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4 A genetic link between whole-plant water use efficiency and leaf carbon isotope
5 composition in the C₄ grass *Setaria*

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7 Short title: Water use efficiency and C₄ leaf carbon isotopes

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21

22 **Abstract**

23 Increasing whole plant water use efficiency (yield per transpiration; WUE_{plant}) through
24 plant breeding can benefit the sustainability of agriculture and improve crop yield
25 under drought. To select for WUE_{plant} , an efficient phenotyping method that reports on
26 the genetic contribution of component traits such as transpiration efficiency (TE_i ; rate of
27 CO_2 assimilation per stomatal conductance) must be developed. Leaf carbon stable
28 isotope composition ($\delta^{13}C_{\text{leaf}}$) has been proposed as a high-throughput proxy for TE_i ,
29 and a negative correlation between $\delta^{13}C_{\text{leaf}}$ and both WUE_{plant} and TE_i has previously
30 been demonstrated in several C_4 grass species. Therefore, the aim of the research
31 presented here was to determine if the same loci control $\delta^{13}C_{\text{leaf}}$, WUE_{plant} , and TE_i under
32 well-watered and water-limited conditions in a recombinant inbred line (RIL)
33 population of closely related C_4 grasses *Setaria viridis* and *S. italica*. Three quantitative
34 trait loci (QTL) for $\delta^{13}C_{\text{leaf}}$ were co-localized with transpiration, biomass, and a linear
35 model of WUE. When WUE_{plant} was calculated for allele classes based on the three QTL
36 for $\delta^{13}C_{\text{leaf}}$, $\delta^{13}C_{\text{leaf}}$ was negatively correlated with WUE_{plant} as theory predicts when
37 WUE_{plant} is in part driven by differences in TE_i . In any population, multiple traits can
38 influence WUE_{plant} ; however, the analysis of $\delta^{13}C_{\text{leaf}}$ in this RIL population demonstrates
39 that there is genetic control of TE_i that significantly contributes to WUE_{plant} .
40 Furthermore, this research suggests that $\delta^{13}C_{\text{leaf}}$ can be used in marker-assisted breeding
41 to select for TE_i and as a tool to better understand the physiology and genetic
42 architecture of TE_i and WUE_{plant} in C_4 species.

43

44 **Significance Statement**

45 Overextended water resources and drought are major agricultural problems
46 worldwide. Therefore, selection for increased plant water use efficiency (WUE_{plant}) in
47 food and biofuel crop species is an important trait in plant breeding programs. Leaf
48 carbon isotopic composition ($\delta^{13}C_{\text{leaf}}$) has potential as a rapid and effective high
49 throughput phenotyping method for intrinsic transpiration efficiency (TE_i), an
50 important leaf-level component trait of WUE_{plant} . Our research shows that $\delta^{13}C_{\text{leaf}}$ and
51 WUE_{plant} share a common genetic architecture through their shared relationship with
52 TE_i . This suggests that $\delta^{13}C_{\text{leaf}}$ can be used as a screen for TE_i in marker-assisted plant
53 breeding programs to improve crop drought resistance and decrease agricultural water
54 consumption.

55

56 Introduction

57 Water availability constrains agricultural production and threatens food security
58 in many drought-prone regions (1). Therefore, improving the harvestable yield relative
59 to water supplied to crop systems (agronomic water use efficiency; WUE_{ag}) has long
60 received attention from researchers and government agencies (2-5). It has been
61 proposed by Passioura (2) that yield could be improved relative to available water by
62 increasing (1) the ratio of transpiration (T) to evapotranspiration (ET), (2) whole plant
63 water use efficiency (ratio of biomass production to total transpiration; WUE_{plant}), and
64 (3) harvest index (HI). To date, increases in WUE_{ag} have been primarily made by
65 improved management practices that increase T/ET through improved water
66 application (6, 7), increased canopy cover (8) and mulching (6). Additionally, selecting
67 for greater HI has also increased WUE_{ag} , for example, with semi-dwarf wheat varieties
68 (9). Unfortunately, in many agricultural settings, these traits appear to have approached
69 their theoretical maximum.

70 To date, there has been limited change to WUE_{ag} through selection for increased
71 WUE_{plant} . This is primarily because WUE_{plant} is a complex trait and is influenced by 1) CO_2
72 assimilation (A_{net}) relative to water loss via stomata conductance (g_s), (i.e. the intrinsic
73 transpiration efficiency, A_{net}/g_s ; TE_i), 2) the portion of carbon loss from whole plant
74 respiration (ϕ_c), 3) “nonproductive” water loss from cuticular and nighttime
75 transpiration (ϕ_w), and 4) the evaporative demand between the atmosphere and the
76 plant (See theory; 10, 11, 12). Theoretically, the first three of these factors can be selected
77 for through plant breeding, but these traits, especially ϕ_c and ϕ_w , are determined by a
78 complex set of composite traits that are difficult to measure and select for in breeding
79 programs (13-15). Alternatively, in theory, TE_i is an ideal trait to select for because it is
80 independent of environmental conditions driving changes in evaporative demands (16,
81 17), and it is an important component of WUE_{plant} as it relates to both CO_2 and H_2O leaf
82 exchange, influencing both photosynthetic capacity and T (15, 18). Unfortunately, the
83 primary means of estimating TE_i with gas exchange measurements of A_{net}/g_s do not
84 integrate over time and generally do not represent TE_i over the lifetime of the plant or
85 even the leaf (18). Furthermore, these measurements are prohibitively time-consuming
86 and laborious, making this method impractical for selecting for WUE_{plant} in a plant-
87 breeding program.

88 Alternatively, leaf carbon isotope composition ($\delta^{13}\text{C}_{\text{leaf}}$) has long been promoted
89 as a proxy for an integrated measurement of TE_i in C_3 and potentially in C_4 species (10-
90 12, 19, 20). In C_3 plants, the relationship between $\delta^{13}\text{C}_{\text{leaf}}$ and TE_i has been tested and
91 even integrated into breeding programs (12, 15, 18, 21-26). However, in C_4 species, it
92 remains uncertain if $\delta^{13}\text{C}_{\text{leaf}}$ is an effective proxy for TE_i and if there is a genetic link
93 between these two traits. Nonetheless, empirical evidence supporting the theoretical
94 relationship between $\delta^{13}\text{C}_{\text{leaf}}$ and TE_i has been documented in multiple C_4 species such
95 as *Setaria viridis*, *S. italica*, *Zea mays*, and *Sorghum bicolor* (27-29). These studies also
96 demonstrated consistent differences in $\delta^{13}\text{C}_{\text{leaf}}$ between well-watered and water-limited
97 plants that negatively correlated with TE_i . Additionally, in *S. viridis* and *S. italica*, TE_i
98 correlated with $\text{WUE}_{\text{plant}}$ (29). However, other studies on both C_3 and C_4 species have
99 found that TE_i and $\text{WUE}_{\text{plant}}$ were not correlated (30-40). Therefore, the contribution of
100 TE_i to $\text{WUE}_{\text{plant}}$ requires further investigations to delineate those factors that collectively
101 contribute to $\text{WUE}_{\text{plant}}$.

102 The genetic loci controlling $\text{WUE}_{\text{plant}}$ and its relationship to TE_i and $\delta^{13}\text{C}_{\text{leaf}}$ can
103 potentially be identified using large mapping populations grown on automated
104 phenotyping systems that measure plant water use and biomass accumulation on
105 hundreds of individual plants (41, 42). In fact, quantitative trait loci (QTL) have been
106 found for $\delta^{13}\text{C}_{\text{leaf}}$ in several C_3 species such as rice (43-45), barley (46), *Brachypodium*
107 *distachyon* (47), wheat (48), tomato (49), Arabidopsis (50, 51), sunflower (52), soybean
108 (53), cotton (54), *Quercus ruber* (55), and *Stylosanthes scabra* (56). Additionally, a few
109 studies on C_3 plants have found co-localized QTL for $\delta^{13}\text{C}_{\text{leaf}}$ and $\text{WUE}_{\text{plant}}$ (52, 57), and,
110 in one case, $\delta^{13}\text{C}_{\text{leaf}}$ and TE_i were associated with a causal gene (ERECTA; 58). However,
111 to date only one publication found that $\delta^{13}\text{C}$ was under genetic control in a C_4 species
112 (maize; 59), but this was not tested in terms of WUE. Therefore, to effectively use
113 marker-assisted breeding to select for $\text{WUE}_{\text{plant}}$ and TE_i in C_4 plants requires a more
114 thorough understanding of the physiological relationship and genetic architecture of
115 $\delta^{13}\text{C}_{\text{leaf}}$, TE_i , and $\text{WUE}_{\text{plant}}$.

116 Here a recombinant inbred line (RIL) population of 189 lines created from
117 accession A10 of *S. viridis* (L.) P. Beauv. and accession B100 of *S. italica* (L.) P. Beauv was
118 used to screen for $\text{WUE}_{\text{plant}}$, TE_i , and $\delta^{13}\text{C}_{\text{leaf}}$ (42, 60, 61). Both *S. viridis* and *S. italica* are
119 model C_4 grasses in the same panicoid clade as important C_4 crops such as maize,
120 sugarcane, sorghum and emerging bioenergy crops Miscanthus and switchgrass. The

121 objectives of this study were to compare $\delta^{13}\text{C}_{\text{leaf}}$ between plants grown under well-
122 watered and water-limited conditions and to determine the genetic and physiological
123 relationship between $\text{WUE}_{\text{plant}}$, TE_i , and $\delta^{13}\text{C}_{\text{leaf}}$.

124 **Results**

125 *Fresh biomass, transpiration, $\text{WUE}_{\text{ratio}}$, WUE_{fit} , and WUE_{res} traits*

126 Although whole plant growth and water use were analyzed throughout the
127 experiment (42) we selected day 25 for our analysis when most genotypes were in the
128 vegetative stage. On this day, the fresh biomass estimated from side view images and
129 validated with final harvest biomass (42) varied across genotypes from 3.70 to 29.00 g
130 and 1.40 to 10.30 g in the well-watered and water-limited treatments, respectively.
131 Cumulative transpiration ranged from 367 to 1433 ml in the well-watered and 96 to 433
132 ml in water-limited plants. There was a significant difference in fresh biomass and
133 transpiration rates between genotypes in both irrigation treatments (Fig. 1), and across
134 genotypes the water-limited treatment significantly reduced fresh biomass and
135 transpiration by 64 and 65 %, respectively (Table 1). The ratio of biomass relative to the
136 amount of total plant transpiration ($\text{WUE}_{\text{ratio}}$) was 20 % higher in the water-limited
137 treatment and ranged across genotypes from 1.2 to 36.7 (g/L) in the well-watered
138 treatment and 3.9 to 41.7 (g/L) in the water-limited treatment, respectively (Table 1).

139 Across genotypes, output of the linear model of fresh biomass versus
140 transpiration (WUE_{fit}) ranged from 1.84 to 36.24 g in the well-watered and 0.49 to 9.94 g
141 in the water-limited treatments (Table 1). In the well-watered treatment, the residual of
142 the WUE model (WUE_{res}) varied from -9.61 to 10.06 g and in the water-limited treatment
143 from -6.22 to 2.25 g, with a significant genotype effect within each treatment. Because
144 WUE_{res} was calculated for each treatment separately and centered around 0, no
145 difference was expected between treatments (Table 1).

146 *Broad-sense heritability, proportional variance and leaf carbon isotopic composition ($\delta^{13}\text{C}_{\text{leaf}}$)*

147 In all traits, except WUE_{res} , 32 to 86 % of the variance in the experiment was
148 explained by the treatment effect (Table 2). Additionally, in all traits, except $\text{WUE}_{\text{ratio}}$
149 and WUE_{res} , the variance ascribed to the genotype effect was relatively small but
150 substantial given the large influence that the water limitation treatment had on these
151 traits.

