from roots in the summer to more reduced
226 substrates in the winter such as lipids and proteins derived from microbial necromass. Previous incubations found that glucose
227 additions increased ARQ (Dilly, 2001; Theenhaus et al., 1997). Other studies attributed a decline in ARQ during the time
228 course of incubation to the depletion of labile carbon sources (Angert et al., 2015; Severinghaus, 1995). Our short-term
229 incubations demonstrated that root respiration has a greater ARQ than microbial respiration from root-free soils. During the
230 growing season, root respiration and exudation increase, which should increase ARQ, as seen in our data. In Eastern U.S.
231 deciduous forests, root exudation rates tend to be lower in the winter and spring than in the summer and fall (Abramoff and
232 Finzi, 2016; Phillips et al., 2008). Mass-specific fine root respiration rates were greater during the growing season (up to 8
233 $\text{nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$) than in the winter ($< 1 \text{ nmol CO}_2 \text{ g}^{-1} \text{ s}^{-1}$) and total belowground carbon flux was greatest from May through
234 October (Abramoff and Finzi, 2016). Though these root studies were not from the western United States, eddy covariance data
235 from a coniferous forest near our study site found that primary production was greatest during the summer months from June
236 through mid-September (Goldstein et al., 2000).

237

238 Beyond the results of our root and root-free soil incubations, there is additional evidence that root and rhizosphere respiration
239 should have a greater ARQ than microbial-derived respiration. For example, respiration of root tips is driven by sugar content
240 and has an RQ of 1.0 (Saglio and Pradet, 1980). Furthermore, recent metabolomic analysis of root exudates identified sugars,
241 carboxylic acids, amino acids, and phenolics as the main metabolites (Zhalnina et al., 2018), most of which are relatively
242 oxidized energy sources with greater respiratory quotients. Thus, we would expect greater ARQ values during the summer due
243 to higher root activity. When trees are dormant, the lack of fresh inputs from roots may lead to more recycling of organic
244 carbon within microbial biomass, wherein proteins and lipids are the first and third largest constituents by weight, making up
245 to 55% and from 10-35% of a typical bacterial cell's dry mass, respectively (Kleber and Reardon, 2017; Neidhardt, 1987).
246 Lipids and proteins tend to be reduced and have the lowest RQ values of common organic substrates, likely explaining the
247 lower wintertime ARQ values in our soils.

248

249 The seasonal pattern in $\delta^{13}\text{CO}_2$ reinforces our interpretation that changes in respiration carbon sources were driving changes
250 in ARQ. Soil $\delta^{13}\text{CO}_2$ was more enriched in the summer and became more depleted in the winter by up to 2‰. In a
251 comprehensive review of carbon isotopes in terrestrial ecosystems, Bowling et al. (2008) showed that plant lipids tend to be



252 more depleted in ^{13}C while sugars and organic acids tend to be more enriched in ^{13}C relative to bulk leaf $\delta^{13}\text{C}$. While these
253 numbers are based on plant lipids, if we assume microbial lipids are similarly depleted relative to other organic compounds,
254 an increase in microbial necromass as an organic matter source relative to root-derived sources during the winter would cause
255 the observed fluctuation in $\delta^{13}\text{CO}_2$. Furthermore, a chemical fractionation of soil organic matter found that the water-soluble
256 fraction, which includes sugars, was 3–4‰ more enriched than the acid-insoluble pool (Biasi et al., 2005). While the
257 interpretation of respiration $\delta^{13}\text{C}$ by itself in C_3 ecosystems can be difficult due to the small ‰ differences among carbon
258 sources (e.g., Bowling et al., 2015), the simultaneous use of ARQ and $^{13}\text{CO}_2$ helps strengthen interpretations.

