

1 **Part 1 RCs**

2 **RC1:** '[Comment on egusphere-2023-1522](#)', Anonymous Referee #1, 29 Aug 2023 [reply](#)
3 Quick et al. present a study of tree water use under elevated CO₂ in a FACE
4 experiment. BIFoR is a unique FACE experiment in that it is located in an old growth
5 forest and could potentially offer better insight into the future of global forests than
6 previous FACE experiments in young plantation forests. The question of plant-water
7 relations under high CO₂ is a critical one and includes processes that are often
8 uncertain so data such as those presented in this paper are extremely valuable.

9 While it is obvious that a lot of effort has gone into data collection, the paper is very
10 poorly written, with little focus or in depth analysis of what was measured. While two
11 hypotheses are raised in the introduction, these neither follow from the rest of the
12 introduction nor are they followed through in the rest of the paper. The statistical
13 analysis is lacking and it is hard to understand what the conclusions of this study are. I
14 hope that my comments can help improve the manuscript so that this very valuable data
15 can be published.

16 **RC1 Major comments:**

17 **RC1 Major Comment 1:**

- 18 • Abstract - this is very detailed and it is unclear what the focus of the paper is. Details
19 of measurements do not generally belong in an abstract and distract from the main
20 point of the paper.

21 **RC1 Major Comment 2:**

- 22 • Introduction - this starts very abruptly and the different sections feel very disjointed,
23 without a clear storyline. In general, paper introductions do not need subsections and
24 removing the headers and joining the paragraphs logically might help the reader.

25 **RC1 Major Comment 3:**

- 26 • Methods - there are a lot of details on the measurements at the site, perhaps too
27 many for the main body of the paper, but little to no details on the statistical analysis
28 of the data. From the results section it is apparent that some linear regressions have
29 been performed, although these are not documented in the methods and it is unclear
30 how and why these particular analyses were chosen. It is unclear if the data have
31 been tested for normality or any other assumptions of the methods employed.

32 **RC1 Major Comment 4:**

- 33 • Statistical analyses - the study contains only very basic stats. There is no attempt
34 made to test whether any of the observed differences are significant or not. Most
35 importantly, since ecological processes are complex and many factors contribute to
36 water use, as highlighted by the authors themselves, it is unlikely that a simple linear
37 regression is the best model to use. Some sort of mixed effect or general linear
38 model that can take into account the multiple factors involved might give a better
39 picture of the response of tree water use to CO₂.

40 **RC1 Major Comment 5:**

- 41 • Initial conditions - this is the case with any manipulative experiment, but one
42 always has to wonder about the effect of initial conditions in each treatment on
43 the actual treatment effect. As a minimum, initial conditions should be shown and
44 discussed. Initial conditions could also be included in any more complex
45 statistical model involved.

46 **RC1 Major Comment 6:**

- 47 • Discussion & interpretation - this is almost entirely lacking. The last two sections
48 are in some part a discussion, but they are very superficial, do not discuss the
49 implications of the findings or any limitations of the study.

50 **RC1 Minor comments:**

51 **RC1 Minor Comment 1:**

52 L 14 “diurnal (i.e. daylight)” maybe use daytime to avoid confusion?

53 **RC1 Minor Comment 2:**

54 L 47 This is a rather abrupt start to the introduction - why large trees?

55 **RC1 Minor Comment 3:**

56 L 50 unclear what the relaxed xylem is

57 **RC1 Minor Comment 4:**

58 L 60 wet deposition of?

59 **RC1 Minor Comment 5:**

60 L 71 photosynthesis is the name of the process even when the pigments are not green.

61 **RC1 Minor Comment 6:**

62 L 71 slightly odd references to model predictions, especially Guerreri et al. there is a lot
63 of literature out there on model predictions of CO₂ fertilization effect.

64 **RC1 Minor Comment 7:**

65 L 110 Do you mean drought?

66 **RC1 Minor Comment 8:**

67 L114 What is meant here by location - geographical coordinates, altitude, slope?

68 **RC1 Minor Comment 9:**

69 L 141 This sentence is very convoluted and it is unclear what separation refers to

70 **RC1 Minor Comment 10:**

71 L 156 monthly distributions of what? Significant differences between what and what?

72 **RC1 Minor Comment 11:**

73 L 182 is 1840 the plantation year?

74 **RC1 Minor Comment 12:**

75 L 318 It is unclear how these characteristics were used

76 **RC1 Minor Comment 13:**

77 L 335 I'm not sure this section and level of detail are necessary

78 **RC1 Minor Comment 14:**

79 L 477 It is unclear why there is a model fitted to each day rather than across days. It
80 would also be good to see R2 values

81 **RC1 Minor Comment 15:**

82 L 482 'smaller', 'similar' Were there any statistical tests done to see if these differences
83 were significant?

84 **RC1 Minor Comment 16:**

85 Figure 7 It is hard to tell what the difference between a) and b) and c) and d) is and why
86 these could not just be one plot.

87 **RC1 Minor Comment 17:**

88 Figures 7 and 8 missing statistical values such as R2 or p

89 **RC1 Minor Comment 18:**

90 Figure 9 Would be good to show where the statistical differences are

91 **RC1 Minor Comment 19:**

92 L 549 So what does this figure show us?

93 **RC1 Minor Comment 20:**

94 L 564 Since you have precipitation data, could you test this deduction?

95 **RC1 Minor Comment 21:**

96 L 591 this information belongs in the methods

97 **RC1 Minor Comment 22:**

98 Sections 3.4 and 3.5 are these meant to be a discussion?

99 **RC2:** ['Comment on egusphere-2023-1522'](#), Anonymous Referee #2, 12 Sep 2023 [reply](#)

100 This study evaluates the possible effects of enhanced CO₂ (eCO₂) on tree-water use,
101 the measurements were conducted at one of the FACE experiments on mature forests.
102 Here, the authors defined tree-water use as the daytime accumulated sapflow (TWU)
103 and proposed two hypotheses: H1 A detectable eCO₂ treatment effect on TWU is
104 present and H2 TWU is greater in the presence of FACE infrastructure. In my opinion
105 the hypotheses are relevant and could provide valuable information that could improve
106 our understanding on ecosystem responses to environmental change and on eCO₂
107 experimental design.

108 Unfortunately, the sections of the manuscript feel unconnected due to inclusion of
109 irrelevant information, which could easily be moved to the supplementary material. The
110 hypotheses are not properly tested, so the interpretation of the results and the
111 conclusions are mostly based on visual interpretation of the plots.

112 The data used in this study is incredibly rare and valuable, so I really hope that the
113 authors can make use of the following comments to better leverage these data.

114 **RC2 Major comments:**

115 **RC2 Major Comment 1:**

116 Abstract:

117 There is an unnecessary amount of detail here, it distracts the reader.

118 It is not clear to me how this info is going to contribute to the development of more
119 “realistic dynamic vegetation models”

120 **RC2 Major Comment 2:**

121 Introduction:

122 The sections inside the introduction feel disconnected and instead of leading the reader
123 to the research questions they feel like tangents.

124 Not a single mention to the Huber value or water use efficiency, which are key concepts
125 to connect CO₂ and water in trees.

126 **RC2 Major Comment 3:**

127 Methods:

128 Almost half of this manuscript is used to describe the methods. It is too much in my
129 opinion. If the authors feel like the figures in this section are really necessary, they could
130 be moved to the appendix section.

131 **RC2 Major Comment 4:**

132 There is no mention to any statistical test to evaluate the hypotheses.

133 **RC2 Major Comment 5:**

134 I don't really think TWU is a good metric to detect effects of eCO₂ in only 5 years of
135 experiment if the trees are 180 years old, the change of sapwood area of trees under
136 aCO₂ and eCO₂ would be negligible. However, if measurements of Huber value and
137 Leaf area index are available, TWU can be converted to transpiration, enhancing the
138 differences between treatments if any.

139 There is no connection on how soil moisture and throughfall are going to be used
140 complement the analysis.

141 **RC2 Major Comment 6:**

142 Results and Discussion

143 There is almost no discussion at all. The results are not put in context, it is hard to tell
144 how the reported values compare to other studies. A relative change between aCO₂ and
145 eCO₂ would be very useful here.

146 **6a:** It is well known that sap flow escalates with sapwood area (proportional to R_b), and
147 R_b to A_c, I don't really see novelty on reporting linear models on these relationships.
148 How these regressions help with the evaluation of the hypotheses?

149 **6b:** All the interpretations to the results rely on visual inspection, I believe this is not
150 enough to evaluate if your hypotheses are true.

151 **6c:** It is very interesting to see here interception, throughfall, and soil moisture
152 measurements, however these results appear disconnected from the other results.

153 **6d:** The word "significant" is used to described key results of the study, however there is
154 no statistical test associated.

155 **6e:** You state that hypothesis 2 is true, however this is hard to tell without proper stats. It
156 is unlikely that the effect of the infrastructure alone is bigger than the effect of the eCO₂.
157 If that were true, we could see the combined effect of infrastructure + eCO₂ effect in the
158 eCO₂ treatment as well.