152 Broad-sense heritability (H^2) was relatively robust for all traits, including $\delta^{13}\text{C}_{\text{leaf}}$,
153 in at least one treatment or when treatments were combined (Table 2). For example,

154 $\delta^{13}\text{C}_{\text{leaf}}$ was significantly heritable in the well-watered treatment but not in the water-
155 limited treatment. This is true of other traits, except WUE_{res} , where the well-watered
156 treatment had higher H^2 than the water-limited treatment or combined treatments. The
157 decrease of H^2 in the water-limited compared to the well-watered treatment was
158 pronounced in all traits except fresh biomass and WUE_{res} . The $\delta^{13}\text{C}_{\text{leaf}}$ values ranged
159 from -14.7 to -12.4 ‰ in the well-watered and -15.6 to -13.2 ‰ in the water-limited
160 treatment, with significant differences across genotypes (Fig. 1; Table 1). The water-
161 limited treatment significantly reduced $\delta^{13}\text{C}_{\text{leaf}}$ on average across genotypes by $0.82 \pm$
162 0.04 ‰ (Table 1).

163 *Correlation of traits with $\delta^{13}\text{C}_{\text{leaf}}$*

164 Over the time course of the experiment, the correlations of $\delta^{13}\text{C}_{\text{leaf}}$ with
165 transpiration, biomass, and WUE_{fit} were constant except for a few days at the beginning
166 and end of the experiment, so at the midpoint of the experiment (day 25), correlations
167 with $\delta^{13}\text{C}_{\text{leaf}}$ were similar in magnitude to most days of the experiment. These
168 correlations were stronger across treatments than within treatment, yet they were
169 significant under the well-watered conditions (Fig. S1). On day 25, the correlation
170 coefficients of $\delta^{13}\text{C}_{\text{leaf}}$ with fresh biomass, transpiration, and WUE_{fit} were between 0.51
171 and 0.61 across treatments and 0.30 to 0.33 for well-watered treatment (Fig. 1). For
172 $\text{WUE}_{\text{ratio}}$ and WUE_{res} , the correlation coefficients were low in both treatments and when
173 the treatments were combined (Fig. S1).

174 *QTL analysis and contributions of allele composition on traits*

175 Three QTL (chr. 7@51 centimorgans (cM), chr. 7@99 cM and chr. 9@34 cM)
176 associated with $\delta^{13}\text{C}_{\text{leaf}}$ were found in the well-watered treatment, but none were
177 detected in the water-limited treatment (Table 3). All three of these QTL co-localized
178 with WUE_{fit} and transpiration in both treatments on day 25 (as well as days 17 through
179 33 as described in Feldman, *et al.* (42). Two QTL (7@99 and 9@34) were also associated
180 with fresh biomass in both treatments (Table 3). Furthermore, one QTL (9@34) of
181 $\text{WUE}_{\text{ratio}}$ was co-localized with $\delta^{13}\text{C}_{\text{leaf}}$. WUE_{res} were associated with two QTL (2@96 and
182 5@109) that co-localized with fresh biomass, transpiration, $\text{WUE}_{\text{ratio}}$, and WUE_{fit} in both
183 treatments, but not with $\delta^{13}\text{C}_{\text{leaf}}$. Having an allele from parental accession A10 (*S. viridis*)
184 at two of the three loci (7@51 and 7@99) increased $\delta^{13}\text{C}_{\text{leaf}}$ in the well-watered treatment,
185 while an allele from parental line B100 (*S. italica*) increased $\delta^{13}\text{C}_{\text{leaf}}$ at 9@34 in both
186 treatments (Fig. S2).

187 Combining alleles from the three QTL associated with $\delta^{13}\text{C}$ produced eight
188 possible allele classes, and all but one (ABA) was present in this population. For six of
189 the seven allele classes, the relationship between fresh biomass and transpiration was
190 significant (Table 4; Fig. 2A). Furthermore, the regression of $\delta^{13}\text{C}_{\text{leaf}}$ against the slopes of
191 fresh biomass versus transpiration showed a strong negative relationship for both the
192 well-watered ($\delta^{13}\text{C}_{\text{leaf}} = -0.027$ slope $- 12.48$; $R^2 = 0.69$; $P < 0.05$) and in the water-limited
193 treatments ($\delta^{13}\text{C}_{\text{leaf}} = -0.012$ slope $- 13.82$; $R^2 = 0.80$; $P < 0.05$; Fig. 2B). Although this
194 relationship was dampened in the water-limited treatment, it followed a similar trend
195 (Fig. 2B). Additionally, the order that the allele classes are positioned along the $\delta^{13}\text{C}_{\text{leaf}}$
196 versus slope regression is the same in both treatments. In the well-watered treatment,
197 the QTL 7@99 (represented by the second letter in three-letter allele class names)
198 appears to have the greatest influence on this relationship, where the A10 allele was
199 associated with a reduced slope and enriched $\delta^{13}\text{C}_{\text{leaf}}$ (Fig. 2B). Alternatively, in the
200 water-limited treatment, the effect of QTL 7@99 on this relationship was reduced
201 relative to the well-watered treatment. Additionally, in both treatments the mean fresh
202 biomass and transpiration for each of these allele classes had a strong significant
203 positive relationship with $\delta^{13}\text{C}_{\text{leaf}}$ (Fig. 3A-B).

204 Discussion

205 Leaf carbon isotope composition ($\delta^{13}\text{C}_{\text{leaf}}$) has been theoretically related to TE_i
206 (A_{net}/g_s) for both C_3 and C_4 species (12, 15, 18, 19, 27). However, $\delta^{13}\text{C}_{\text{leaf}}$ is less well-
207 understood in C_4 species because the CO_2 -concentrating mechanism dampens
208 variability in leaf CO_2 discrimination and ultimately $\delta^{13}\text{C}_{\text{leaf}}$ (62). However, in this study
209 we demonstrate significant genetic and environmental influence on $\delta^{13}\text{C}_{\text{leaf}}$ in a C_4
210 species that is driven by differences in water use efficiency. Although the variation in
211 $\delta^{13}\text{C}_{\text{leaf}}$ in C_4 species is typically less than in C_3 species, in this study, a relatively large
212 range in $\delta^{13}\text{C}_{\text{leaf}}$ (2.4 ‰) was observed across genotypes, and there was a significant
213 mean 0.82 ± 0.04 ‰ difference between treatments, showing considerable genotypic by
214 treatment response. These results are similar to previous studies of well-watered and
215 water-limited C_4 plants (29, 63, 64), suggesting $\delta^{13}\text{C}_{\text{leaf}}$ in C_4 plants is both genetically
216 determined and responsive to environmental conditions such as water limitation.

217 In the well-watered treatment, the positive correlations of fresh biomass,
218 transpiration, and WUE_{fit} with $\delta^{13}\text{C}_{\text{leaf}}$ and similarities in the genetic architecture of these
219 traits, suggest that TE_i (as represented by $\delta^{13}\text{C}_{\text{leaf}}$) is important in determining the

220 amount of biomass produced for a given volume of water transpired. However, in the
221 water-limited plants the low variation of $\delta^{13}\text{C}_{\text{leaf}}$, the lack of $\delta^{13}\text{C}_{\text{leaf}}$ correlations with
222 other traits (e.g. WUE_{ratio} and WUE_{res}), and no QTLs for $\delta^{13}\text{C}_{\text{leaf}}$ was likely due to
223 restricted stomatal conductance (g_s) across most genotypes, minimizing individual
224 differences in TE_i , as found for C_3 species (52, 65). As stated previously, WUE_{res}
225 represents how individuals deviate from WUE_{fit} either by A_{net}/g_s or other whole plant
226 process. However, the lack of overlap in genetic architecture and the lack of correlation
227 between WUE_{res} and $\delta^{13}\text{C}_{\text{leaf}}$ suggest that TE_i is not the primary driver of WUE_{res} . The
228 major QTL associated with WUE_{res} (2@96, 5@109) may potentially allocate carbon to
229 non-transpiring biomass such as structural and stem tissue (66). Additionally, the
230 similar genetic architecture for WUE_{res} and WUE_{ratio} , indicating that in this population
231 ϕ_w , ϕ_c , and r influence variation in WUE_{res} of individual plants relative to WUE_{fit} (42).

232 Three QTL of WUE_{fit} co-localized with $\delta^{13}\text{C}_{\text{leaf}}$ (7@51, 7@99, 9@34), suggesting that
233 these loci are related to genotypic variation in TE_i . Based on the theoretical and
234 empirical relationships, TE_i and $\delta^{13}\text{C}_{\text{leaf}}$ should be negatively correlated (20, 27, 62).
235 Therefore, if TE_i is a major contributor to WUE_{plant} within an allele class, then WUE_{plant}
236 should negatively correlate with $\delta^{13}\text{C}_{\text{leaf}}$. The WUE_{plant} , defined as the slope of the linear
237 regression of fresh biomass versus transpiration (1, 67), was negatively correlated with
238 $\delta^{13}\text{C}_{\text{leaf}}$ for each allele class. This suggests that $\delta^{13}\text{C}_{\text{leaf}}$ is genetically and physiologically
239 related to WUE_{plant} , likely through TE_i . This relationship between the allele class-specific
240 WUE_{plant} and $\delta^{13}\text{C}_{\text{leaf}}$ existed in the water-limited treatment as well, although no QTL for
241 $\delta^{13}\text{C}_{\text{leaf}}$ were found. Given that the water limitation did not remove the underlying
242 relationship between TE_i and WUE_{plant} , the inability to detect QTL is likely due to the
243 reduced variation in $\delta^{13}\text{C}_{\text{leaf}}$, which reduced the magnitude of the genotypic response
244 and decreased the signal to noise ratio. Alleles increasing WUE were contributed by
245 both the weedy *S. viridis* (A10 parental accession) and the domesticated *S. italica* (B100
246 parental accession). In addition, allele classes followed the same order along the $\delta^{13}\text{C}_{\text{leaf}}$
247 versus slope regression in both well-watered and water-limited treatment blocks. For
248 example, under both well-watered and water-limited conditions the allele class AAB
249 had the lowest WUE_{plant} and most enriched $\delta^{13}\text{C}_{\text{leaf}}$; whereas allele classes BBB and BBA
250 had the highest WUE_{plant} and most depleted $\delta^{13}\text{C}_{\text{leaf}}$ under both treatments. This trend
251 indicates a strong allelic effect on relationship between $\delta^{13}\text{C}_{\text{leaf}}$, TE_i , and WUE_{plant} that
252 will allow the selection for $\delta^{13}\text{C}_{\text{leaf}}$ to improve TE_i , WUE_{plant} , and drought tolerance.

253 Across allele classes, $\delta^{13}\text{C}_{\text{leaf}}$ also formed significant correlations with biomass
254 and transpiration. A positive correlation between $\delta^{13}\text{C}_{\text{leaf}}$ and fresh biomass is expected
255 if large plants have low TE_i and small plants have high TE_i . This scenario is possible, if
256 the photosynthetic capacity (i.e. chlorophyll and Rubisco) are diluted across large
257 leaves. This will decrease the photosynthetic rates per unit leaf area but on a per plant
258 basis would remain large relative to small plants with fewer small leaves. In this case,
259 the strong, positive relationship between $\delta^{13}\text{C}_{\text{leaf}}$ and transpiration would be driven by
260 increased leaf area in the large plants. However, C_4 photosynthesis is mostly CO_2
261 saturated at ambient CO_2 concentrations, so variation in TE_i is likely due to variation in
262 g_s across genotypes such that T decreases without negative impacts on A_{net} . This is
263 further supported from studies of C_3 species that showed that g_s and not A_{net} varied
264 with TE_i and $\delta^{13}\text{C}_{\text{leaf}}$ (67-70).