259

260 Seasonality encompasses changes to phenology and soil climate. Both ARQ and $\delta^{13}\text{C}$ had significant positive relationships
261 with soil temperature. In addition to the importance of plant phenology described above, temperature could have direct effects
262 on respiration sources. Specifically, warmer temperatures can increase root exudation rates (Yin et al., 2013) and the relative
263 contribution of autotrophic-derived, if not directly autotrophic, respiration to total soil respiration. In two subarctic ecosystems,
264 warming increased the proportion of ecosystem respiration derived from autotrophs (which, using natural abundance
265 radiocarbon as a tracer, included heterotrophic respiration of root exudates) relative to heterotrophs (Hicks Pries et al., 2015).
266 However, temperatures can affect ARQ through more than just changing the contributions of autotrophic sources. Lower
267 temperatures increase the thermodynamic favorability of the oxidation of reduced carbon in compounds like lipids (LaRowe
268 and Van Cappellen, 2011), which could also explain the decrease in ARQ values at lower temperatures. For $\delta^{13}\text{C}$, it is likely
269 that phenological changes to organic carbon sources were more important than temperature per se. Several soil incubation
270 studies show that increases in temperature cause respired $\delta^{13}\text{CO}_2$ to become depleted by about 0.12–0.35‰ for each 1°C rise
271 in temperature—the opposite of the relationship we found (Andrews et al., 2000; Biasi et al., 2005; Hicks Pries et al., 2013).
272 In these incubations, which were devoid of new organic carbon inputs, unlike in situ conditions, the shift was attributed to
273 changes to the microbial community that affected carbon source preferences (Andrews et al., 2000; Biasi et al., 2005).
274 Furthermore, in a Mediterranean climate, phloem sap from trees has been shown to become more enriched in $\delta^{13}\text{C}$ during the
275 summer (Merchant et al., 2010), matching our pattern in soil $\delta^{13}\text{CO}_2$.

276

277 Soil temperature and soil moisture were so strongly negatively correlated due to our study site's Mediterranean climate that it
278 is difficult to separate their effects. ARQ and $\delta^{13}\text{CO}_2$ were negatively correlated with volumetric water content, which was
279 greatest when soil temperatures were coldest. Volumetric water content has the potential to control ARQ in several ways. First,
280 increased soil moisture reduces O_2 availability, which could increase ARQ values >1 as CO_2 is produced without additional
281 O_2 consumption. However, during our study the soil remained oxic (soil O_2 averaged 20% and the minimum was 17.38%).
282 The negative relationship between ARQ and soil moisture indicates that anaerobic respiration was not a driver, and we only
283 measured one ARQ value greater than one (1.03) during our study. However, diffusion rates are lower with higher soil
284 moisture, which could make detection of high ARQ values difficult if anoxic conditions occur within microaggregates. In
285 anoxic microaggregates, iron (II) is produced anaerobically, which is subsequently oxidized to iron (III) as the aggregate dries



286 and becomes aerobic, a process that consumes O₂ without producing CO₂, resulting in low ARQ values that can be detected
287 as drying soils increase diffusion (Angert et al., 2015). In our soils, which tend to contain relatively high amounts of iron
288 oxides (Rasmussen et al., 2005), iron oxidation could explain the 15% of ARQ values that were less than the reduced organic
289 matter value of 0.7. Lastly, since CO₂ is more soluble in water than is O₂, more CO₂ relative to O₂ is expected to dissolve in
290 soil water, which would reduce ARQ values at higher moisture contents. However, different dissolution rates and iron
291 oxidation do not fully explain our data as the wide variability in ARQ values (0.44 to 0.94) at high volumetric water contents
292 (0.27 to 0.31) can be best explained by time of year (Fig. A1), which again points to phenology as the main driver; the greater
293 ARQ values are from April and June while the lower values are from February and March. Furthermore, there was a stronger
294 relationship between observed and predicted ARQ in the temperature model than from the soil moisture model.

295

296 The reasons for $\delta^{13}\text{CO}_2$ becoming more depleted with increasing volumetric water content are not clear. Based on kinetics, we
297 would expect that as more CO₂ dissolves in water, the soil air should become enriched in ¹³CO₂ because dissolution
298 discriminates against the heavy isotope and increasingly so at lower temperatures (Zhang et al., 1995), but our data were not
299 consistent with this explanation.

300

301 The warming treatment caused slightly greater ARQ values (only in the winter) and slightly more enriched $\delta^{13}\text{CO}_2$ relative to
302 the controls; trends that are similar to those followed by ARQ and $\delta^{13}\text{CO}_2$ with soil temperature. The warming treatment effect
303 in terms of ARQ and $\delta^{13}\text{CO}_2$ was relatively small (0.07 units and 0.48‰), but its effect on soil carbon flux was large. Warming
304 increased CO₂ production by 34 to 37% with about 50% of the respiration and 40% of the warming response occurring below
305 15 cm in the soil profile (Hicks Pries et al., 2017).