159 **RC2 Minor comments:**

160 **RC2 Minor Comment 1:**

161 L 70 "...could be beneficial for individual tree Productivity" hard to tell without
162 considering respiration and temperature increase. These ideas would be helpful to
163 contextualize better.

164 **RC2 Minor Comment 2:**

165 L 155 how did you test for significant differences?

166 **RC2 Minor Comment 3:**

167 L 335 Is this relevant?

168 **RC2 Minor Comment 4:**

169 Figure 9. unreadable

170 **RC2 Minor Comment 5:**

171 L 570 how can you tell there are” significant differences” without any test? Why do you
172 attribute similar general responses to daylength? What about temperature or VPD?

173 **RC2 Minor Comment 6:**

174 L 585 “Influence of ... and herbivory on water usage dynamics” is misleading, you just
175 point out that an attack of herbivory happened.

176 [Reply](#)

177 **Citation:** <https://doi.org/10.5194/egusphere-2023-1522-RC2>
178

179 **Part 2 AC**

180 **General (RC1, RC2):**

181 **AC re General (RC1, RC2):** *Thank you for highlighting the importance of this topic and*
182 *suggesting general and specific improvements. We recognise there are some*
183 *shortcomings in the submitted paper. We are very grateful for the efforts to improve, and*
184 *the points made, by all referees/ community to bring this paper to the required standard,*
185 *both in terms of the quality and flow of the paper, explaining the statistical analysis and*
186 *in respect of testing the hypotheses. Obtaining the data reported has occupied 6 years*
187 *of intensive field work and we are very pleased and relieved that referee RC1 agrees*
188 *that the data are of great value and that RC2 agrees that the hypotheses are relevant*
189 *and will aid understanding of ecosystem responses to environmental change and*
190 *improvement of eCO2 experimental design. The two hypotheses and objectives are*
191 *reviewed in section 3.5.*

192 **Abstract:**

193 **RC1 Major Comment 1 and RC2 Major Comment 1:**

194 **AC lines 13-41 re Major Comment 1 and RC2 Major Comment 1:** *The abstract is*
195 *being refocused on the main methodology, objectives and findings. Its redraft will be*
196 *finalised in the marked up revision.*

197 *We also make reference to where dynamic vegetation models can be improved in our*
198 *Introduction. (old LL123-127)*

199 **Introduction:**

200 **RC1 Major Comment 2 and RC2 Major Comment 2:**

201 **AC re RC1 Major Comment 2 and RC2 Major Comment 2:** *We have improved the*
202 *starting sentences and clarified the story line by reordering some of the sentences to*
203 *improve the flow and logic and moving previous section 1.2 nearer the beginning. The*
204 *aims of the paper have been clarified and sharpened. Some headers are retained as we*
205 *believe they help the reader frame the sections.*

206 *See also AC on RC2 Major Comment 5 in respect of Huber value and water use*
207 *efficiency.*

208 **Materials and Methods:**

209 **RC1 Major Comment 3 and RC2 Major Comment 3:**

210 **AC re Major Comment 3 and RC2 Major Comment 3:** *We recognise that this may be*
211 *too detailed for the reader and have moved some material to the SI (e.g. re non sap flow*
212 *instrumentation) and some re sap flow to a new Appendix.*

213 *In respect of the Figures in this Section of the paper, some of them add knowledge to*
214 *the characteristics of water flow within the xylem. We have reviewed all these figures*
215 *and explained their purpose within captions as appropriate. We moved any remaining*
216 *figures to the new Appendix C which covers detail of the sap flux methodology. (e.g.*

217 *Fig.3 is now Fig.C1). We show the remaining text for Sections 2.7 to 2.8 in part 3 of our*
218 *response as an example.*

219 *A summary of the analysis tools used is contained in Section 2.9.*

220 *Regression analysis in respect of tree size was progressed following visualisation,*
221 *facilitating normalisation of the water usage data in respect of tree parameters and*
222 *consequent comparison of data from different treatments; the linear model provided a*
223 *simple best fit for this range of data – e.g. See old LL 473&474.*

224 *There is reference to the minimum detectable sap flux using this method (old LL425-*
225 *426) showing that the effects are tree-size dependent and referring to Appendix A2,*
226 *Stage 5. The limitations and truncation effect of the sap flow transducer method is also*
227 *outlined in Appendix B. The first paragraph of this has been expanded to explain the*
228 *influence of truncation on the normality of the data. Some text has also been moved*
229 *here from the Results section.*

230 *Further response on statistical methods is outlined below.*

231 **Statistical analysis**

232 **RC1 Major Comment 4:**

233 **AC re RC1 Major Comment 4:** *Timely sharing of sap flux data from this plant-water*
234 *experiment is important to enable its use in land surface models such as JULES and*
235 *vegetation models. As a result we have restricted our statistical approach to show the*
236 *validity of the data and pointed to where this fills gaps in existing knowledge.*

237 **RC2 Major Comment 4:**

238 *There is no mention to any statistical test to evaluate the hypotheses.*

239 **AC re RC2 Major Comments 4:**

240 *We recognise there were some limitations in our pre-print in respect of statistical testing*
241 *of the hypotheses. We have now undertaken additional analyses to underpin our*
242 *findings in respect of the hypotheses, focusing on 2019-2021. A brief insertion of new*
243 *figure content and p values tables (draft below) has been added to the Results. The*
244 *Discussion will be modified to reflect the new findings.*

245 *A summary of the analysis tools used is contained in Section 2.9 – see revised section*
246 *below. Regression against tree size allowed normalisation of the water usage data in*
247 *respect of tree parameters and consequent comparison of treatments; the linear model*
248 *provided a simple best fit for this range of data – e.g. See old LL 473&474.*

249 **RC1 Major Comment 5:**

250 **AC re RC1 Major Comment 5:** *With the exception of reporting of precipitation and*
251 *throughfall data where this is available pre-treatment (sections 2.4, 2.5 and 3.4),*
252 *reporting of the initial conditions is not possible. To assist the reader some data has*
253 *been added to the SI document clarifying this point.*

254 **RC2 Major Comment 5:**

255 **AC re RC2 Major Comment 5:** *Our normalised daylight TWU is a whole-tree integrative*
256 *measure of transpiration and, as such, a measure suitable for determining the effects of*
257 *the CO₂ treatment. No other FACE studies of mature trees have had sufficient periods of*
258 *sap flux data across treatment and control replicates to investigate seasonal CO₂*
259 *treatment effects over successive years. Estimates of water usage as a function of land*
260 *area across each of the 9 arrays for all oaks in the experiment is outside the scope of*
261 *the present study.*

262 *Huber value and leaf area index for the years reported cannot be calculated directly on a*
263 *per tree basis.*

264 **Results and Discussion**

265 **RC1 Major Comment 6 and RC2 Major Comment 6:**

266 **AC re RC1 Major Comment 6 and RC2 Major Comment 6:** *Refinement of the Results*
267 *and Discussion section is being carried out and will be presented in the marked up*
268 *revision as an improvement to the pre-print.*

269 **AC re RC2 Major Comment 6 sub comments**

270 **6a:** *There is a paucity of results relating sap flow with sapwood area for mature Q. robur*
271 *of the range of DBH considered here. The relationship in this study enables*
272 *normalisation of the TWU data. This is not the only option for normalisation but, as we*
273 *include all sap flux data across all days of our treatment season April to October,*
274 *including dark and/ or rainy days, we do not use the tree-size normalisation method*
275 *adopted in a similar study (i.e. Web-FACE, reported Leuzinger and Körner, (2007))*
276 *which compared species water usage. They normalised each species by the highest tree*
277 *specific accumulated sap flux for a given period of monitoring (22 days).*

278 **6b:** *We agree that the analyses should not rely on visual qualitative observations. We*
279 *have added statistical tests to underpin our findings in respect of the hypotheses. These*
280 *are reported in a new section of the results.*

281 **6c:** *Please see response to RC2 Major Comment 5.*

282 **6d:** *We report the significance of our normalisation linear models in old LL500-507 and*
283 *LL522-530 including reference to p-values in the supporting tables. Old lines 571-572*
284 *and 643-651 use the term significant incorrectly and these paragraphs have been*
285 *reviewed and rephrased.*

286 **and 6e:** *Please see response to RC2 Major Comments 4*

287 **Minor comments RC1 and RC2 in line order:**

288 **RC1 Minor Comment 1:**

289 L 14 **AC:** *Noted. We have selected the terms most used in supporting plant-water and*
290 *assimilation literature i.e. diurnal as the antonym to nocturnal. We recognise that the*
291 *term can also be used in place of circadian, hence the clarification of definition and*
292 *usage for this paper.*

293 **RC1 Minor Comment 2:**

294 L 47 **AC:** *Noted. We have now added couple of sentences forming a better start to the*
295 *Introduction. See below.*