265 For $\delta^{13}\text{C}_{\text{leaf}}$ to be an effective screen for TE_i and $\text{WUE}_{\text{plant}}$ of a C_4 species, it must
266 show genotypic variation, be responsive to environmental conditions that influence TE_i ,
267 and be physiologically related to $\text{WUE}_{\text{plant}}$. In this C_4 grass RIL population, $\delta^{13}\text{C}_{\text{leaf}}$
268 showed significant and consistent response to water limitation, significant genotypic
269 variation, and high heritability. Additionally, $\delta^{13}\text{C}_{\text{leaf}}$ correlated with transpiration,
270 biomass, and the linear relationship between biomass and transpiration (WUE_{fit}),
271 suggesting a physiological relationship among these traits. This is further supported by
272 the fact that $\delta^{13}\text{C}_{\text{leaf}}$ and $\text{WUE}_{\text{plant}}$ had a strong negative correlation within the allele
273 classes defined by $\delta^{13}\text{C}_{\text{leaf}}$.

274 Both $\text{WUE}_{\text{plant}}$ and WUE_{fit} are driven by a balance between the intrinsic
275 relationship between carbon assimilation and water lost via stomates (TE_i) and other
276 whole plant processes such as ϕ_c , ϕ_w , and r (when only aboveground biomass is
277 measured). Hence, the relationship between $\delta^{13}\text{C}_{\text{leaf}}$ and $\text{WUE}_{\text{plant}}$ is only apparent if TE_i
278 has a strong influence on $\text{WUE}_{\text{plant}}$. Nevertheless, $\delta^{13}\text{C}_{\text{leaf}}$ can be used to detect variation
279 in TE_i that is not apparent from measuring $\text{WUE}_{\text{plant}}$ only. Therefore, this study
280 advances our understanding of $\text{WUE}_{\text{plant}}$, TE_i , and $\delta^{13}\text{C}_{\text{leaf}}$ in a C_4 species. Furthermore, it
281 illustrates that $\delta^{13}\text{C}_{\text{leaf}}$ has potential to be used in screening for TE_i in marker-assisted C_4
282 plant breeding and to better understand the genetic controls of $\text{WUE}_{\text{plant}}$ and its
283 components. Additional work is needed to explore the use of $\delta^{13}\text{C}_{\text{leaf}}$ in other C_4 species
284 and under field settings to better understand the complex interaction of traits and
285 causal genes that influence $\text{WUE}_{\text{plant}}$, TE_i , and $\delta^{13}\text{C}_{\text{leaf}}$.

286 **Methods**

287 *Plant material and growth conditions*

288 An interspecific *Setaria* F7 recombinant inbred line (RIL) population comprised of
289 189 genotypes was previously generated through a cross between the wild-type green
290 foxtail *S. viridis* accession, A10, and the domesticated *S. italica* foxtail millet accession,
291 B100 (60, 61, 71). Seeds from this population were sowed in 10 cm diameter white pots
292 pre-filled with ~470 cm³ of Metro-Mix 360 soil (Hummert, USA) and 0.5 g of Osmocote
293 Classic 14-14-14 fertilizer (Everris, USA) and placed on the Bellwether Phenotyping
294 System using a random block design. Two to three replicates per genotype, including
295 the A10 and B10 parental accessions, per treatment (1138 individuals) were grown for
296 25 days with a photoperiod of 16 h light / 8 h night, light intensity of 500 $\mu\text{mol}/\text{m}^2/\text{s}$, a
297 temperature regime of 31°C day/21°C night and relative humidity was maintained
298 between 40 – 80 %. Plants were watered 2-3 times per day to maintain plants at 100 %
299 pot capacity (PC) in the well-watered treatment and at 40% PC in the water-limited
300 treatment as determined by Fahlgren, *et al.* (41). Prescribed soil water content across
301 both treatment blocks was achieved by 15 days after planting. Additional detail on the
302 experimental design and plant growth can be found in Feldman, *et al.* (42), Feldman, *et*
303 *al.* (66).

304 *Measurements of biomass and transpiration*

305 The volume of water transpired by individual plants at each pot weighing was
306 calculated as the difference between the measured pot weight and the weight of the pre-
307 filled pot at pot capacity (100% PC for the well-watered treatment) or the difference
308 between current pot weight and the weight measurement on the previous day if no
309 water was added. The pots were watered to return the pot weight to the pre-set pot
310 weight, which maintains the gravimetric water content at either 100 % or 40 % PC for
311 the well-watered or water-limited treatment blocks, respectively. Initially all seedlings
312 were watered to pot capacity for the first two days on the Bellwether system. After two
313 days, the potting medium in the water-limited treatment was allowed to dry down to 40
314 % of PC, then maintained at the water content level of 40 %.

315 Plants were imaged every other day to measure side view area, which was used
316 to calculate fresh biomass (41, 42). The relationship between side view area and fresh
317 biomass was developed based on the final aboveground biomass and the last imaging
318 before harvesting. The sensitivity of the measurements limited reliable measurements of

319 plant size and transpiration from day 17 to 33 when plants were not too small for image
320 analysis and measurements of transpiration (42). Water use efficiency (WUE_{ratio}) was
321 calculated as the ratio of aboveground biomass and total water transpired. Instead of
322 comparing $\delta^{13}C_{leaf}$ to each day, the midpoint of the experiment was chosen (day 25) to
323 conduct the analysis. The plants on day 25 were in vegetative phase and growing
324 rapidly, typical of when physiological and gas exchange measurements would be made.
325 Additionally, day 25 was representative in terms of QTL and other analyses conducted
326 throughout the experiment (42).

327 In this population, plant size and cumulative whole-plant transpiration were
328 strongly correlated (42), so an ordinary least squares linear regression was used within
329 treatment blocks to model this relationship (Fig. S3). The within treatment model was
330 used to predict the accumulation of fresh biomass based on the amount of water
331 transpired (WUE_{fit}). The residuals surrounding WUE_{fit} (WUE_{res}) represent the fresh
332 biomass of individual plants not explained by WUE_{fit} (42).

333 *Leaf carbon stable isotopic composition ($\delta^{13}C_{leaf}$)*

334 The youngest, fully expanded leaf was collected during the final harvest at the
335 end of the experiment (day 34) and dried at 55 °C for three days. Once the leaves were
336 dried, 8-12 discs, having a total leaf area of 0.47 – 0.71 cm², were sampled from each leaf
337 and placed in tin capsules for stable isotopic analysis. A comparison of $\delta^{13}C_{leaf}$ from
338 sampling leaf discs to grinding and sampling an aliquot of the completely homogenized
339 powered leaf tissue was made on a subset of 47 leaves. The slope of $\delta^{13}C_{leaf}$ from the
340 punches regressed against $\delta^{13}C_{leaf}$ from the ground leaf tissue was 0.93 ± 0.03 ($R^2 = 0.96$;
341 Fig. S4) and the mean difference between methods was 0.06 ± 0.04 ‰, which was
342 similar to the IRMS precision and significantly less than the sample standard deviation
343 of 0.5 ‰. Considering the similarity between sampling methods, all leaves were
344 sampled using the more rapid hole punching method.

345 Leaf tissue was converted to CO₂ with an elemental analyzer (ECS 4010, Costech
346 Analytical, Valencia, CA) and analyzed with a continuous flow isotope ratio mass
347 spectrometer (Delta PlusXP, ThermoFinnigan, Bremen; (72, 73). The Santrock correction
348 was used by the IRMS software to correct for ¹⁷O (74). Final δ values were the mean of 5
349 sample peaks calibrated to the international standards NBS 19, RM 8542, and IAEA-CO-
350 9 to calculate $\delta^{13}C$ relative to Vienna Pee Dee belemnite (V-PDB). Quality control

351 standards were also included to determine the correction quality. Overall standard
352 deviation for $\delta^{13}\text{C}$ values was 0.07 ‰.

353 The stable isotope composition of carbon ($\delta^{13}\text{C}_{\text{leaf}}$) was reported in δ notation in
354 parts per thousand (‰),

$$\delta = \left(\frac{R_{\text{sample}}}{R_{\text{standard}}} - 1 \right) \quad \text{Equation 1}$$

355 where R_{sample} and R_{standard} is the molar ratios of heavy to light isotope ($^{13}\text{C}/^{12}\text{C}$) of the
356 sample and international standard, respectively. The international standard used for
357 oxygen was Vienna- PeeDee Belemite (VPDB).

358 *Statistical analysis*

359 Statistical analyses were conducted in R version 3.4.0 (R Team 75), using car
360 (version 2.0-26) packages for statistical tests and lmodel2 (version 1.7-2) package for
361 Model II regression analysis (standard major axis). Model II regressions were calculated
362 because neither variable was controlled, both varied naturally with their own associated
363 error, and the physical units of both variables were not the same. Homogeneity was
364 tested based on plotting predicted fit versus residuals. Using the extRemes package
365 (version 2.0-8), normality was tested by plotting residuals on quantiles-quantiles plots.
366 Within treatment comparisons were made on each trait using a two-factor analysis of
367 variance (ANOVA) where the factors were treatment and genotype. Slope of the
368 relationship between fresh biomass and transpiration was calculated with Model II
369 linear regression and determined to be different using analysis of co-variance
370 (ANCOVA). Methods used in QTL analysis were explained in Feldman, *et al.* (42),
371 Feldman, *et al.* (66).

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375

Appendix I: Glossary of terms

Term	Definition
$\Delta^{13}\text{C}$	Photosynthetic carbon discrimination ($\delta^{13}\text{C}_{\text{ambient}} - \delta^{13}\text{C}_{\text{leaf}}$)
$\delta^{13}\text{C}_{\text{leaf}}$	Leaf carbon isotopic composition (‰)
$\delta^{13}\text{C}_{\text{ambient}}$	Isotopic composition of ambient CO_2 (‰)
WUE_{ag}	Agricultural water use efficiency (crop yield/water applied to crop)
$\text{WUE}_{\text{plant}}$	Total biomass/water transpired
$\text{WUE}_{\text{biomass}}$	Theoretical ratio of aboveground biomass and water transpired
$\text{WUE}_{\text{ratio}}$	Measured ratio of aboveground biomass and water transpired
WUE_{fit}	Linear model fit of biomass regressed on transpiration
WUE_{res}	Residuals of the linear model between biomass and transpiration
TE_i	Intrinsic transpiration efficiency (A_{net}/g_s)
HI	Harvestable index
C_i/C_a	Intercellular to ambient CO_2 concentration
g_s	Stomatal conductance
A_{net}	Net photosynthetic rate
T	Transpiration rate
ET	Evapotranspiration
v	Evaporative demand; ($e_i - e_a$)
$e_i - e_a$	Water vapor molar difference between intercellular and ambient air at leaf temperature
ϕ_w	Proportion of water used by plant that is unproductive water loss (e.g. nighttime and cuticular transpiration)
ϕ_c	Proportion of fixed carbon lost through respiration
r	Proportion of biomass
a	Fractionation during diffusion of CO_2 in air through stomata (4.4 ‰)
b_3	Fractionation by Rubisco (30 ‰)
b_4	Fractionation of PEP carboxylation and isotopic equilibrium during dissolution and hydration of CO_2 (-5.2 ‰ at a leaf temperature of 30 °C)
s	Fractionation during the CO_2 leakage from the bundle sheath cells (1.8 ‰)
ϕ	Leakiness of CO_2 from the bundle sheath
H^2	Broad sense heritability