306

307 The direction of the shift in ARQ and $\delta^{13}\text{CO}_2$ with warming to slightly higher ARQ and more enriched $\delta^{13}\text{CO}_2$ values indicates
308 proportionately more respiration of relatively oxidized, labile organic substrates. In principle, the shifts in ARQ and $\delta^{13}\text{CO}_2$
309 could also result from a relative increase in availability of labile organic substrates, perhaps as a result of enhanced root growth
310 and exudation (Yin et al., 2013). Given that ARQ only differed by treatment in the winter when trees were less active, however,
311 the former situation of preferential decomposition seems more likely. If so, this could lead to exhaustion of that pool and
312 eventually lower warming-induced CO₂ losses as seen at Harvard Forest (Melillo et al., 2002, 2017). Further measurements of
313 CO₂ production, ARQ, and $\delta^{13}\text{CO}_2$ as warming progresses will help distinguish these cases.

314

315 Depth was the only parameter by which ARQ and $\delta^{13}\text{CO}_2$ did not change in concert with one another. ARQ decreased with
316 depth while $\delta^{13}\text{CO}_2$ became more enriched. The decrease in ARQ with depth, which was more dramatic in the root-free soil
317 incubations than in soil air (difference of 0.08 versus 0.03), is likely due to decreased plant inputs with fewer fine roots and
318 less root exudation at depth (Hicks Pries et al., 2018; Tückmantel et al., 2017). The enrichment of soil $\delta^{13}\text{CO}_2$ likely reflects
319 the near-universal enrichment of soil organic carbon with depth due to catabolic carboxylation reactions (as microbial



320 byproducts and necromass become a larger proportion of soil organic matter; Ehleringer et al., 2000; Torn et al., 2002) or the
321 Suess effect (the continuing depletion of atmospheric CO₂ over time due to the burning of fossil fuels). In our soils, there was
322 about a 2‰ enrichment in bulk soil organic δ¹³C with depth (Hicks Pries et al., 2018).

323 **5 Conclusion**

324 Here we have shown, for the first time, both annual patterns in soil ARQ and how ARQ is affected by experimental warming.
325 The seasonal patterns in ARQ were likely due to changes in the substrates providing the energy for soil respiration with root-
326 derived sugars and organic acids being the dominant substrates during the growing season and microbial necromass being the
327 dominant substrate during the winter. Our inferences of organic substrates based on ARQ were supported by soil δ¹³CO₂
328 measurements, which showed clear patterns despite our study system containing only C₃ plants. We have shown how ARQ
329 measurements can help to disentangle the biological sources contributing to soil respiration and to understand how sources are
330 shifting due to global change. This application of ARQ worked well in our soils, which were well-drained, oxygenated, and
331 lacked carbonates. The interpretation of soil ARQ values becomes more complex if those conditions are not met (Angert et
332 al., 2015). The autotrophic and heterotrophic source separation in our incubations indicates ARQ has the potential to be used
333 to partition soil respiration in a similar manner to natural abundance δ¹³C (e.g., Dorrepaal et al., 2009; Hicks Pries et al., 2013).
334 To enable further applications of ARQ, more characterization is needed of the controls of the ratio, including incubation studies
335 of sterile and ‘live’ soils under aerobic and anaerobic conditions and co-located measurements of ARQ fluxes and the oxidative
336 ratio of organic matter sources as in Masiello et al. (2008). Such future investigations will help determine whether ARQ
337 deserves a prominent place alongside natural abundance isotopes in the ecosystem ecology and biogeochemistry toolkit.

338 **Data Availability**

339 Data is being made publicly available on ESS-DIVE (<http://ess-dive.lbl.gov/>).

340 **Author Contribution**

341 CHP, AA, and MST conceived of the study. Field measurements were conducted by CHP and CC. Lab analyses were
342 conducted by CHP and BH. Statistical analyses were conducted by CHP. CHP wrote the manuscript with feedback from all
343 authors.