296 **RC1 Minor Comment 3:**

297 L 50 **AC:** *Whilst we previously referred to relaxed xylem to reflect changes in diameter*
298 *due to lack of plant water tension (i.e. due to lack of or reduced transpiration), this*
299 *sentence lacked clarity and the references had been incorrectly grouped. Above ground*
300 *both ambient air temperature and lack of transpiration affects the stem DBH and related*
301 *physical traits such as stem circumference. We intended to point to the tree anatomy*
302 *response to a variety of stressors. Sentences in pre-print (old LL49-51) have been*
303 *regrouped and redrafted. See below.*

304 **RC1 Minor Comment 4:**

305 L 60 **AC:** *We have deleted this sentence as part of refocusing content of the*
306 *Introduction.*

307 **RC2 Minor Comment 1:**

308 L 70 **AC:** *old LL71-74 have been rephrased to clarify.*

309 **RC1 Minor Comment 5:**

310 L 71 **AC:** *Noted. Changed to refer to chlorophyll.*

311 **RC1 Minor Comment 6:**

312 L 71 **AC:** *We are not discussing the CO₂ fertilization effect here, but rather the*
313 *implications of CO₂ fertilisation for water usage. Guerrieri et al. (2016) is quoted as a*
314 *recent example paper as we are concerned here with the interface of forest carbon*
315 *uptake and water use. We have changed the references to include some additions. (e.g.*
316 *Norby et al. 2016)*

317 **RC1 Minor Comment 7:**

318 L 110 **AC:** *Yes, water availability and drought are relevant. We have modified the*
319 *sentence.*

320 **RC1 Minor Comment 8:**

321 L114 **AC:** *yes this is ambiguous, we have redrafted the phrase to be more specific.*

322 **RC1 Minor Comment 9:**

323 L 141 **AC:** *Dendrometer sentences have been removed as not relevant to this paper.*

324 **RC2 Minor Comment 2:**

325 L 155 **AC:** *A summary of the analysis tools used is contained in Section 2.9.*

326 **RC1 Minor Comment 10:**

327 L 156 **AC:** *monthly distributions are of ‘...seasonal and inter-year patterns of daily water*
328 *usage by old growth oak trees...’ as stated. Significant differences sentence has been*
329 *rephrased to clarify.*

330 **RC1 Minor Comment 11:**

331 L 182 **AC:** *See also L796-798. Dendrochronology was used to date the age range of*
332 *oaks in the FACE experimental forest. It is unclear what age the trees were when they*
333 *were planted hence the reference to circa 1840.*

334 **RC1 Minor Comment 12:**

335 L 318 **AC:** *See stage 5 in Appendix Table A2 also Figure 2. The tree radius at insertion*
336 *point derived from the circumference minus the bark thickness gives R (m).*

337 **RC1 Minor Comment 13 and RC2 Minor Comment 3:**

338 L 335 **AC:** *Moved to Appendix C*

339 **RC1 Minor Comment 14:**

340 L 477 **AC:** *the model is fitted across all the days of July. See also comments on Figs 7 &*
341 *8. Seasonal water usage modelling would need to account for day length and this*
342 *modelling has not been undertaken.*

343 **RC1 Minor Comment 15:**

344 L 482 **AC:** *See L501 and Table S5, also comments on Figs 7 & 8. It would be possible*
345 *to do more extensive treatment comparisons, but these would not result in significant*
346 *differences, as discussed.*

347 **RC1 Minor Comment 16:**

348 Figure 7 **AC:** *putting all the treatments, separated by year, on one plot makes the results*
349 *difficult to visualise. b) and d) combine the regression line results for all the years*
350 *available for a particular treatment.*

351 **RC1 Minor Comment 17:**

352 Figures 7 and 8 **AC:** *p is stated in the related tables (Fig. 7 relates to Table 2 and Table*
353 *S5, Fig.8 relates to Table 3 as stated in the text).*

354 **RC1 Minor Comment 18:**

355 Figure 9 **AC:** *Noted. Please see additional ANOVA analyses, see also response to **RC2***
356 **Major Comments 4**

357 **RC2 Minor Comment 4:**

358 Figure 9 (L543). **AC** *We shall improve readability by increasing the size of this Figure*
359 *and its labels in the final manuscript. The Figure can then be opened in a Figure viewer.*

360 **RC1 Minor Comment 19:**

361 L 549 **AC:** *We describe and discuss the effect of TWU normalisation here (Fig. 9). The*
362 *caption and related paragraphs show the seasonal and inter-year patterns. These are*
363 *also now explored statistically in the new ANOVA models.*

364 **RC1 Minor Comment 20:**

365 L 564 **AC:** *We have deleted this reference to precipitation as it is known that there are*
366 *multiple environmental and soil drivers affecting transpiration which we will not be testing*
367 *statistically within this paper.*

368 **RC2 Minor Comment 5:**

369 L 570 **AC** *We have rephrased this sentence. We previously described Figure 11 rather*
370 *than statistically tested the conditions in each month. The assumption in this paper is*
371 *that all trees experience the same or very similar temperatures and VPDs on a given*
372 *day, so that comparison of treatments will be evident visually and give pointers for our*
373 *statistical analysis of the hypotheses (see comments on ANOVA) and exploration in our*
374 *future research.*

375 **RC2 Minor Comment 6:**

376 L 585 **AC** *We describe the effects of herbivory in respect of decreasing leaf area and*
377 *canopy closure timing. By removal of leaf tissue herbivory reduces the whole tree*
378 *transpiration (less leaf so less water usage). The supporting reference describes this in*
379 *more detail. LL622-623 also explain the effects of the herbivory attack on canopy*
380 *closure and hence throughfall.*

381 **RC1 Minor Comment 21:**

382 L 591 **AC:** *Noted. We have moved this paragraph to the end of Section 2.2*
383 *Measurements overview .*

384 **RC1 Minor Comment 22:**

385 **AC:** *Section 3.4 is a combined results and discussion as are all subsections in section 3.*
386 *The two hypotheses and objectives are reviewed in section 3.5. If a separate Discussion*
387 *section would be clearer we will incorporate this when we redraft the manuscript.*

388

389 **Part 3 AC – changed text.**

390 **Abstract: (RC1, RC2)**

391 Revised Abstract, 2nd, 3rd & 4th sentences (old LL14-22).

392 The present study investigates diurnal (i.e. daylight) water usage of old growth oaks within an
393 experimental treatment season from April to October inclusive. Over five years, from 2017 to 2022,
394 we collected individual tree data from eighteen oaks (*Quercus robur* L.) within a large-scale
395 manipulative experiment at the Birmingham Institute of Forest Research (BIFoR) Free-Air CO₂
396 Enrichment (FACE) temperate forest in central England, UK. Diurnal tree water usage per day (TWU)
397 across the leaf-on seasons was derived from these data....

398 **Introduction: (RC1, RC2)**

399 Revised old LL70-71

400 Primary producers may respond to elevated CO₂ (eCO₂) levels by assimilating and storing more
401 carbon, which for plants containing chlorophyll happens during photosynthesis.

402 Revised old LL71-74

403 Global carbon and water cycle models (Guerrieri et al., 2016; De Kauwe et al., 2013; Medlyn et al.,
404 2015; Norby et al., 2016) predict that, at least until the middle of the 21st century, trees and plants
405 could potentially photosynthesise more efficiently, which may induce increased carbon storage. This
406 could be beneficial for individual tree productivity.

407 Revised old LL109-111

408 Within the UK maritime temperate climate, only a few ecohydrological studies (e.g. Herbst et al.,
409 2007; Renner et al., 2016) have previously considered the sap flow responses to water availability
410 and drought for old growth *Quercus* species.

411 Revised old LL113-

412 The response of woody plants to drought varies considerably by species (Leuzinger et al., 2005;
413 Vitasse et al., 2019), location (e.g. north versus south in Europe Stage et al., 2017)), soil
414 characteristics such as soil texture (Lavergne et al., 2020) and combinations thereof (Fan et al., 2017;
415 Salomón et al., 2022; Sulman et al., 2016; Venturas et al., 2017).