Appendix II. Theory

Agricultural water use efficiency (WUE_{ag}) can be defined as the crop yield per unit water supplied to the crop system. Therefore, yield relative to water use can be calculated as a function of evapotranspiration (ET), the proportion of ET that is transpired (T/ET), WUE_{plant} , and the harvest index (harvested proportion of biomass; Equation 1; 2, 15) as

$$Yield = ET \times \frac{T}{ET} \times WUE_{plant} \times HI. \quad \text{Equation S1}$$

Where WUE_{plant} relates to net CO_2 assimilation rates (A_{net}) relative to transpiration rate (T) at the whole plant level, and accounts for the proportion of fixed carbon lost by whole plant respiration (ϕ_c) and the proportion of "unproductive" water loss (ϕ_w) such as nighttime transpiration or cuticular evaporation (Equation 2; 11, 76). The relationship of these parameters to WUE_{plant} can be defined as

$$WUE_{plant} = \frac{A_{net}(1 - \phi_c)}{T(1 + \phi_w)} \quad \text{Equation S2}$$

where A_{net} and T are related through stomatal conductance (g_s) as

$$A_{net} = g_{sCO_2}(C_a - C_i) \quad \text{Equation S3}$$

and

$$T = g_{sH_2O}(e_i - e_a). \quad \text{Equation S4}$$

The parameter $(e_i - e_a)$ is the water vapor molar difference between intercellular and ambient air at leaf temperature, $(C_a - C_i)$ is the CO_2 molar difference between intercellular and ambient CO_2 , and g_{sCO_2} and g_{sH_2O} are the conductance values for CO_2 and H_2O , respectively (10-12). Substituting Eqs 3 and 4 into Eq 2 gives

$$WUE_{plant} = \frac{g_{sCO_2}(C_a - C_i)(1 - \phi_c)}{g_{sH_2O}(e_i - e_a)(1 + \phi_w)} = \frac{C_a \left(1 - \frac{C_i}{C_a}\right) (1 - \phi_c)}{1.6v(1 + \phi_w)} \quad \text{Equation S5}$$

where v is the evaporative demand ($e_i - e_a$), and the ratio of diffusivities of H₂O and CO₂ in air is 1.6. The molar ratio of intercellular to ambient CO₂ (C_i/C_a) influences WUE_{plant} because it represents the relative drawdown of intercellular CO₂ (C_i) by photosynthesis and the conductance of CO₂ into the leaf and the conductance of water vapor out the leaf via the stomata. Intrinsic TE (A_{net}/g_s ; TE_i) is equal to the CO₂ gradient from ambient to intercellular spaces ($C_i - C_a$), which can be rewritten as $TE_i = C_a(1 - C_i/C_a)$.

Therefore, Eq 5 can be simplified as a function of TE_i as

$$WUE_{plant} = \frac{TE_i(1 - \phi_c)}{1.6v(1 + \phi_w)} \quad \text{Equation S6}$$

When WUE is calculated with respect to aboveground biomass ($WUE_{biomass}$) as is the case with a bioenergy crop, biomass yield excludes the root fraction (r) and $WUE_{biomass}$ can be defined as

$$WUE_{biomass} = \frac{TE_i(1 - \phi_c)(1 - r)}{1.6v(1 + \phi_w)} \quad \text{Equation S7}$$

The relationship between $\delta^{13}C_{leaf}$ and TE_i is based on 1) variation in $\delta^{13}C_{leaf}$ (‰) of plants grown in the same atmospheric conditions is primarily controlled by leaf CO₂ isotope discrimination ($\Delta^{13}C$), 2) $\Delta^{13}C$ is influenced by changes in C_i/C_a and 3) C_i/C_a , as stated above, is affected by the interrelationship A_{net} and g_s . Therefore, TE_i (A_{net}/g_s) is related to C_i/C_a and, in turn, $\Delta^{13}C$ (10, 27).

Finally, leaf carbon composition ($\delta^{13}C$) is related to $\Delta^{13}C$ as

$$\Delta^{13}C = \frac{\delta^{13}C_{ambient} - \delta^{13}C_{leaf}}{1 + \delta^{13}C_{leaf}/1000} \quad \text{Equation S8}$$

where, $\delta^{13}C_{ambient}$ is the CO₂ isotopic signature of the CO₂ in the air surrounding the leaf, and $\delta^{13}C_{leaf}$ is the leaf carbon isotopic composition (20). In C₄ species, $\Delta^{13}C$ is primarily determined by fractionations associated with CO₂ carboxylation and diffusion, the ratio of bundle sheath CO₂ leak rate to PEP carboxylase rate (leakiness; ϕ), and C_i/C_a (Equation 9). Leakiness (ϕ) determines the slope of the relationship between $\Delta^{13}C$ and C_i/C_a and has been shown to be relatively constant in many C₄ species across light intensities, temperatures, and CO₂ partial pressures (77-80). Based on this mathematical

relationship, if ϕ is less than 0.37, then $\Delta^{13}\text{C}$ increases as C_i/C_a decreases, which corresponds with increasing $\delta^{13}\text{C}_{\text{leaf}}$. If ϕ is greater than 0.37, then the relationship reverses where $\Delta^{13}\text{C}$ increases with C_i/C_a . In *Setaria*, ϕ has been found to be less than 0.37, so C_i/C_a is expected to form a negative relationship with $\Delta^{13}\text{C}$ and positive relationship with $\delta^{13}\text{C}_{\text{leaf}}$ (Ellsworth et al. unpublished; 81). The relationship of $\Delta^{13}\text{C}$ and C_i/C_a can be defined by the simplified relationship as originally described by Farquhar (1984) as

$$\Delta^{13}\text{C} = a + (b_4 + \phi(b_3 - s) - a) \frac{C_i}{C_a} \quad \text{Equation S9}$$

where a is the fractionation during diffusion of CO_2 in air through stomata (4.4 ‰), b_4 is the fractionations of PEP carboxylation and the preceding isotopic equilibrium during dissolution and hydration of CO_2 (-5.7 ‰ at a leaf temperature of 25 °C) as described in (82, 83), b_3 is the Rubisco fractionation (29 ‰), and s is the fractionation during the leakage of CO_2 out of the bundle sheath cells (1.8 ‰) (27, 82).

Table 1: Analysis of variance of traits. Means \pm SD of fresh biomass, transpiration, WUE_{ratio} , WUE_{fit} , and WUE_{res} were determined on day 25. $\delta^{13}C_{leaf}$ was collected at the end of the experiment on day 34.

Trait	Treatment block		Treatment		Genotype		Treatment \times Genotype	
	Well-watered	Water-limited	$F_{ddf,ndf}$	P	$F_{ddf,ndf}$	P	$F_{ddf,ndf}$	P
Fresh biomass (g)	14.87 \pm 6.78	5.29 \pm 2.04	4248 _{1,744}	<0.0001	14.83 _{188,744}	<0.0001	6.11 _{182,744}	<0.0001
Transpiration (ml)	765.4 \pm 60.1	268.7 \pm 193.4	8395 _{1,764}	<0.0001	7.432 _{189,764}	<0.0001	3.321 _{183,764}	<0.0001
WUE_{ratio} (g/L)	19.7 \pm 5.0	24.5 \pm 5.0	515.8 _{1,744}	<0.0001	6.613 _{188,744}	<0.0001	1.063 _{182,744}	0.29
WUE_{fit} (g)	14.87 \pm 6.12	5.29 \pm 1.54	3299 _{1,744}	<0.0001	7.441 _{188,744}	<0.0001	3.942 _{182,744}	<0.0001
WUE_{res} (g)	-1.85 \pm 2.93	-1.85 \pm 1.34	0 _{1,744}	1.00	2.785 _{188,744}	<0.0001	0.97 _{182,744}	0.59
$\delta^{13}C_{leaf}$	-13.50 \pm 0.50	-14.33 \pm 0.55	569 _{1,351}	<0.0001	2.054 _{184,351}	<0.0001	1.003 _{175,351}	0.49

Table 2: Broad-sense heritability (H^2) and proportional variance of traits on day 25 after sowing. $\delta^{13}\text{C}_{\text{leaf}}$ was collected at the end of the experiment.

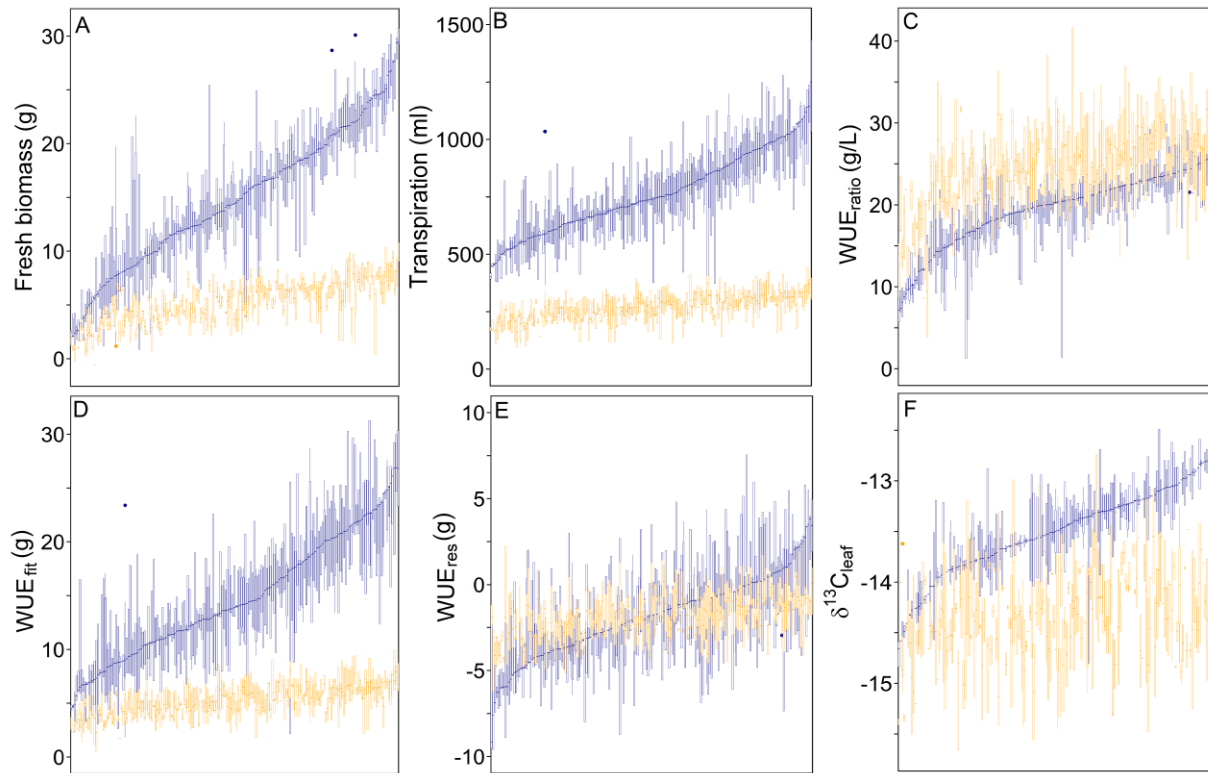
Trait	Proportional variance			H^2		
	Genotype	Treatment	G x Treatment	Both treatments	Well-watered treatment	Water-limited treatment
Fresh biomass	0.13	0.64	0.14	0.60	0.77	0.65
Transpiration	0.04	0.86	0.04	0.56	0.61	0.34
$\text{WUE}_{\text{ratio}}$	0.34	0.32	0.01	0.83	0.62	0.37
WUE_{fit}	0.07	0.69	0.11	0.48	0.63	0.33
WUE_{res}	0.24	0.0	0.0	0.62	0.23	0.27
$\delta^{13}\text{C}_{\text{leaf}}$	0.09	0.55	0.01	0.45	0.49	0.04

Table 3. QTL found across all traits in both treatments. Filled cells represent QTL, and the number in the cell is the proportion of additive variance explained (%) by the QTL, which can have a positive or negative effect on the trait.