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480

Molecule	RQ ^a	$\delta^{13}\text{C}$ (relative to bulk leaf) ^b
Organic acids	1.4 (1-4)	+0.75
Sugars	1.0	+1.5-2
Phenolics	0.95	NA
Proteins	0.77 (0.67-1.0)	+1
Lignin	0.88	-3
Lipids	0.73	-4

481 ^aData from Masiello et al. 2008

482 ^bData from Bowling et al. 2008

483 **Table 1. Respiration quotient (RQ) and relative isotopic enrichment of common molecules/substrates for respiration found in soils.**

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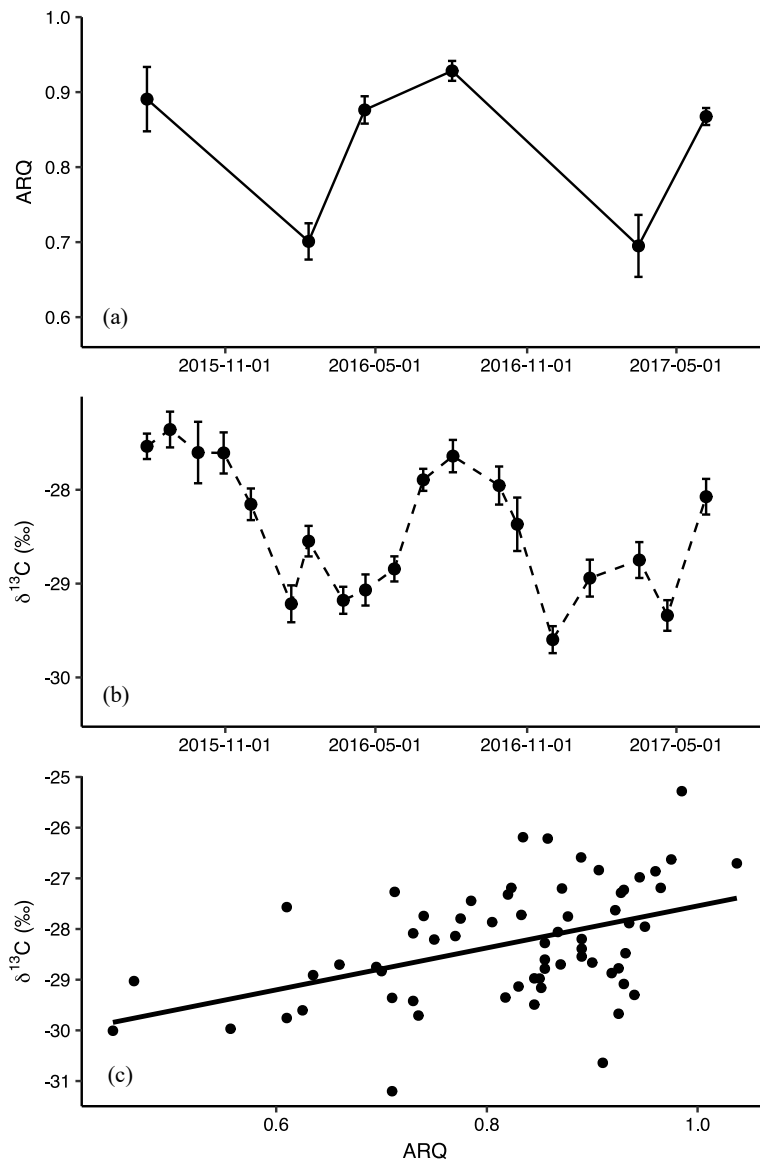
Depth (cm)	$\delta^{13}\text{CO}_2$ (‰)	ARQ
30	-29.0 ± 0.09	0.84 ± 0.02
50	-28.6 ± 0.08	
70	-28.4 ± 0.07	
90	-28.3 ± 0.08	0.81 ± 0.02

489 **Table 2. Mean (± SE) corrected $\delta^{13}\text{CO}_2$ and ARQ of soil pore space by depth.**