416 Revised L156

417 We test for significant differences between treatments within these water usage distributions and
418 patterns.

419 Revised complete sections(old LL 46-159)

420 Long-term manipulation experiments enable prediction of how, under climate change, increased
421 atmospheric carbon dioxide levels and climate environmental extremes might affect plants and
422 ecosystems. Plant hydraulics are adapted to expected ranges of environmental parameters, with larger
423 plants exhibiting greater resilience to wider parameter variation due to their ability to maintain water and
424 food reserves. Large trees can maintain their transpiration rates even during water stress but remain
425 vulnerable (Süßel and Brüggemann, 2021). To maintain transpiration demands, trees accommodate to:
426 diel variation in solar radiation; respiration fluctuation; high temperatures; and seasonal soil water

427 deficits. Short-term mechanisms include stomatal regulation, stem diameter variations (Sánchez-Costa
428 et al., 2015), use of stored plant water and use of available water at variable soil depth (David et al.,
429 2013; Flo et al., 2021; Gao and Tian, 2019; Nehemy et al., 2021). Longer term strategies include
430 development of resilient root structures, (David et al., 2013; Flo et al., 2021) and minimisation of
431 embolism mitigated by different xylem structures (Gao and Tian, 2019). The ability of mature trees to
432 withstand climate extremes may rely in part on using these buffering traits which act to prevent
433 permanent damage and maintain viability (Moene, 2014; Iqbal et al., 2021). This prompts the further
434 question of how increasing atmospheric carbon dioxide levels will affect the hydraulic resilience of trees.
435 The response of woody plants to drought varies considerably by species (Leuzinger et al., 2005;
436 Vitasse et al., 2019), location (e.g. north versus south in Europe Stagge et al., 2017)), soil
437 characteristics such as soil texture (Lavergne et al., 2020) and combinations thereof (Fan et al., 2017;
438 Salomón et al., 2022; Sulman et al., 2016; Venturas et al., 2017). Trees require water/ water vapour
439 at all stages of life experiencing insufficient water at times (e.g. under elevated temperatures and
440 drought) so tree species have evolved different root traits (Montagnoli, 2022) and hydraulic
441 characteristics (Sperry, 2003) to maintain their fitness to their environment. Volkmann et al., (2016)
442 used rainwater isotopes to track soil water sources for sessile oak (*Quercus petraea*) and beech
443 (*Fagus sylvatica*). Sánchez-Pérez et al., (2008) studied oak (*Quercus robur*), ash (*Fraxinus excelsior*)
444 and poplar (*Populus alba*). Both studies found that use of soil water at different depths varied between
445 species and seasonal variation of climatic conditions. Trees therefore exhibit variable resilience to
446 water shortage/ excess and other environmental stressors (Brodribb et al., 2016; Choat et al., 2018;
447 Grossiord et al., 2020; Landsberg et al., 2017; Martínez-Sancho et al., 2022; Niinemets and
448 Valladares, 2006; Schäfer, 2011; Süßel and Brüggemann, 2021) with a broad spectrum of sometimes
449 species-specific strategies and coping mechanisms (Schreel et al., 2019).

450 **Future-forest atmospheric carbon dioxide and water usage**

451 Primary producers may respond to elevated CO₂ (eCO₂) levels by assimilating and storing more
452 carbon, which for plants containing chlorophyll happens during photosynthesis. Global carbon and
453 water cycle models (Guerrieri et al., 2016; De Kauwe et al., 2013; Medlyn et al., 2015; Norby et al.,
454 2016) predict that, at least until the middle of the 21st century, trees and plants could potentially
455 photosynthesise more efficiently, which may induce increased carbon storage. This could be
456 beneficial for individual tree productivity. Stomatal regulation determines the trade-offs between
457 carbon assimilation and water loss and determines the rate and quantity of water usage seen in the
458 stems of woody plants. Water usage at tree level is, therefore, a strongly integrative measure of the
459 whole plant response to environmental drivers (such as temperature and precipitation) and
460 experimental treatments (such as eCO₂).

461 Untangling the canopy water exchange and soil moisture hydraulic recharge dynamics within forest
462 Free-Air CO₂ Enrichment (FACE) experiments can be complex, but responses to eCO₂ manipulations
463 (including stepwise increases (Drake et al., 2016)) inform our understanding of plant responses to
464 climate change scenarios. Specific studies concerning transpiration and water savings of eCO₂
465 responses (Ellsworth, 1999; Li et al., 2003) have already improved the model predictions (De Kauwe

466 et al., 2013; Donohue et al., 2017; Warren et al., 2011) and here we seek to enable further prediction
467 improvements.

468 Experimental research into ecohydrological responses of old growth deciduous forest to changing
469 atmospheric CO₂ levels has been limited. The Web-FACE study (Leuzinger and Körner, 2007)
470 reported on temperate old growth species and found that eCO₂ reduced water usage in *Fagus*
471 *sylvatica* L. (dominant) and *Carpinus Betula* L. (subdominant) by about 14% but had no significant
472 effect on the water usage of *Quercus petraea* (Matt.) Liebl., the other dominant species present.
473 There were a small number of trees (six) of a *Quercus* species included in Leuzinger and Körner's
474 (2007) study, with water savings monitored by accumulated sap flux (normalised against peak values
475 in each tree) over two 21-day periods. Changes in water usage by old growth oak trees at eCO₂ when
476 measuring for longer periods (greater than a month) across the leaf-on season have not previously
477 been reported.

478 The paucity of studies of the water usage of mature temperate trees under eCO₂ significantly
479 weakens model-data comparisons at FACE sites (De Kauwe et al., 2013). Warren et al., (2011)
480 reviewed the forest FACE experiments which, apart from Web-FACE, all constituted younger
481 deciduous and mixed plantations less than thirty years old (Schäfer et al., 2002; Tricker et al., 2009;
482 Uddling et al., 2008; Wullschleger & Norby, 2001; Wullschleger et al., 2002). Some of these studies
483 are long-term (> ten years) but all are limited in their period of monitoring sap flow, maximum
484 continuous data periods being covered by Schäfer et al., (2002) at Duke forest USA (1997-2000) and
485 lesser periods by Oak Ridge National Environmental Research Park (ORNL) USA and POP/
486 EuroFACE (Wullschleger & Norby, 2001; Tricker et al., 2009). Larger numbers of young trees (252
487 aspen–birch) were monitored for sap flux by Uddling et al., (2008), whereas most recent sap flow
488 studies of oak have either been single trees of different species (e.g. Steppe et al., 2016) or short-
489 term proof-of-concept studies using experimental instrumentation (Asgharinia et al., 2022).

490 There are further (2010 onwards) sap studies of deciduous oak which do not manipulate CO₂ but
491 which offer helpful data for comparison, for example within Europe (Aszalós et al., 2017; Hassler et
492 al., 2018; Perkins et al., 2018; Schoppach et al., 2021; Süßel and Brüggemann, 2021; Wiedemann et
493 al., 2016) and North America (Fontes and Cavender-Bares, 2019). Robert et al. (2017) have reviewed
494 the characteristics of these old growth species from multiple studies which help us to place our results
495 in context. Within the UK maritime temperate climate, only a few ecohydrological studies (e.g. Herbst
496 et al., 2007; Renner et al., 2016) have previously considered the sap flow responses to water
497 availability and drought for old growth *Quercus* species.

498 **Improving global vegetation models and questions of scale.**

499 Global vegetation models have been developed based on leaf-level plant knowledge alongside that of
500 soil-tree-atmosphere exchange (e.g. Medlyn et al., 2015). These models have predicted reduced
501 canopy conductance, G_s and increased run-off in future climate scenarios, but an important gap has
502 been identified between estimated and observed water fluxes (De Kauwe et al., 2013).

503 Canopy/ leaf transpiration estimates from stem xylem sap flux (Granier et al., 2000; Wullschleger and
504 Norby, 2001; Wullschleger et al., 2002), use the parameter canopy conductance (G_s) to reflect how

505 the whole canopy transpires rather than concentrating on individual leaf stomatal conductance to
506 water. Measurements of G_s and transpiration and partitioning of evapotranspiration in deciduous
507 forests (Tor-ngern et al., 2015; Wehr et al., 2017) have now clarified relationships between canopy
508 parameters and environmental variables PAR, VPD and precipitation. Long-term carbon and water
509 flux data from flux towers in forest ecosystems (e.g. Ameriflux (Baldocchi et al., 2001), Euroflux
510 (Valentini, 2003), FluxNet (Baldocchi et al., 2005)) and satellite datasets such as EOS/Modis
511 worldwide (Huete et al., 1994), have provided canopy level and landscape wide data. Plant focused
512 environmental manipulation studies, such as FACE, can provide data on individual parameters and
513 processes to inform and challenge the models.

514 At the forest scale, studies of the effects of the European drought (2018-2019) on forested landscapes
515 have shown that recovery time for surviving trees may be several years, affecting both plant growth,
516 stem shrinkage (Dietrich et al., 2018) and branch mortality during that time, especially for old growth
517 deciduous species (Salomón et al., 2022). At this forest scale (Keenan et al., 2013; Renner et al., 2016),
518 there is also a more complex impact on ecosystem and atmospheric demands, as planetary-scale CO_2
519 levels increase affecting boundary layer feedbacks.

520 In contrast to forest- and leaf-scale studies, the present study is tree-focused and bridges the data
521 gaps identified previously (Medlyn et al., 2015; De Kauwe et al., 2013) in respect of model-data scale
522 mismatch. Tree-scale studies have provided essential data for calibration and validation of tree-water
523 models (De Kauwe et al., 2013; Wang et al., 2016;), identified key parameters driving responses to
524 expected water shortages (Aranda et al., 2012) and compared species differences in mature tree
525 responses to ambient (Catovsky et al., 2002) or eCO_2 (Catoni et al., 2015; Tor-ngern et al., 2015). Xu
526 and Trugman, (2021) have updated the previous empirical parameter approach to global vegetation
527 modelling, reinforcing the need to use measured tree parameters (such as sapwood area) to improve
528 model predictions of climate change response.