Trait	Treatment	Genomic position of QTL							
		2@96	3@48	5@109	6@61	7@51	7@99	9@34	9@126
Fresh biomass	well-watered	15.47	5.5	-6.22	4.8		-8.24	11.88	
	water-limited	10.3		-8.52			-5.19	20.37	9.5
Transpiration	well-watered	13.45		-9.92	4.8	-5.14	-10.54	16.49	
	water-limited	8.15					-11.54	21.84	
WUE _{fit}	well-watered	13.45		-9.92		-5.14	-10.53	16.49	
	water-limited	8.15					-11.54	21.84	
$\delta^{13}\text{C}$	well-watered					-6.49	-8.2	14.52	
	water-limited								
WUE _{ratio}	well-watered	20.27		-10.67				6.45	
	water-limited	13.56		-11.77				9.15	7.50
WUE _{res}	well-watered	7.96		-13.55					
	water-limited	11.01		-22.72					

Table 4: $\delta^{13}\text{C}_{\text{leaf}}$ and the regression slope of the relationship between fresh biomass and transpiration at the allele class level. The slopes are Model II regression (standard major axis). The allele class 'ABA' was not present in this RIL population. These slopes \pm SEM are from the relationship found in figure 3A and are also plotted against $\delta^{13}\text{C}_{\text{leaf}} \pm$ SEM in figure 3B.

Allele class	Well-watered treatment				Water-limited treatment			
	Slope	R ²	P	$\delta^{13}\text{C}_{\text{leaf}}$	Slope	R ²	P	$\delta^{13}\text{C}_{\text{leaf}}$
AAA	36.8 \pm 3.7	0.85	< 0.0001	-13.7 \pm 0.08	45.4 \pm 3.9	0.87	< 0.0001	-14.36 \pm 0.07
AAB	35.1 \pm 2.3	0.78	< 0.0001	-13.15 \pm 0.04	35.7 \pm 3.2	0.59	< 0.0001	-14.28 \pm 0.05
ABA	Not present in RIL population							
ABB	38.4 \pm 15.9	0.49	0.19	-13.8 \pm 0.07	31.1 \pm 16.5	0.16	0.51	-14.24 \pm 0.16
BBB	61.1 \pm 6.6	0.91	< 0.0001	-13.91 \pm 0.07	53.1 \pm 9.2	0.59	0.009	-14.42 \pm .13
BBA	56.7 \pm 11.2	0.81	0.006	-14.0 \pm 0.09	62.7 \pm 8.8	0.90	0.001	-14.60 \pm 0.10
BAB	35.1 \pm 1.7	0.86	< 0.0001	-13.44 \pm 0.04	39.6 \pm 2.5	0.76	< 0.0001	-14.23 \pm 0.05
BAA	47.9 \pm 4.0	0.86	< 0.0001	-13.79 \pm 0.09	42.6 \pm 7.8	0.32	0.006	-14.43 \pm 0.08

Correlation with $\delta^{13}\text{C}_{\text{leaf}}$ (r)

Treatment	Fresh biomass	Transpiration	$\text{WUE}_{\text{ratio}}$	WUE_{fit}	WUE_{res}
Well-watered	0.30	0.33	0.20	0.33	-0.03
Water-limited	-0.05	-0.05	-0.06	-0.05	-0.01
Both treatments	0.51	0.60	-0.19	0.55	-0.12

Figure 1. Ordered boxplots of fresh biomass (A), transpiration (B), $\text{WUE}_{\text{ratio}}$ (C), WUE_{fit} (D), WUE_{res} (E), and $\delta^{13}\text{C}_{\text{leaf}}$ (F). All traits were measured on day 25 at peak growth, except $\delta^{13}\text{C}_{\text{leaf}}$, which $\delta^{13}\text{C}_{\text{leaf}}$ measured on leaves collected at the end of the experiment on day 34. Treatment effect was significant for all traits except WUE_{res} . The table below shows the correlation coefficients of each trait with $\delta^{13}\text{C}_{\text{leaf}}$.

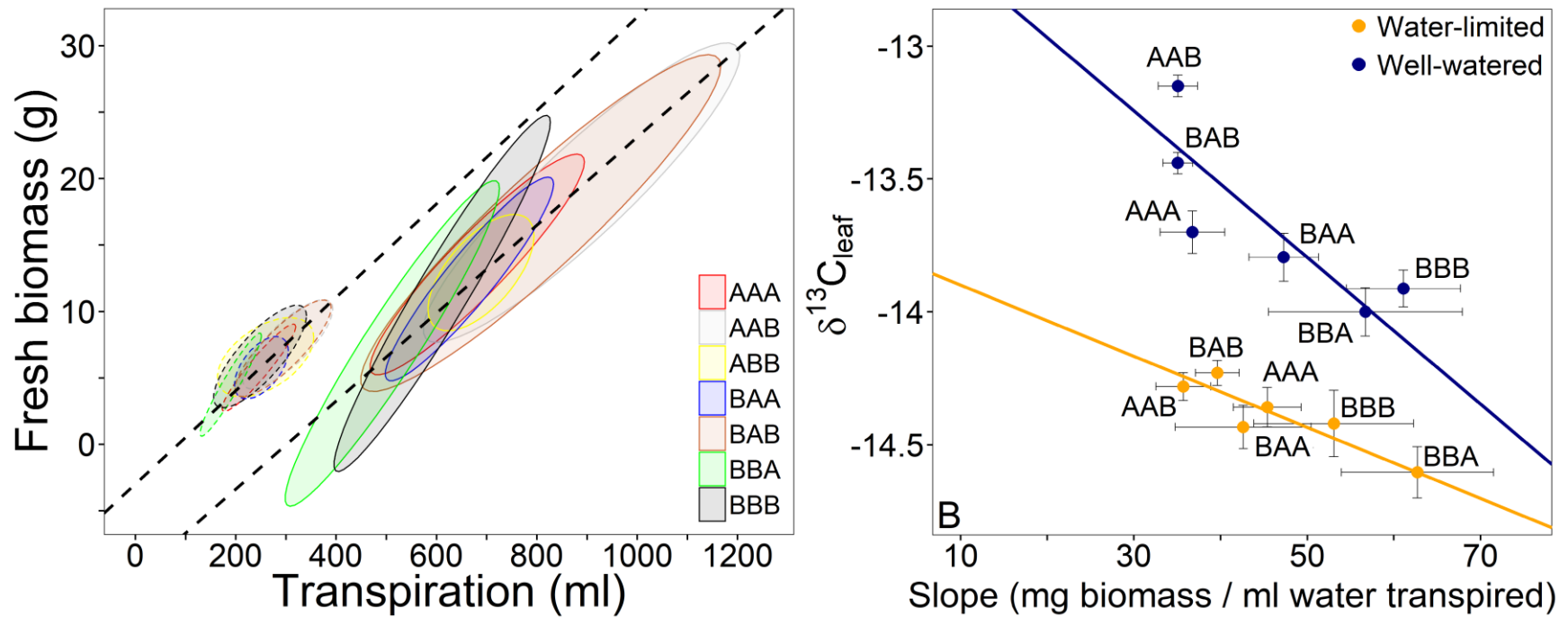


Figure 2. The effect of allele class on fresh biomass, transpiration and $\delta^{13}\text{C}_{\text{leaf}}$. In panel A, QTL 7@51, 7@99, and 9@34 were combined to produce seven allele classes where the first letter represents the allele at QTL 7@51, the second letter represents the allele at QTL 7@99, and the third letter represents the allele at QTL 9@34. The letter 'A' represents the allele from the A10 parental accession (*Setaria viridis*), and 'B' represents the allele from the B100 parental accession (*Setaria italica*). Ellipses represent 95 % confidence intervals for the relationship of fresh biomass and transpiration, and the slope of this relationship for each allele class was significant, except for allele class 'ABB' ($P < 0.0001$). In panel B, $\delta^{13}\text{C}_{\text{leaf}} \pm \text{SEM}$ is regressed against the slope of relationship $\pm \text{SE}$ in panel A, excluding the non-significant slope for 'ABB'. The slope is the water use efficiency for an entire allele class. The regression for $\delta^{13}\text{C}_{\text{leaf}}$ versus slope was significant in the well-watered ($\delta^{13}\text{C}_{\text{leaf}} = -0.027 \text{ slope} - 12.48$; $R^2 = 0.71$; $P < 0.05$) and in the water-limited treatments ($\delta^{13}\text{C}_{\text{leaf}} = -0.012 \text{ slope} - 13.82$; $R^2 = 0.80$, $P < 0.05$).