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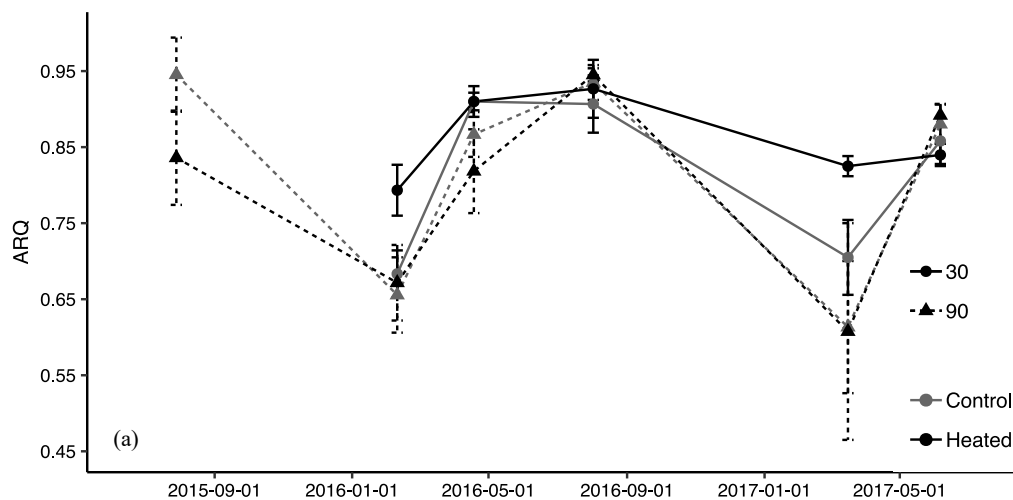
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493 **Figure 1.** Mean (\pm SE) apparent respiration quotient (ARQ; a) and corrected $\delta^{13}\text{C}$ (b) in soil pore space air averaged across all
494 depths and treatments by sampling month. The relationship between ARQ and $\delta^{13}\text{C}$ values over the months when they were
495 sampled simultaneously (c). The line shows the fit of a mixed model regression where gas well was treated as a random effect.

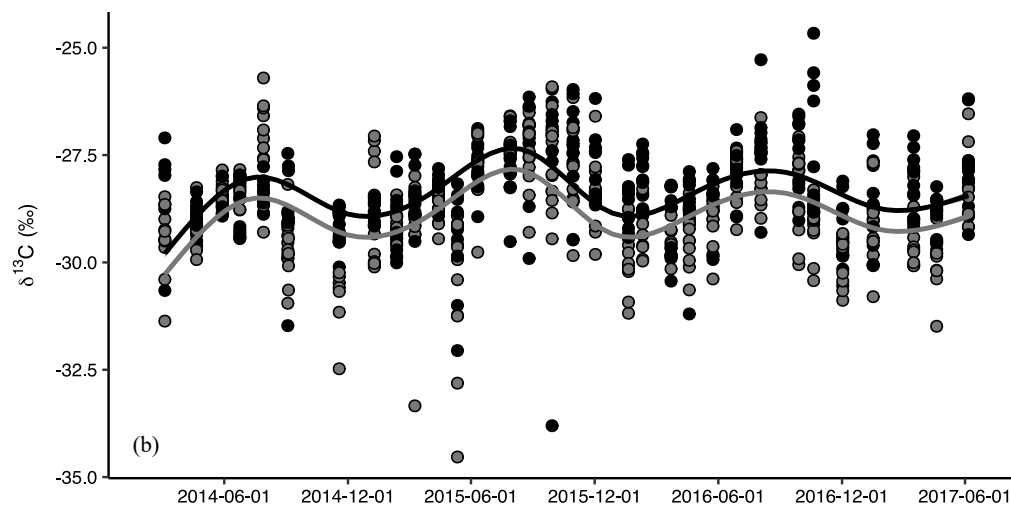
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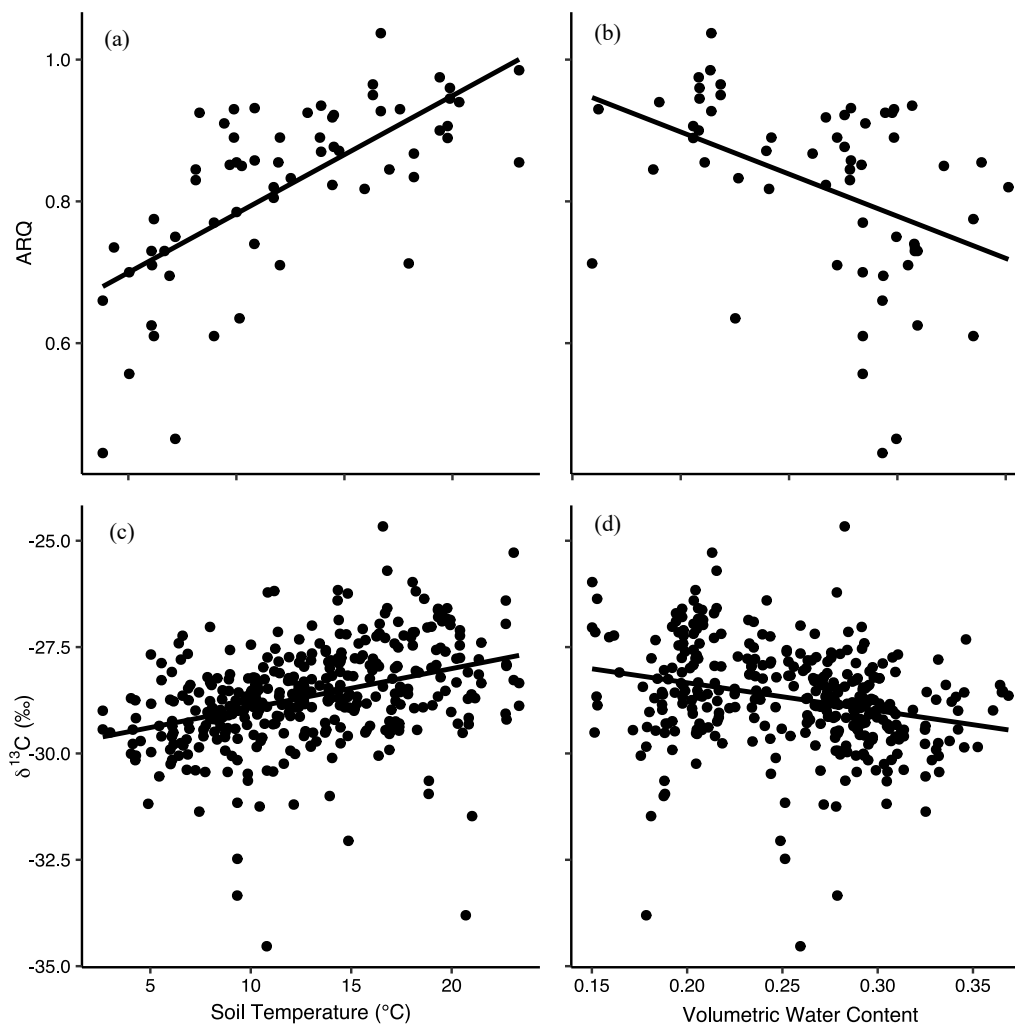
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502 **Figure 2.** Mean (\pm SE) apparent respiration quotient (ARQ) by sampling date for heated (black) and control (grey) treatments at 30
503 cm (circles) and 90 cm (triangles) depths (a). ARQ only differed significantly among treatments during the winter at 30 cm.
504 Corrected $\delta^{13}\text{CO}_2$ for all depths (30, 50, 70, and 90 cm) and months sampled (b). The lines represent the predicted fit of a sinusoidal
505 regression (see text) for an average soil depth in control (grey) and heated (black) treatments.