529 Here we focus on whole tree species traits and link these parameters to diurnal (i.e. daylight) tree
530 water usage per day (TWU, litres d⁻¹) from stem xylem sap measurements, affirming the influence of
531 leaf-on season precipitation and solar radiation/ air temperature.

532 Measurements of xylem sap flux are marginally intrusive, providing highly time-resolved plant water
533 usage data for several years with minimal maintenance. Heat-based measurement techniques (Forster,
534 2017; Granier et al., 1996; Green & Clothier, 1988) have been used over the past 40 years in
535 measurement of plant xylem hydraulic function (Landsberg et al., 2017; Steppe and Lemeur, 2007) with
536 automated data capture enabling increasingly realistic models of whole tree xylem function.

537 **Objectives, research questions and hypotheses**

538 This study provides new gap filling data to characterise seasonal and inter-year patterns of daily water
539 usage by old growth oak trees using monthly distributions. We test for significant differences between
540 treatments within these water usage distributions and patterns. The paper examines the limitations of
541 water usage measurement by compensation heat pulse (HPC) sap transducers. It also relates diurnal
542 tree water usage per day (TWU, litres d⁻¹) to measurable tree traits (bark radius and canopy area) and
543 examines variation of TWU with environmental drivers and soil moisture.

544 **Materials and Methods: (RC1, RC2)**

545 (modified text old LL 211)

546 (MacKenzie et al., 2021). Table S2 shows instrumentation types and related parameters used for
547 analysis within this paper.

548 ((modified old LL212-256)

549 We experienced early leaf-on herbivory attacks in oak by Winter moth larvae, especially in 2018 and
550 2019 (Roberts et al., 2022) decreasing leaf area by 20-30% and affecting canopy closure timing. A
551 longer dry period occurred in meteorological summer 2018 (Rabbai et al., 2023) with wide variation in
552 summer monthly precipitation across the study years.

553 **Seasonal definitions**

554 The seven months of CO₂ treatment per year (with six months of leaf-on photosynthesis) do not easily
555 divide into standardised meteorological seasons (Spring, Summer), so we define our months of interest,
556 including non-treatment months as shown in Table 1. The table includes two months, March to April of
557 pre-leaf growth when oak sap starts to rise.

558 **Soil and throughfall precipitation data collection**

559 Pre-treatment (2015-2017 for infrastructure arrays) and pre-project (i.e. 2017 onwards data for all *Ghost*
560 arrays) on-site soil and throughfall data were used to characterise the site. Supplementary (2018
561 onwards) throughfall/ soil monitoring sites were added (see Mackenzie et al. (2021)). Shallow soil
562 moisture and soil temperature data were captured at least half-hourly by the same CR1000 datalogger
563 as the sap flow data.

564 For plants, incident precipitation affects their function in several ways during the leaf-on season. Firstly
565 water droplets incident on the leaves which when combined with lack of sun prevent full photosynthesis.
566 The canopy water mostly evaporates or may drip to ground. Secondly throughfall (P_{ts} , mm) reaching
567 ground level may either runoff the surface being lost to the soil, infiltrate (providing some necessary
568 support for root rehydration and plant water intake or evaporate. Lastly the soil water percolates through
569 the soil layers to replenish the water table.

570 Water inputs of throughfall precipitation under the oak canopy (within 2 to 3 metres of a stem and
571 situated near a soil moisture monitoring position) were measured in all arrays, with Fig. S2 showing a
572 typical installation set-up.

Calendar months	FACE Treatment season label	Note	Oak phenology at BIFoR FACE				
			2017	2018	2019	2020	2021
March – April (eCO ₂ starts beginning April)	Budburst & first leaf	March is pre-treatment. First leaf dates for oak shown	6 April *	25 April *	29 Mar *	No data* (c. 6 th April)	27 April *
May – June	Early leaf-on	Includes canopy closure early leaf of oak	-	-	-	-	-

July – August	Mid leaf-on		-	-	-	-	-
September – October (eCO ₂ until end October)	Late leaf-on	Includes start of senescence i.e. first tint	6 Sept	12 Sept	1 Oct	15 Sept	28 Sept
November - Feb	Dormant	All remaining non-treatment months	-	(after 21 Nov)**	26 Nov**	(after 03 Nov)**	07 Dec**
		Assumed leaf-fall season	6 Sept 2017 to 25 April 2018	12 Sept 2018 to 29 Mar 2019	1 Oct 2019 to c. 6 April 2020	15 Sept 2020 to 27 April 2021	28 Sept 2021 to 24 th April 2022

573 Table 1: Definition of treatment season periods and dates for oak phenology at BIFoR FACE according to
574 Nature’s Calendar criteria for years 2017–2021 (note this excludes canopy closure data - not recorded).
575 First tint is also recorded for year 2016 as 4th Oct. * On-site first leaf data (not obtained in 2020 due to the
576 Covid-19 pandemic; 6th April 2020 was noted as budburst, unverified first leaf is recorded as 24th April
577 2020). Note: Separate records of leaf-fall season are recorded for LAI calculation purposes as Nature’s
578 Calendar data does not discriminate first leaf fall by leaf colour. ** First bare tree date recorded. Nature’s
579 Calendar link: (<https://naturescalendar.woodlandtrust.org.uk/>). Phenocam additionally available for all
580 years (https://phenocam.nau.edu/webcam/roi/millhaft/DB_1000/).

581 **FACE and meteorological measurements**

582 Local precipitation (from a mixture of sources including Met. towers, see MacKenzie et al., (2021))
583 was recorded. Treatment levels of eCO₂, diurnal CO₂ treatment period, top canopy air temperature (T_a
584 , °C) and total solar radiation (TG , Watt m²) were available from the FACE control system (Hart et al
585 2020; MacKenzie et al 2021). Data were averaged across the six infrastructure arrays for TG and T_a
586 as the *Ghost* arrays have no FACE measurements (see supporting information Fig. S9).

587 **Tree selection**

588 There is variation inherent in biological individuals, in the same or different treatment types (Chave,
589 2013), which may not behave typically in space or time. This individual-tree experiment design aims to
590 minimise untypical variation. Accordingly the following criteria were used to select trees for sap flow
591 monitoring:

- 592 • canopy cover completely within the array (eCO₂ & aCO₂ arrays)
- 593 • central within the plot near logger and adjacent to access facilities at height (eCO₂ & aCO₂
594 arrays – for sampling and porometry access)
- 595 • straight stem, preferably with little epicormic growth
- 596 • no large dead branches within the canopy which might affect the comparative biomass of the
597 tree
- 598 • unlikely to experience seasonal standing or stream water at the base

599 Target oak trees for monitoring were also chosen to suit the physical limitations of the transducer to
600 logger constraints rather than randomly.

601 (modified Section 2.7-2.8 old LL257-401)

602 **Tree characteristics**

603 The tree size measurement approach is shown in Figure S3 and the range and mean-per-treatment
604 values of bark circumference (metres) for all target trees are tabulated (Table S3). All oak trees were
605 of similar height (circa 25 m). Tree stem circumference at insertion height of probes was measured at
606 installation (from 2017 onwards), and in subsequent winters (Jan 2020-Feb 2022). We note that tree
607 size will affect TWU (Bütikofer et al., 2020; Lavergne et al., 2020; Verstraeten et al., 2008).

608 Canopy spread of all target oak trees was measured around installation date (2017-2018) and repeated
609 for all oaks in early 2022 (Fig. S3). We assume that the two-dimensional canopy area, derived from the
610 mean canopy diameter plus stem diameter, is a good approximation to actual canopy spread and hence
611 the whole canopy surface experiencing leaf transpiration. For trees of similar height we assume that
612 allometric shape to estimate whole canopy volume will be similar. On the second occasion in 2022 we
613 measured the asymmetry of each tree stem across the probeset cardinal positions (East-West) and
614 right-angles to this (North-South) as a check of mean bark radius value for sap flux calculations.

615 Short incremental wood cores (circa 100mm long, 4mm diameter) were taken from two old growth oaks
616 outside of the experimental arrays. Microcores were also taken near all 36 target oak probeset
617 installation positions. These cores were used to determine wood hydraulic properties (Edwards and
618 Warwick, 1984; Marshall, 1958) for sap flux calculations (see also stage 4, Table A2 and definitions
619 Table A1). In summer 2021 woodcores taken from some of the target oaks were further analysed to
620 check the conversion (xylem woody matrix) factors from heat velocity to sap velocity and to verify the
621 active xylem radial width. The visibly active xylem (sapwood) is typically between seven and 50 mm
622 when viewed in wet cores but can more easily be measured in dried cores or disks. The uncertainty in
623 heartwood boundary H (m), as described in Appendix B, could be resolved in future similar studies by
624 taking short cores prior to installing instrumentation.