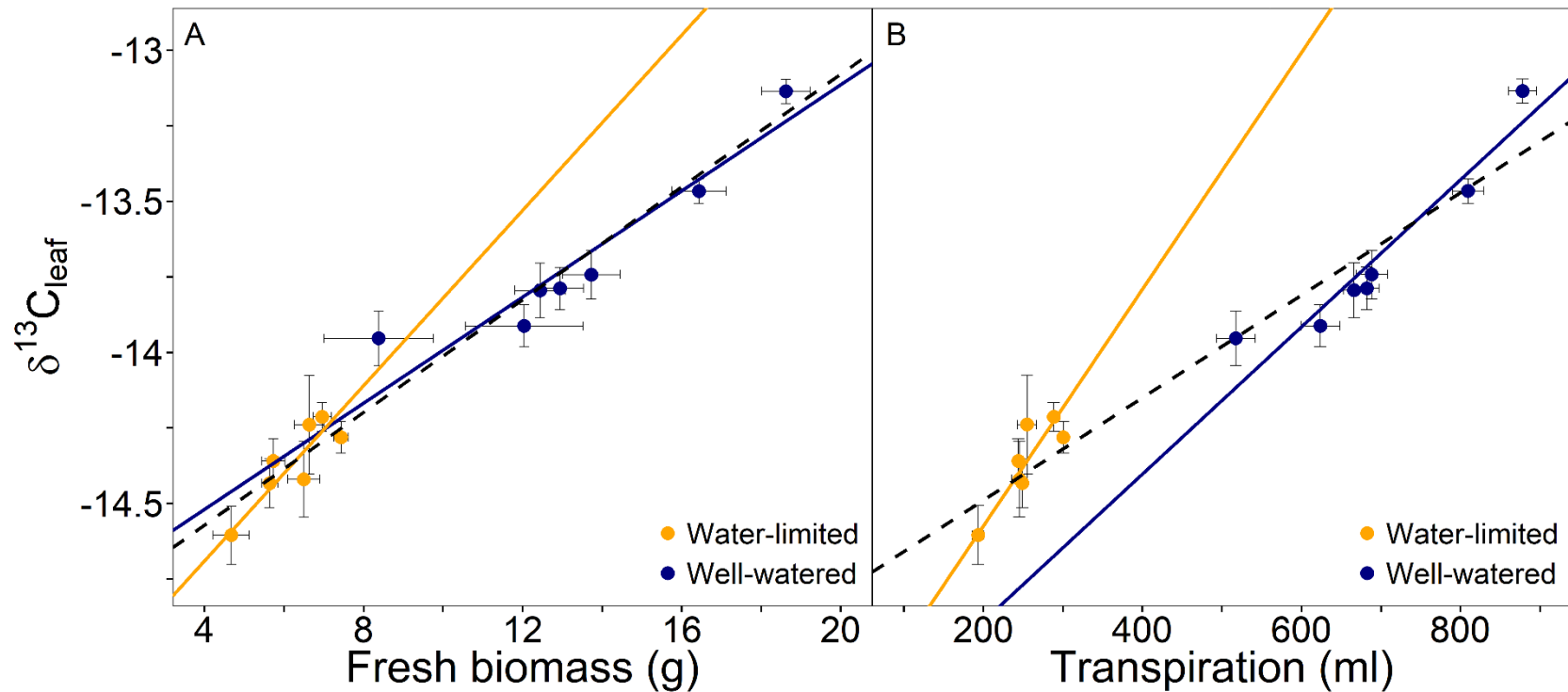


Figure 3. The effect of allele class on the relationship of fresh biomass and transpiration with $\delta^{13}\text{C}_{\text{leaf}}$. Like in Fig. 2, QTL 7@51, 7@99 and 9@34 were combined to produce seven allele classes. In panel A, the mean $\delta^{13}\text{C}_{\text{leaf}}$ was regressed against fresh biomass for both treatments separately and both combined ($\delta^{13}\text{C}_{\text{leaf}} = 0.088$ fresh biomass $- 14.87$; $R^2 = 0.88$; $P = 0.002$ for well-watered; $\delta^{13}\text{C}_{\text{leaf}} = 0.145$ fresh biomass $- 15.14$; $R^2 = 0.75$; $P = 0.01$ for water-limited; $\delta^{13}\text{C}_{\text{leaf}} = 0.093$ fresh biomass $- 14.94$; $R^2 = 0.95$; $P < 0.0001$ for both treatments). In panel B, $\delta^{13}\text{C}_{\text{leaf}}$ is regressed against transpiration for both treatments and both treatments combined ($\delta^{13}\text{C}_{\text{leaf}} = 0.0024$ transpiration $- 15.38$; $R^2 = 0.90$; $P = 0.001$ for well-watered; $\delta^{13}\text{C}_{\text{leaf}} = 0.0039$ transpiration $- 15.36$; $R^2 = 0.75$; $P = 0.01$ for water-limited; $\delta^{13}\text{C}_{\text{leaf}} = 0.0017$ transpiration $- 14.83$; $R^2 = 0.93$; $P < 0.0001$ for both treatments). Regression of both treatments together is identified by black, dashed line.

Supplemental Figures

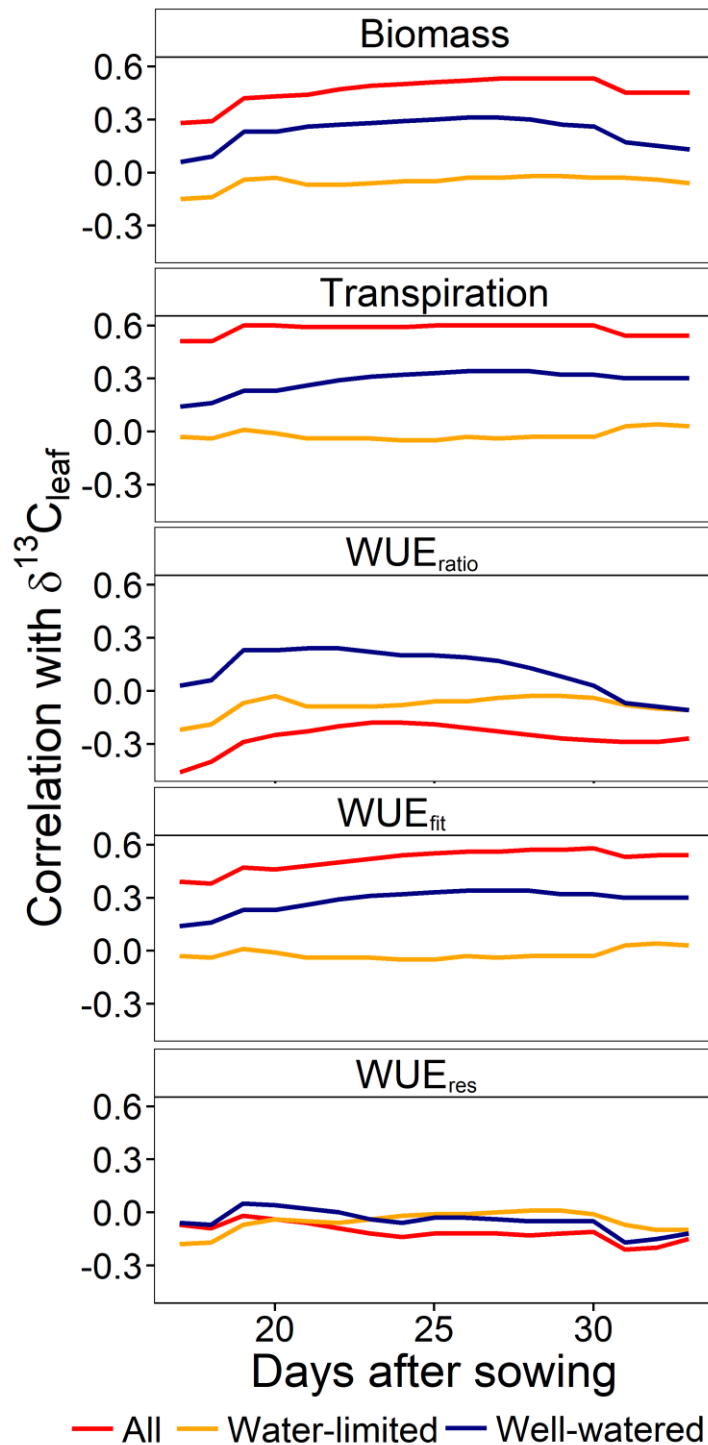


Figure S1. Correlation of fresh biomass, transpiration, WUE_{ratio} , WUE_{fit} , and WUE_{res} with $\delta^{13}C_{leaf}$ through the course of the experiment. Fresh biomass, transpiration, WUE_{fit} , and WUE_{res} were from Feldman, *et al.* (42).

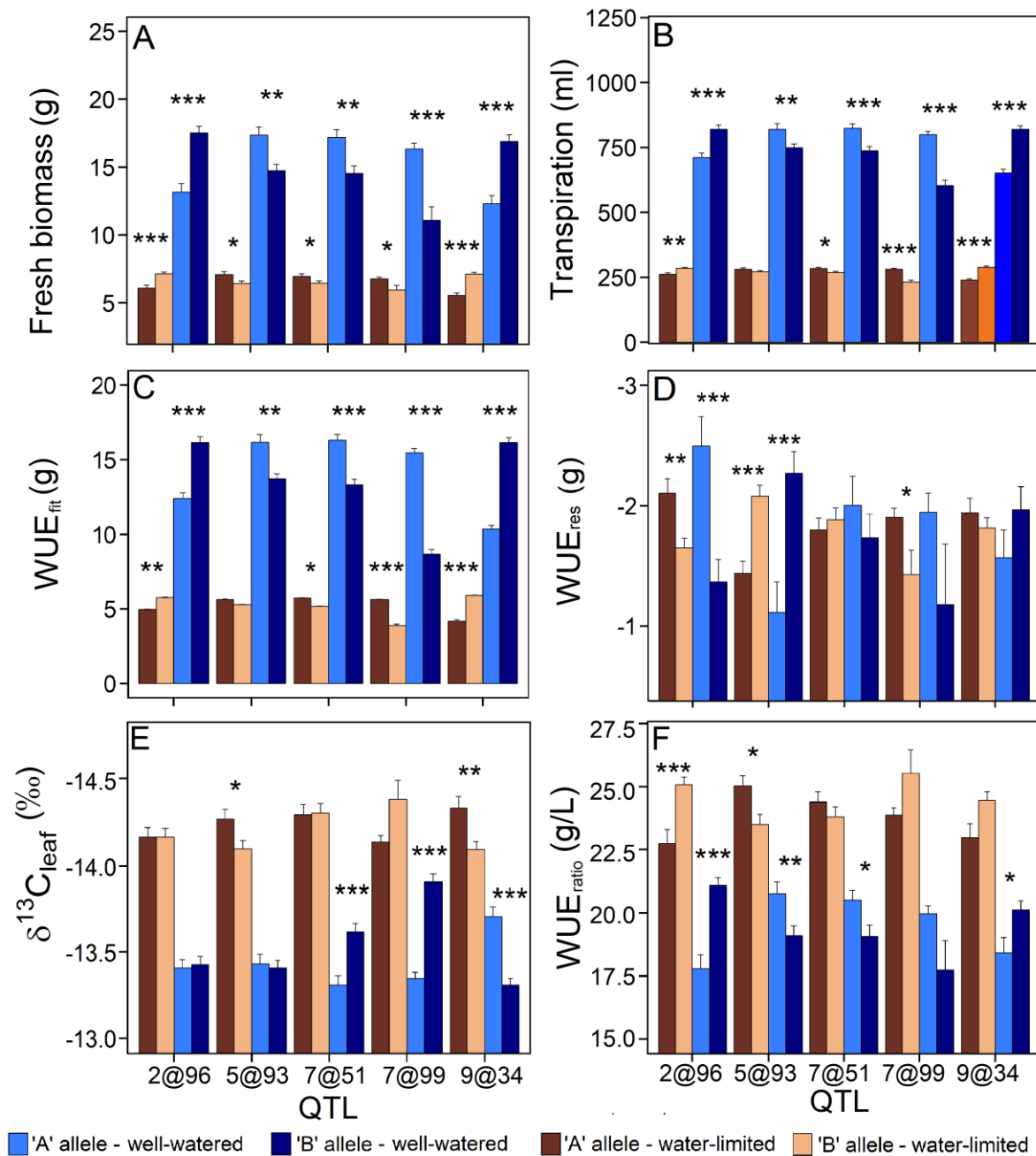


Figure S2. Effect of alleles of the five principal QTL identified for fresh biomass (A), transpiration (B), WUE_{fit} (C), WUE_{res} (D), and $\delta^{13}\text{C}_{\text{leaf}}$ (E), and WUE_{ratio} (F). 'A' represents the allele from the A10 parental line (*Setaria viridis*), and 'B' represents the allele from the B100 parental line (*Setaria italica*). Level of significance is denoted as the following: *, **, *** represent P < 0.05, 0.01, 0.0001, respectively.