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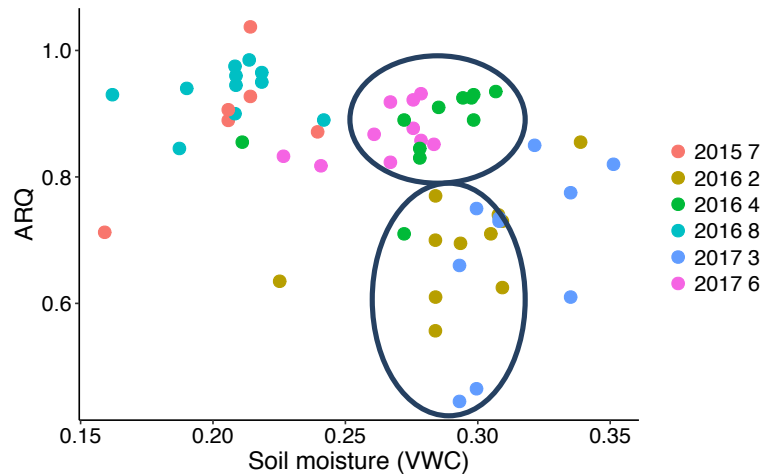
507 **Figure 3.** The relationships of apparent respiration quotient (ARQ; a, b) and $\delta^{13}\text{CO}_2$ (c, d) by soil temperature (a, c) and soil moisture
508 (b, d). The points represent data taken from an individual gas well. The lines show the fit of a mixed model regression between each
509 variable where individual gas well was treated as a random effect.

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514 **Figure A1. The variability in the apparent respiration quotient's (ARQ) relationship with soil moisture at 0.25 to 0.30 VWC can be**
515 **explained by time of year. The greater ARQ values are from April and June while the lower values are from February and March.**
516 **Colors are by sampling date shown as year and month.**

517

518