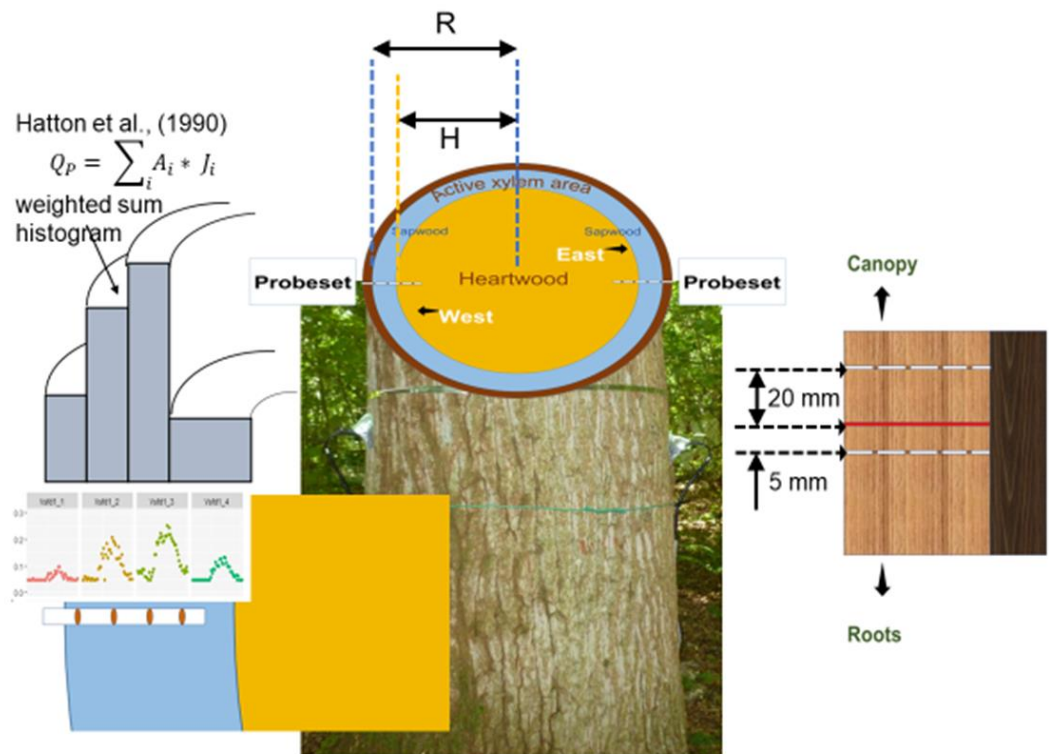
625 **Xylem sap flux**

626 Details of the xylem sap flux measurement method and associated calculations are in Table A2 and
627 Appendix C. Each target oak tree had two probesets, East and West facing and contained a central
628 heat pulse probe and two measurement probes (each containing four thermocouples upstream and
629 downstream of the heater (Fig. 2). Appendix B discusses the limitations which the time-out characteristic
630 places on this set of HPC data and consequent results. Validity of high outlier values is considered
631 within our analysis.

632 To determine whole tree sap flux several tree characteristics were used: (a) tree stem circumference at
633 insertion point, (b) bark thickness, from which we derived tree stem cambium radius at insertion point
634 R (m) and subsequently heartwood radius H (m) from sensor spacing. H (m) could not be determined
635 from 10 cm cores as these were not taken for all sap trees monitored.

636 The xylem sap flow installations in target trees commenced in Jan. 2017. All no-treatment-no-
637 infrastructure control (*Ghost*) oak trees provided data from August 2017 and commissioning of all 18
638 oak trees was completed by autumn 2018. All oak sap flow installations were successful and a total of
639 12,259 days of individual tree data (770,667 diurnal sap flux measurements across all months) were

640 processed for the 2017–2021 TWU analysis. Resulting data gaps in the earliest installations affected
 641 four of the 36 probesets installed in four trees August 2017 until September 2019.
 642



643
 644 **Figure 1: Showing sap probeset layout, spacing dimensions between probes and indicative illustration of**
 645 **Hatton et al., (1990) weighted sum histogram, where R (m) is the radius to the cambium and H (m) is the**
 646 **heartwood estimated radius, both at the probeset insertion height. All equations and variables also defined**
 647 **in Tables A1 and A2. Graphical insert is Figure C1(b).**

648 **Quality Assurance of raw HPC data**

649 Commissioning and failure data were recorded for each probeset. This enabled a combination of data
 650 file amendment (especially for the earliest installations on separate loggers) and post capture filtering
 651 to eliminate periods of invalid data for each probeset.

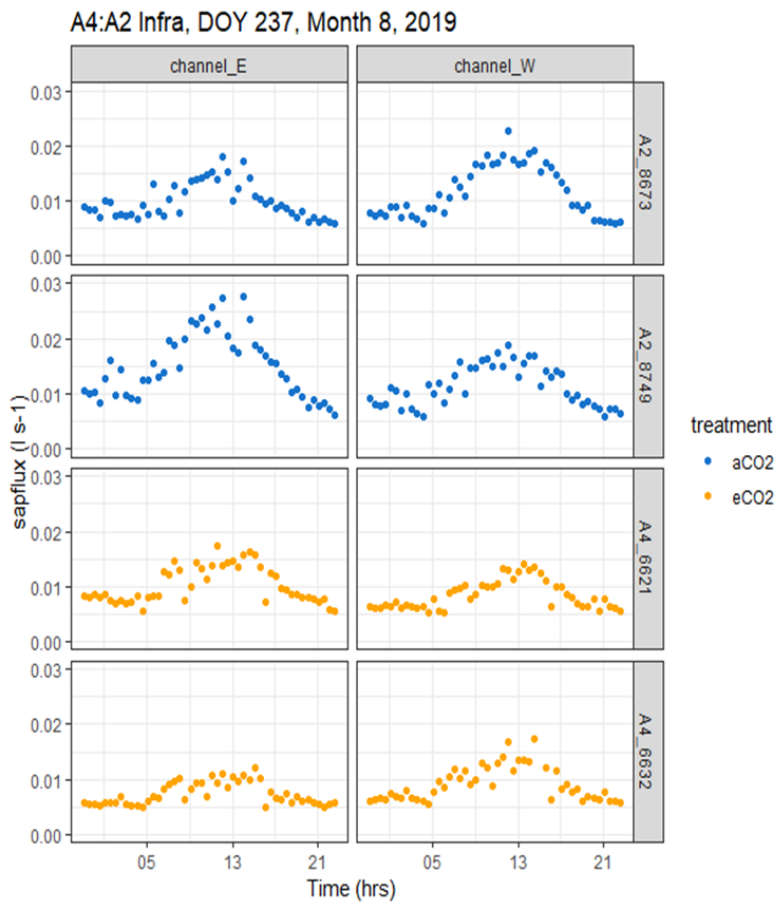
652 **Heat pulse to xylem sap flux calculations.**

653 Figure C3 shows example positional (i.e. thermocouple-specific), point sap flux density data from four
 654 probesets in two trees. Data from the thermocouple radial position giving the highest diurnal values
 655 (one thermocouple position for each probeset) are selected from the four-position data and shown
 656 across a 24-hour period (Fig. C1(a)). Note the increase in sap flux density towards the centre of the
 657 sapwood, decreasing again towards the heartwood.