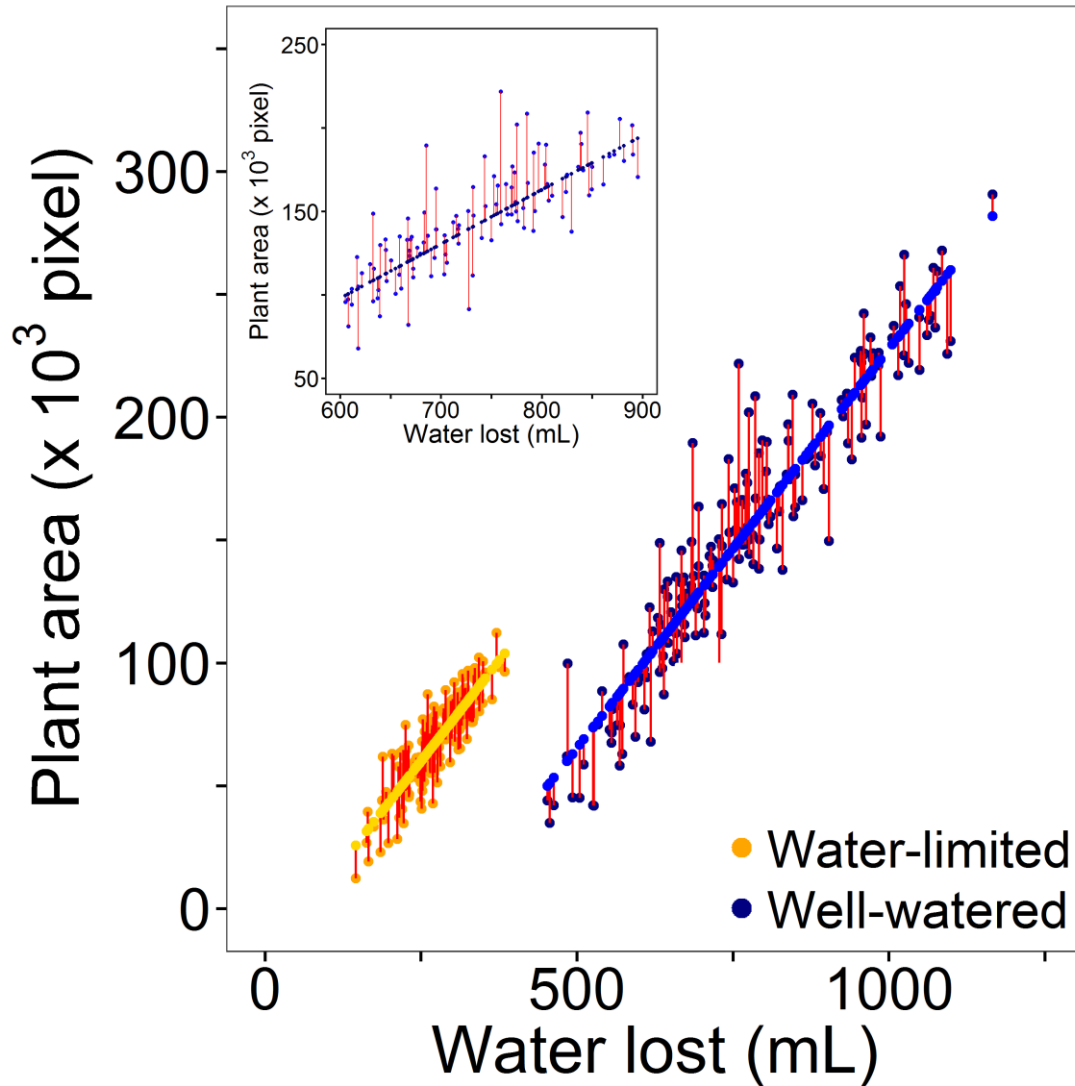


Figure S3. Method of measuring the residuals of an ordinary least squares (OLS) linear regression for plant size versus whole-plant transpiration on day 25. Blue and yellow points represent the model fit (WUE_{fit}) in each treatment. Red vertical lines connecting points to the model fit (WUE_{res}) represent residuals. Inset zooms in to the range from 600-900 ml of cumulative transpiration for the well-watered treatment only.

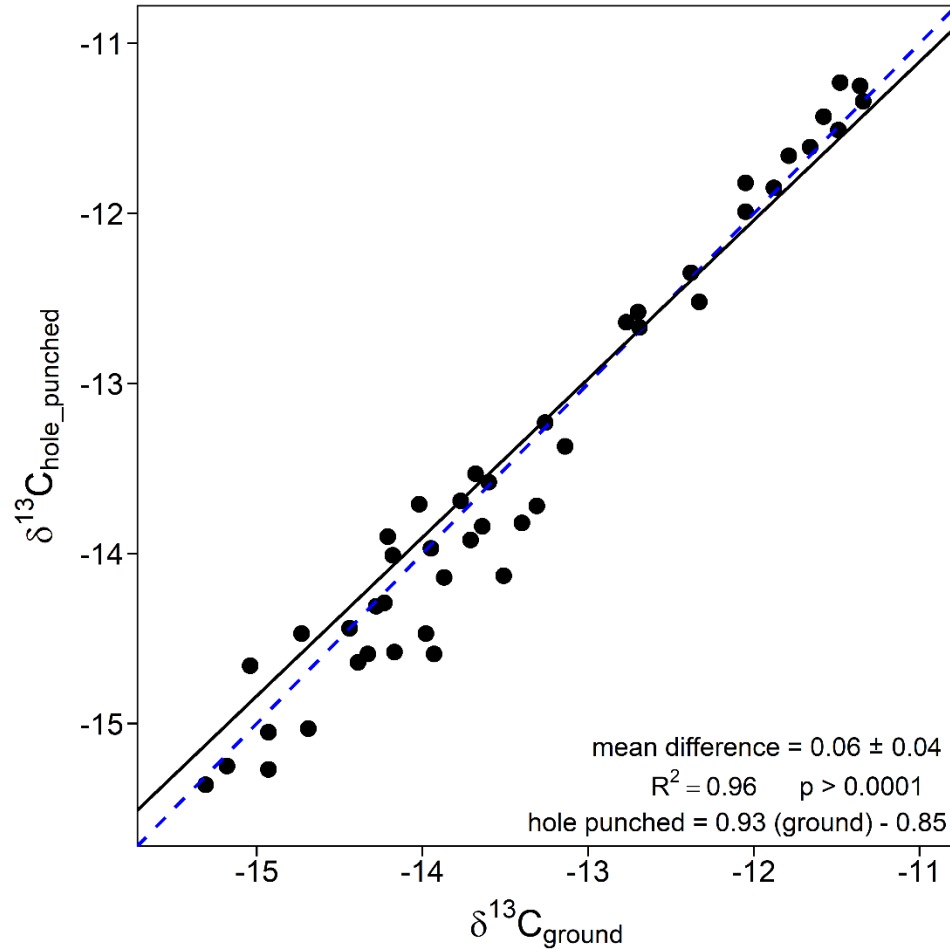


Figure S4. Comparison of the two sampling methods for $\delta^{13}\text{C}_{\text{leaf}}$. Principal method used to sample for $\delta^{13}\text{C}_{\text{leaf}}$ was hole punching, which provided similar values to grinding the entire leaf and weighing out a subsample. Solid, black line represents the regression line. The dashed, blue represents 1:1 line.

References Cited

1. Morison JI, Baker NR, Mullineaux PM, & Davies WJ (2008) Improving water use in crop production. *Philos Trans R Soc Lond B Biol Sci* 363(1491):639-658.
2. Passioura J (1977) Grain yield, harvest index, and water use of wheat. *J Aust Inst Agric Sci*.
3. Sinclair TR, Tanner C, & Bennett J (1984) Water-use efficiency in crop production. *Bioscience* 34(1):36-40.
4. Vadez V, Kholova J, Medina S, Kakker A, & Anderberg H (2014) Transpiration efficiency: new insights into an old story. *J Exp Bot* 65(21):6141-6153.
5. Bierhuizen J & Slatyer R (1965) Effect of atmospheric concentration of water vapour and CO₂ in determining transpiration-photosynthesis relationships of cotton leaves. *Agricultural Meteorology* 2(4):259-270.
6. Medrano H, *et al.* (2015) Improving water use efficiency of vineyards in semi-arid regions. A review. *Agronomy for Sustainable Development* 35(2):499-517.
7. Deng X-P, Shan L, Zhang H, & Turner NC (2006) Improving agricultural water use efficiency in arid and semiarid areas of China. *Agricultural Water Management* 80(1):23-40.
8. Westgate M, Forcella F, Reicosky D, & Somsen J (1997) Rapid canopy closure for maize production in the northern US corn belt: radiation-use efficiency and grain yield. *Field Crops Res* 49(2-3):249-258.
9. Richards RA, Hunt JR, Kirkegaard JA, & Passioura JB (2014) Yield improvement and adaptation of wheat to water-limited environments in Australia—a case study. *Crop and Pasture Science* 65(7):676-689.
10. Farquhar GD, Hubick KT, Condon AG, & Richards RA (1989) Carbon isotope fractionation and plant water-use efficiency. *Stable isotopes in ecological research*, (Springer), pp 21-40.
11. Farquhar GD, Ehleringer JR, & Hubick KT (1989) Carbon isotope discrimination and photosynthesis. *Annu Rev Plant Biol* 40(1):503-537.
12. Farquhar GD & Richards RA (1984) Isotopic composition of plant carbon correlates with water-use efficiency of wheat genotypes. *Funct Plant Biol* 11(6):539-552.
13. Coupel-Ledru A, *et al.* (2016) Reduced nighttime transpiration is a relevant breeding target for high water-use efficiency in grapevine. *Proc Natl Acad Sci U S A* 113(32):8963-8968.
14. Flexas J, *et al.* (2010) Improving water use efficiency in grapevines: potential physiological targets for biotechnological improvement. *Australian Journal of Grape and Wine Research* 16:106-121.
15. Condon AG, Richards RA, Rebetzke GJ, & Farquhar GD (2002) Improving intrinsic water-use efficiency and crop yield. *Crop Sci* 42(1):122-131.
16. Ghannoum O (2016) How can we breed for more water use-efficient sugarcane? *J Exp Bot* 67(3):557-559.
17. Jackson P, *et al.* (2015) Genetic variation in transpiration efficiency and relationships between whole plant and leaf gas exchange measurements in *Saccharum* spp. and related germplasm. *J Exp Bot* 67(3):861-871.
18. Condon AG, Richards RA, Rebetzke GJ, & Farquhar GD (2004) Breeding for high water-use efficiency. *J Exp Bot* 55(407):2447-2460.

19. Condon AG, Richards RA, & Farquhar GD (1987) Carbon isotope discrimination is positively correlated with grain yield and dry matter production in field-grown wheat. *Crop Sci* 27(5):996-1001.
20. Farquhar GD (1983) On the nature of carbon isotope discrimination in C₄ species. *Funct Plant Biol* 10(2):205-226.
21. Cabrera-Bosquet L, Molero G, Nogués S, & Araus JL (2009) Water and nitrogen conditions affect the relationships of $\Delta^{13}\text{C}$ and $\Delta^{18}\text{O}$ to gas exchange and growth in durum wheat. *J Exp Bot* 60(6):1633-1644.
22. Cabrera-Bosquet L, Albrizio R, Nogués S, & Araus JL (2011) Dual $\Delta^{13}\text{C}/\delta^{18}\text{O}$ response to water and nitrogen availability and its relationship with yield in field-grown durum wheat. *Plant Cell Environ* 34(3):418-433.
23. Elazab A, Molero G, Serret MD, & Araus JL (2012) Root traits and $\delta^{13}\text{C}$ and $\delta^{18}\text{O}$ of durum wheat under different water regimes. *Funct Plant Biol* 39(5):379-393.
24. Yousfi S, Serret MD, Márquez AJ, Voltas J, & Araus JL (2012) Combined use of $\delta^{13}\text{C}$, $\delta^{18}\text{O}$ and $\delta^{15}\text{N}$ tracks nitrogen metabolism and genotypic adaptation of durum wheat to salinity and water deficit. *New Phytol* 194(1):230-244.
25. Araus JL, Cabrera-Bosquet L, Serret MD, Bort J, & Nieto-Taladriz MT (2013) Comparative performance of $\delta^{13}\text{C}$, $\delta^{18}\text{O}$ and $\delta^{15}\text{N}$ for phenotyping durum wheat adaptation to a dryland environment. *Funct Plant Biol* 40(6):595-608.
26. Cabrera-Bosquet L, Crossa J, von Zitzewitz J, Serret MD, & Luis Araus J (2012) High-throughput phenotyping and genomic selection: The frontiers of crop breeding converge. *J Integr Plant Biol* 54(5):312-320.
27. Henderson S, Von Caemmerer S, Farquhar GD, Wade L, & Hammer G (1998) Correlation between carbon isotope discrimination and transpiration efficiency in lines of the C₄ species *Sorghum bicolor* in the glasshouse and the field. *Funct Plant Biol* 25(1):111-123.
28. Cabrera-Bosquet L, Sánchez C, & Araus JL (2009) Oxygen isotope enrichment ($\Delta^{18}\text{O}$) reflects yield potential and drought resistance in maize. *Plant Cell Environ* 32(11):1487-1499.
29. Ellsworth PZ, Ellsworth PV, & Cousins A (2017) Leaf oxygen and carbon isotopic signatures reflect drought resistance and water use efficiency in the C₄ grasses *Setaria viridis* and *Setaria italica*. *J Exp Bot*.
30. Xu L-K & Hsiao TC (2004) Predicted versus measured photosynthetic water-use efficiency of crop stands under dynamically changing field environments. *J Exp Bot* 55(407):2395-2411.
31. Medrano H, *et al.* (2015) From leaf to whole-plant water use efficiency (WUE) in complex canopies: limitations of leaf WUE as a selection target. *The Crop Journal* 3(3):220-228.
32. Tomás M, *et al.* (2014) Variability of water use efficiency in grapevines. *Environ Exper Bot* 103:148-157.
33. Tomás M, *et al.* (2012) Water-use efficiency in grapevine cultivars grown under controlled conditions: effects of water stress at the leaf and whole-plant level. *Australian Journal of Grape and Wine Research* 18(2):164-172.
34. Tarara JM, Peña JEP, Keller M, Schreiner RP, & Smithyman RP (2011) Net carbon exchange in grapevine canopies responds rapidly to timing and extent of regulated deficit irrigation. *Funct Plant Biol* 38(5):386-400.