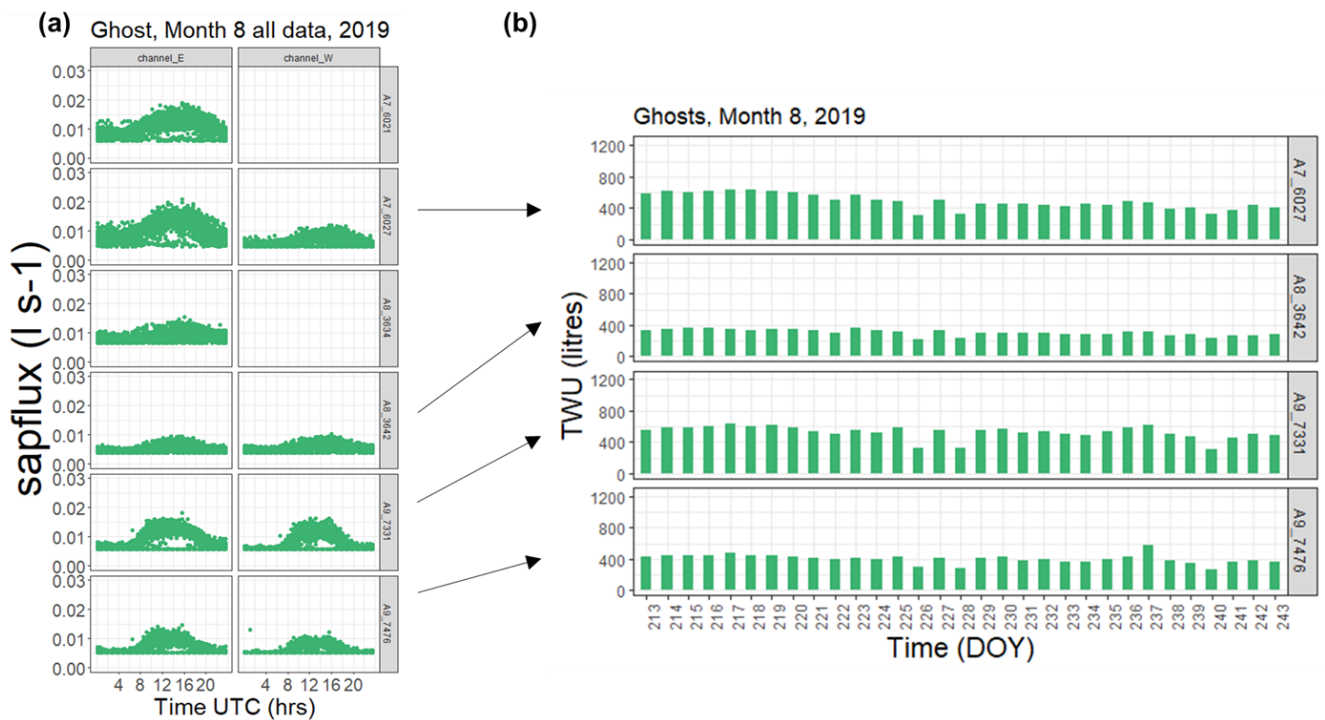
658 **Converting point xylem sap flux data to whole tree water usage.**

659 Using tree cambium radius (R) data, estimated heartwood radius (H) (0.05 m smaller than the inner
 660 sensor radial position), along with transducer radius positions (r_z), point sap flux density is converted to
 661 volumetric (half tree) total sap flux by using the integration of the point sap fluxes over the active

662 sapwood conducting area. Output from this stage (Stage 5, Appendix Table A2 and Appendix C) gives
 663 a combined sap flux for each probeset (Fig. 3).



664
 665 **Figure 2: An example data visualisation from a sunny August day in 2019 showing output of Stage 5**
 666 **combined point sap flux (litres s⁻¹) for four infrastructure trees with E facing (left) and W facing (right)**
 667 **probesets working: Array 4 (eCO₂), at top and A2 (aCO₂), at bottom. Time is in UTC.**



669 **Figure 3: Example *Ghost* Array xylem sap responses in August 2019. (a) diel (24hour) tree sap flux for all**
670 **days in August 2019 are superimposed. E (left) and W (right) facing probesets for six *Ghost* trees show**
671 **circumferential imbalance in xylem flux. All data for the individual month is superimposed across time-of-**
672 **day sampling (hours, UTC). Frequency of sampling is every 0.5 hrs. Faulty probeset positions are shown**
673 **blank. (b) Example of accumulative daily diurnal water usage (TWU) per tree totalled for E and W facing**
674 **probesets across month 8 2019 for four *Ghost* trees having both E and W probesets functioning with the**
675 **other two *Ghost* trees omitted due to faulty probesets. Time DOY.**

676 Diel sap flux patterns in August 2019 (before filtering to eliminate nocturnal data) are shown in Figure
677 4(a) as an example, for the *Ghost* arrays with East- and West-facing probesets in each column. The
678 sap flux data still show minimum threshold levels (which vary by tree size) determined by the post heat
679 pulse sampling period. Once again it is noticeable that there is often circumferential imbalance in xylem
680 sap flux in the East (lefthand column of Fig. 4(a)) and West (righthand column of Fig. 4(a)) probeset
681 position data, which reflect the asymmetry in growth ring width around the stem typical in these old
682 growth trees. The blank panels represent faulty probes (in two *Ghost* trees) corrected by autumn 2019.
683 To compare individual tree responses across the leaf-on seasons, further data filtering is required. We
684 filter the half-tree sap flux parameters using the solar azimuth and solar radiation parameters captured
685 from the FACE control instrumentation (solar azimuth > -6°, solar radiation > 0 W m⁻²) to give just
686 daylight (diurnal) data. Where both probesets in a tree are providing good data, a mean whole tree sap
687 flux is then derived and accumulated into TWU (Fig. 4(b)) as we had sufficient tree data to not include
688 results where probesets had failed. In future analysis we could use half-tree data once we understand
689 the proportion of sap flux and TWU exhibited by each probeset (e.g. following failure of contact with
690 sapwood of a previously functioning probeset).

691 The TWU data reported here compare well to results from other studies (Table S4: David et al., 2013;
692 Sánchez-Pérez et al., 2008; Tatarinov et al., 2005; Baldocchi et al., 2001).

693 (modified Section 2.9 old LL410-411)

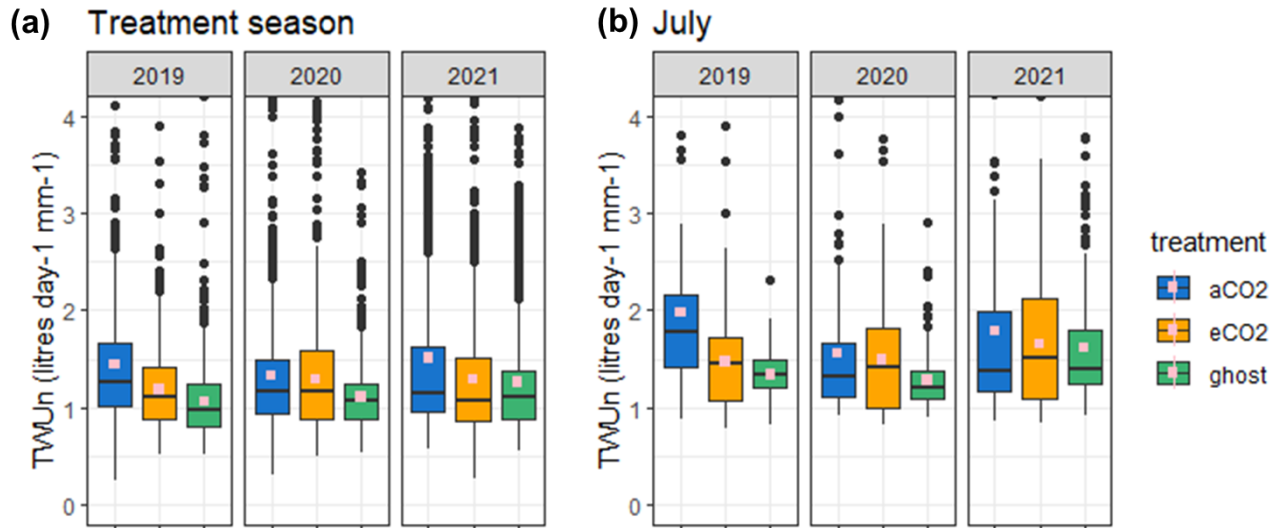
694 **Data processing, visualization and analysis**

695 ... points for outliers are used. ANOVA models to test hypotheses use the functions *anova* and
696 *summary*. Function *autoplot* from the *ggfortify* library is used to check the assumptions of normality of
697 the residuals.

698 **Results and Discussion (RC1, RC2)**

699 (additional figures and tables)

700 **ANOVA testing of hypotheses**



701

702 **Figure 4: Treatment comparison of TWU.** For years 2019- 2021 the TWU_n (litres d-1 mm⁻¹) data is shown for
 703 the three treatment types. The distributions are shown as box and whisker plots showing median and
 704 interquartile range (IQR, 25%ile to 75%ile) with whiskers calculated as 1.5 x IQR from the hinge and points
 705 for outliers. Mean values, calculated from the entire range of data, are shown as spots (pink). (a) The
 706 season data April to October is combined for each year. (b) July for each year is shown.

	2019 p value	2019 %	2020 p value	2020 %	2021 p value	2021 %
Season	< 0.001	-19%	p>0.05, actual value 0.079	-3%	<0.001	-13.9%
July only	< 0.001	-26%	p>0.05, actual value 0.37	-4.5%	p>0.05, actual value 0.19	-7.3%

707 **Table 2: ANOVA model_CO₂ p-value and % difference summary for eCO₂ TWU_n compared with aCO₂ TWU_n.**

	2019 p value	2019 %	2020 p value	2020 %	2021 p value	2021 %
Season	< 0.001	+37%	< 0.001	+20%	<0.001	+20%
July only	< 0.001	+48%	< 0.001	+22%	p>0.05, actual value 0.071	+9.9%

708 **Table 3: ANOVA model_inf p-value and % difference summary for aCO₂ TWU_n compared with ghost TWU_n.**

709 (revisions to old LL571-572)

710 ... The seasonality of TWU and TWU_n are different in the two years, but both exhibit similar general
711 responses to daylength within each treatment season.