35. Terashima I & Hikosaka K (1995) Comparative ecophysiology of leaf and canopy photosynthesis. *Plant Cell Environ* 18(10):1111-1128.
36. Poni S, *et al.* (2009) Performance and water-use efficiency (single-leaf vs. whole-canopy) of well-watered and half-stressed split-root Lambrusco grapevines grown in Po Valley (Italy). *Agric Ecosyst Environ* 129(1):97-106.
37. Gibberd MR, Walker RR, Blackmore DH, & Condon AG (2001) Transpiration efficiency and carbon-isotope discrimination of grapevines grown under well-watered conditions in either glasshouse or vineyard. *Australian Journal of Grape and Wine Research* 7(3):110-117.
38. Chaves MM, *et al.* (2007) Deficit irrigation in grapevine improves water-use efficiency while controlling vigour and production quality. *Ann Appl Biol* 150(2):237-252.
39. Cerasoli S, *et al.* (2004) Carbon and nitrogen winter storage and remobilisation during seasonal flush growth in two-year-old cork oak (*Quercus suber* L.) saplings. *Ann For Sci* 61:721-729.
40. Pinto H, Powell JR, Sharwood RE, Tissue DT, & Ghannoum O (2015) Variations in nitrogen use efficiency reflect the biochemical subtype while variations in water use efficiency reflect the evolutionary lineage of C₄ grasses at inter-glacial CO₂. *Plant Cell Environ* 39(3):514-526.
41. Fahlgren N, *et al.* (2015) A versatile phenotyping system and analytics platform reveals diverse temporal responses to water availability in *Setaria*. *Molecular Plant* 8(10):1520-1535.
42. Feldman MJ, *et al.* (2018) Trait components of whole plant water use efficiency are defined by unique, environmentally responsive genetic signatures in the model C₄ grass *Setaria*. *bioRxiv*.
43. Xu Y, *et al.* (2009) Leaf-level water use efficiency determined by carbon isotope discrimination in rice seedlings: genetic variation associated with population structure and QTL mapping. *Theor Appl Genet* 118(6):1065-1081.
44. Takai T, Fukuta Y, Sugimoto A, Shiraiwa T, & Horie T (2006) Mapping of QTLs controlling carbon isotope discrimination in the photosynthetic system using recombinant inbred lines derived from a cross between two different rice (*Oryza sativa* L.) cultivars. *Plant Prod Sci* 9(3):271-280.
45. Takai T, *et al.* (2009) Detection of a quantitative trait locus controlling carbon isotope discrimination and its contribution to stomatal conductance in japonica rice. *Theor Appl Genet* 118(7):1401-1410.
46. Teulat B, *et al.* (2002) QTLs for grain carbon isotope discrimination in field-grown barley. *TAG Theoretical and Applied Genetics* 106(1):118-126.
47. Des Marais DL, Razzaque S, Hernandez KM, Garvin DF, & Juenger TE (2016) Quantitative trait loci associated with natural diversity in water-use efficiency and response to soil drying in *Brachypodium distachyon*. *Plant Science*.
48. Rebetzke GJ, Condon AG, Farquhar GD, Appels R, & Richards RA (2008) Quantitative trait loci for carbon isotope discrimination are repeatable across environments and wheat mapping populations. *Theor Appl Genet* 118(1):123-137.
49. Xu X, *et al.* (2008) Fine mapping a QTL for carbon isotope composition in tomato. *Theor Appl Genet* 117(2):221-233.

50. Juenger TE, *et al.* (2005) Identification and characterization of QTL underlying whole-plant physiology in *Arabidopsis thaliana*: $\delta^{13}\text{C}$, stomatal conductance and transpiration efficiency. *Plant Cell Environ* 28(6):697-708.
51. McKay JK, *et al.* (2008) Genetics of drought adaptation in *Arabidopsis thaliana* II. QTL Analysis of a new mapping population, Kas-1 \times Tsu-1. *Evolution* 62(12):3014-3026.
52. Adiredjo AL, *et al.* (2014) Genetic control of water use efficiency and leaf carbon isotope discrimination in sunflower (*Helianthus annuus* L.) subjected to two drought scenarios. *PLoS ONE* 9(7):e101218-e101218.
53. Dhanapal AP, *et al.* (2015) Genome-wide association study (GWAS) of carbon isotope ratio ($\delta^{13}\text{C}$) in diverse soybean [*Glycine max* (L.) Merr.] genotypes. *Theor Appl Genet* 128(1):73-91.
54. Saranga Y, Jiang CX, Wright R, Yakir D, & Paterson A (2004) Genetic dissection of cotton physiological responses to arid conditions and their inter-relationships with productivity. *Plant Cell Environ* 27(3):263-277.
55. Brendel O, *et al.* (2008) Quantitative trait loci controlling water use efficiency and related traits in *Quercus robur* L. *Tree Genet Genomes* 4(2):263-278.
56. Thumma BR, *et al.* (2001) Identification of causal relationships among traits related to drought resistance in *Stylosanthes scabra* using QTL analysis. *J Exp Bot* 52(355):203-214.
57. Easlon HM, *et al.* (2014) The physiological basis for genetic variation in water use efficiency and carbon isotope composition in *Arabidopsis thaliana*. *Photosynth Res* 119(1-2):119-129.
58. Masle J, Gilmore SR, & Farquhar GD (2005) The ERECTA gene regulates plant transpiration efficiency in *Arabidopsis*. *Nature* 436(7052):866-870.
59. Gresset S, *et al.* (2014) Stable carbon isotope discrimination is under genetic control in the C₄ species maize with several genomic regions influencing trait expression. *Plant Physiol* 164(1):131-143.
60. Wang Z, Devos K, Liu C, Wang R, & Gale M (1998) Construction of RFLP-based maps of foxtail millet, *Setaria italica* (L.) P. Beauv. *Theor Appl Genet* 96(1):31-36.
61. Devos KM, Wang Z, Beales J, Sasaki T, & Gale M (1998) Comparative genetic maps of foxtail millet (*Setaria italica*) and rice (*Oryza sativa*). *Theor Appl Genet* 96(1):63-68.
62. von Caemmerer S, Ghannoum O, Pengelly JJ, & Cousins AB (2014) Carbon isotope discrimination as a tool to explore C₄ photosynthesis. *J Exp Bot* 65(13):3459-3470.
63. Cabrera-Bosquet L, Sánchez C, & Araus JL (2009) How yield relates to ash content, $\Delta^{13}\text{C}$ and $\Delta^{18}\text{O}$ in maize grown under different water regimes. *Ann Bot* 104(6):1207-1216.
64. Monneveux P, Sheshshayee MS, Akhter J, & Ribaut J-M (2007) Using carbon isotope discrimination to select maize (*Zea mays* L.) inbred lines and hybrids for drought tolerance. *Plant Science* 173(4):390-396.
65. Lambrides C, Chapman S, & Shorter R (2004) Genetic variation for carbon isotope discrimination in sunflower. *Crop Sci* 44(5):1642-1653.
66. Feldman MJ, *et al.* (2017) Time dependent genetic analysis links field and controlled environment phenotypes in the model C₄ grass *Setaria*. *PLoS Genet* 13(6):e1006841.
67. Nakhforoosh A, Bodewein T, Fiorani F, & Bodner G (2016) Identification of water use strategies at early growth stages in durum wheat from shoot phenotyping and physiological measurements. *Frontiers in Plant Science* 7.

68. Lopez JR, *et al.* (2017) QTLs associated with crown root angle, stomatal conductance, and maturity in Sorghum. *Plant Genome*.
69. Impa S, *et al.* (2005) Carbon isotope discrimination accurately reflects variability in WUE measured at a whole plant level in rice. *Crop Sci* 45(6):2517-2522.
70. Flanagan LB & Farquhar GD (2014) Variation in the carbon and oxygen isotope composition of plant biomass and its relationship to water-use efficiency at the leaf-and ecosystem-scales in a northern Great Plains grassland. *Plant Cell Environ* 37(2):425-438.
71. Doust AN, Kellogg EA, Devos KM, & Bennetzen JL (2009) Foxtail millet: a sequence-driven grass model system. *Plant Physiol* 149(1):137-141.
72. Brenna JT, Corso TN, Tobias HJ, & Caimi RJ (1997) High-precision continuous-flow isotope ratio mass spectrometry. *Mass Spectrometry Reviews* 16(5):227-258.
73. Qi H, Coplen TB, Geilmann H, Brand WA, & Böhlke JK (2003) Two new organic reference materials for $\delta^{13}\text{C}$ and $\delta^{15}\text{N}$ measurements and a new value for the $\delta^{13}\text{C}$ of NBS 22 oil. *Rapid Commun Mass Spectrom* 17(22):2483-2487.
74. Santrock J, Studley SA, & Hayes JM (1985) Isotopic analyses based on the mass spectra of carbon dioxide. *Analytical Chemistry* 57(7):1444-1448.
75. R_Core_Team (2013) R: A language and environment for statistical computing (Vienna, Austria).
76. Seibt U, Rajabi A, Griffiths H, & Berry JA (2008) Carbon isotopes and water use efficiency: sense and sensitivity. *Oecologia* 155(3):441-454.
77. Ubierna N, Sun W, & Cousins AB (2011) The efficiency of C₄ photosynthesis under low light conditions: assumptions and calculations with CO₂ isotope discrimination. *J Exp Bot* 62(9):3119-3134.
78. Sun W, Ubierna N, Ma J-Y, & Cousins AB (2012) The influence of light quality on C₄ photosynthesis under steady-state conditions in *Zea mays* and *Miscanthus × giganteus*: changes in rates of photosynthesis but not the efficiency of the CO₂ concentrating mechanism. *Plant Cell Environ* 35(5):982-993.
79. Kromdijk J, Ubierna N, Cousins AB, & Griffiths H (2014) Bundle-sheath leakiness in C₄ photosynthesis: a careful balancing act between CO₂ concentration and assimilation. *J Exp Bot* 65(13):3443-3457.
80. Sage RF (2014) Stopping the leaks: new insights into C₄ photosynthesis at low light. *Plant Cell Environ* 37(5):1037-1041.
81. Kubásek J, Šetlík J, Dwyer S, & Šantrůček J (2007) Light and growth temperature alter carbon isotope discrimination and estimated bundle sheath leakiness in C₄ grasses and dicots. *Photosynth Res* 91(1):47-58.
82. Henderson SA, Caemmerer SV, & Farquhar GD (1992) Short-term measurements of carbon isotope discrimination in several C₄ species. *Funct Plant Biol* 19(3):263-285.
83. Mook WG, Bommerson JC, & Staverman WH (1974) Carbon isotope fractionation between dissolved bicarbonate and gaseous carbon dioxide. *Earth Planet Sci Lett* 22(2):169-176.