712 (revisions to old LL643-651)

713 VWC at the start of each annual treatment season affects canopy usage during early leaf-on, but we
714 have seen that reduction of soil moisture across leaf-on season appears not to affect total water
715 usage. In our forest, as reported elsewhere from previous research, oak trees are most likely using
716 deeper > 1 m depth water resources by a combination of hydraulic recharge from deep tap roots and
717 capillarity from the perched water table. Our TWU results indicate that, other than on wet/ cloudy
718 days, oaks do not diminish their xylem sap flux and water usage across the treatment season, or
719 respond to depletion of shallow soil moisture, during very dry periods. For example, during the most
720 pronounced continuous dry period of the observation period (June to July 2018, Fig. 12), there
721 appears to be no inter-year difference in median *Ghost* tree diurnal sap flux (Fig. 6(a)) or trends in
722 95%ile sap flux (Fig. S5), or median TWU ((Fig. S6) in all trees, despite depletion of shallow soil
723 moisture...

724 **Appendix B: Limitations of the time-out characteristic and outliers**

725 (revisions to old LL 743. Addition of a sentence)

726 The limitations of the time-out characteristic and the effect this places on HPC data are recorded in
727 several references (e.g. see Tranzflo (New Zealand) manual). The limitations impact on our choice of
728 data processing (e.g. diurnal versus diel) and feed through into the statistics we report. These limitations
729 also introduce a truncation effect at lower heat velocities so that the distribution of the resulting raw and
730 processed data is not symmetrical.

731 **Appendix C: Details of xylem sap flow measurements and calculations**

732 In each research array a datalogger and multiplexer (CR1000+AM25T, Campbell Scientific, Logan,
733 Utah, USA) was used for year-round 24-hour capture of raw data from sap flux HPC probesets
734 manufactured by Tranzflo (New Zealand), soil and throughfall measurement devices.

735 The logger was programmed for data capture using CRBasicEditor under LoggerNet (versions to 4.6.2),
736 also by Campbell Scientific, Logan, Utah, USA. We tested our prototype installation set-up in mid-
737 summer 2017 to determine if we could capture the expected range of heat velocities and applied similar
738 capture programs to all array loggers.

739 Each target oak tree had two probesets, East and West facing using long (7 cm four-sensor) probes.
740 Each probeset was inserted at a stem height between 1.1 and 1.3 m and contained a central heat pulse
741 probe and two measurement probes (each containing four thermocouples for long probes respectively)
742 upstream and downstream of the heater (Fig. 2). Transducers were positioned radially in the stem (to
743 suit the ring-porous characteristics and bark thickness of old growth *Q. robur*). Each probeset was
744 protected from natural heating by reflective insulation covers during the treatment season.

745 During monitoring, a heater pulse of duration 1.5 to 2.5 secs was applied half-hourly through a heater
746 box (one per tree) to the heater probes. The pulse duration was dependant on the number of heaters
747 pulsed simultaneously. A 2 second pulse was standard for the two oak per array (four long probeset)
748 configuration. Each thermocouple pair in the upstream and downstream positions takes up to 330sec

749 (5.5 minutes) to reach a differential heat balance point and this time determines the minimum detectable
 750 heat velocity, for a time just within this timeout period. The thermocouple datalogger sampling rate of
 751 0.5 secs determines the maximum detectable speed (minimum time-to-balance), which, given normal
 752 interference levels, is adequate for deriving maximum heat velocity. 16 differential thermocouple
 753 configurations are sampled per array in one 6 minute timeslot every 30 minutes, giving time-to-balance
 754 t_2 data in seconds.

755 Data collection problems, due to logger earthing and sap probe misconnections at manufacture, caused
 756 data loss early in the project. Contact with sapwood was maintained for all oak trees from installation
 757 to March 2021, when two out of the 36 probesets failed.

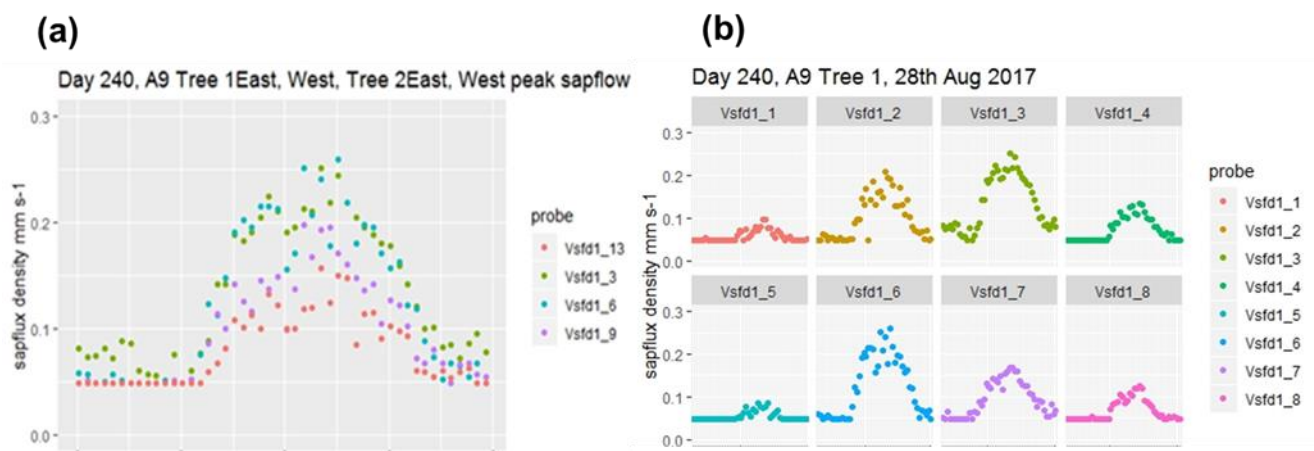
758 **Raw file processing**

759 Logger data from the nine C1000 FACE research loggers were collated by array and transducer type
 760 (i.e. 7 cm probeset datasets for oaks only) using 'R', then combined into year files for further data
 761 processing.

762 **Xylem sap flux calculations.**

763 Following quality checks, each stage of calculation to produce wound-corrected sap velocity and sap
 764 flux density at each transducer position (four per probeset) was performed in stages (see Table A2).
 765 Table A2 lists the methodology and equations along with associated literature sources for each stage.
 766 At stage 3 (Table A2), the Green and Clothier (1988) polynomial factors were used for wound
 767 compensation. For stage 4, the conversion factor c_1 was derived (Eq.(A4) and Eq.(A5)) to calculate
 768 xylem sap velocity from heat velocity in oakwood (Edwards and Warwick, 1984; Marshall, 1958).
 769 Measurement of wet and dry woodcores and microcores previously described provided data for
 770 derivation.

771 Figure C1 shows further details of positioning of peak sap flux through the sapwood in two trees. Figure
 772 C1(a) pools results from both trees. The diurnal maxima from the larger tree are larger than those for
 773 the smaller tree. Figure C1(b) pools probeset results from the larger tree, E facing (top) and W facing
 774 (bottom), illustrating stem imbalance.



775
 776 **Figure C 1: a) Example Stage 4 output showing peak point sap flux density in two trees for one sunny day**
 777 **in August 2017. Tree 1 (vsfd1_3 and vsfd1_6) bark radius is larger than Tree 2 (vsfd1_9 and vsfd1_13). b)**
 778 **Example Stage 4 output showing changes to point sap flux density across the active xylem for E facing**

779 **(top) and W facing (bottom) probesets of one tree (Tree 1) on the same day in August 2017. The lefthand**
780 **probe position is nearest to the bark and the righthand probeset position is nearest to the heartwood. Note**
781 **the peak value occurs at different sensor positions for the two probesets.**

782 The nocturnal/ pre-dawn response for the smaller tree in 1(a) (vsfd1_9 and vsfd_13)) and the less
783 vigorous thermocouple positions in the larger tree in Figure C1(b) (vsfd1_1, vsfd_4, vsfd1_5 and vsfd_8
784) have their minima determined by the previously mentioned time-out limit (i.e. t_z of 330 secs). These
785 minima do not affect the processing of diurnal values but influence nocturnal value accuracy of the
786 lowest point sap flux density (see Appendix B). The radial pattern of sap flux density increases in
787 amplitude to a peak position within the probeset measurement zone and then decreases again towards
788 the heartwood boundary as depth from the cambium increases (Fig. C1(a) and (b)), a characteristic of
789 these ring porous oak species. The radial amplitude patterns vary across seasons.

790 **Converting point xylem sap flux data to whole tree water usage.**

791 An adapted simple integration method (Hatton et al. 1990), based on a weighted average approach
792 was used where the point sap flux density is weighted by the areas of the annular rings associated with
793 each r_z . Fig. C1. Hatton et al. (1990) consider their method, in comparison with alternatives (e.g. fitting
794 a least-squares polynomial), a simpler and more accurate approach for estimation of the volume flux.

795 **(end Appendix C)**