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5 **A fruit growth approach to estimate oil content in olives**

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20 **Abstract**

21 Harvest timing in olive orchards has a strong effect on the quality and quantity of oil yield,  
22 but many farmers still lack simple and affordable quantitative tools for rationally deciding  
23 appropriate harvest dates. This study presents and tests a conceptual model for predicting  
24 fruit oil content ( $O_f$ , g oil fruit<sup>-1</sup>) from inexpensive measurements of fruit dry weight ( $w_f$ ).  
25 The model presents two physiologically relevant parameters, the fruit dry weight at the  
26 onset of the oil accumulation phase ( $w_{f0}$ ) and the ratio of accumulated oil per unit of fruit  
27 dry weight increase during the oil accumulation period ( $\beta$ ), the latter assumed invariable  
28 throughout ripening. A compilation of data on  $w_f$  and  $O_f$  dynamics collected from four  
29 experiments including six olive cultivars and contrasting conditions of water supply and  
30 crop load was used to test the model. Our results suggest that  $\beta$  could be fairly independent  
31 of crop load or watering regime and, probably, genetically controlled. By contrast,  $w_{f0}$  is  
32 clearly affected by both the cultivar and the availability of assimilates for fruit growth  
33 preceding oil accumulation, which makes it orchard- and year-specific. According to those  
34 premises, once cultivar-specific  $\beta$  values are available  $w_{f0}$  could be easily calibrated by  
35 either a single determination of  $O_f$  and  $w_f$  at any time during the oil accumulation phase  
36 (Approach A) or by directly measuring  $w_{f0}$  if the date for the onset of oil accumulation can  
37 be estimated (Approach B). Validation tests with an independently calibrated  $\beta$  showed an  
38 excellent performance for reproducing  $O_f$  patterns from  $w_f$  data using Approach A.  
39 Approach B satisfactory predicted oil accumulation rates, but absolute estimates of  $O_f$  were  
40 less reliable. Regardless of the calibration approach, the model is easy to implement and  
41 has a minimal cost, which satisfies the demand for inexpensive tools for monitoring oil  
42 accumulation dynamics.

43

44 **Keywords:** cultivar variability, fruit growth and development, crop modelling, oil  
45 accumulation, *Olea europaea* L.

46 **1. Introduction**

47 The recognition of the nutritional characteristics and health benefits of olive oil has led to  
48 an increased demand for this product, which has triggered the expansion of this tree crop in  
49 the last decades, both in traditional growing regions in the Mediterranean basin and new  
50 areas around the world. Covering more than 10 Mha nowadays, olive orchards represent  
51 one of the main oil crops worldwide (FAOSTAT, 2017).

52 As in any other oil crop, oil yield results from the product of fruit/seed number, fruit/seed  
53 weight and oil concentration at maturity. Understanding the dynamics of these components  
54 may be useful for establishing the optimal harvest date, as it is a pivotal agronomical  
55 decision that determines the yield and quality of olive oil, the two major revenue  
56 determinants in olive orchards (Mailer et al., 2007; Trentacoste et al., 2012).

57 In olive trees, oil synthesis takes place mainly in the parenchymatic cells of the fruit  
58 mesocarp (Rapoport and Moreno-Alías, 2017), but it is not until pit hardening has been  
59 completed that oil accumulation starts properly becoming the main sink for the assimilates  
60 allocated to fruits (Beltrán et al., 2017; Rapoport et al., 2013; Rapoport et al., 2017).  
61 Existing evidence suggest that the rate of oil accumulation is very high in late summer/early  
62 autumn and then decreases until the fruit reaches physiological maturity (Beltrán et al.,  
63 2005; García-Martos and Mancha, 1992; Trentacoste et al., 2010). However, both oil  
64 accumulation rate and ripening duration are substantially affected by fruit load (Barone et  
65 al., 1994; Dag et al., 2011, Fernández et al., 2015, 2018), cultivar characteristics  
66 (Camposeo et al., 2013; Lavee and Wodner, 1991) and both environmental and  
67 agronomical conditions (Gucci et al., 2019; Lazzez et al., 2011; Mailer et al., 2007). The  
68 concurrent effects of these factors challenge the definition of simple rules for determining  
69 the date at which fruits reach their maximum oil content.

70 Establishing rational criteria for deciding the best harvest date still represents a major  
71 challenge due to the existence of several trade-offs acting simultaneously. On the one hand,  
72 late harvests obviously ensure achieving high oil contents while simultaneously favor low  
73 fruit detachment force (Beltrán et al., 2017; Gamli and Eker, 2017), which can be critical  
74 for the harvesting operation for some cultivars. On the other hand, harvesting early prevents  
75 yield losses associated to natural fruit abscission and leads to oils of higher quality due to  
76 higher contents of some minor components that are responsible of some of the  
77 nutraceutical, organoleptic and gastronomic attributes of olive oil, such as polyphenols and  
78 tocopherols (Aguilera et al., 2017; Alagna et al., 2012; Caponio et al., 2001; Dag et al.,  
79 2011). In fact, the increasing pressure for obtaining oils of the maximum quality is already  
80 promoting an advance in the harvest date among the olive oil production sector.

81 In any case, farmers still lack simple inexpensive methods that allow them to decide the  
82 harvest date with some rational basis. In the best case, oil concentration is determined from  
83 fruit samples and compared to threshold values indicating on how far the orchard is from  
84 exploiting its oil accumulation potential (Zipori et al., 2016). This information is used to  
85 decide whether harvest should start or not. Obviously, recurrent olive samplings are  
86 required during the autumn to have a clear idea on how oil accumulation develops, which  
87 comes at an unaffordable cost for many farmers. The maturity index, based on fruit color  
88 (Beltrán et al., 2017), has also been used as an orientating approach for the decision-making  
89 of harvest timing, but the correlation between color and oil concentration is poor in many  
90 genotypes (Mickelbart and James, 2003; Navas-Lopez et al., 2019). Many other  
91 physiological and biochemical parameters such as fruit respiration (Ranalli et al., 1998),  
92 fruit detachment force (Almeida et al., 2016; Camposeo et al., 2013), changes in oil  
93 composition (Beltrán et al., 2017) and sugar content kinetics (Trapani et al., 2016) have

94 also been related to optimal harvesting periods, but they are more difficult to implement in  
95 practice and still require further research to assess the robustness of their relationships with  
96 oil concentration.

97 The aim of this article is to provide and test new simple approaches for predicting oil  
98 content dynamics based on inexpensive measurements of fruit dry weight that could be  
99 easily used by growers as a support for deciding the optimal harvest timing. Briefly, we  
100 consider that, since the start of oil accumulation, fruit oil content, “ $O_f$ ” (g oil fruit<sup>-1</sup>), can be  
101 linearly related to fruit dry weight “ $w_f$ ” (g fruit<sup>-1</sup>) as:

$$102 \quad O_f = \beta (w_f - w_{f0}) = \beta w_f - \beta w_{f0} \quad (1)$$

103 Where  $w_{f0}$  (g fruit<sup>-1</sup>) is the fruit dry weight at the onset of oil accumulation (i.e. at the end  
104 of pit hardening) and  $\beta$  (g oil g<sup>-1</sup>) is the amount of oil accumulated per g of fruit dry weight  
105 increase since the start of oil accumulation (Fig. 1).  $w_{f0}$  should depend on fruit growth rate  
106 from bloom to the end of pit hardening and must be cultivar dependent.  $\beta$  may also be  
107 cultivar-specific, but we hypothesize that it remains constant during the whole oil  
108 accumulation period and that it is independent of any factor affecting the availability of  
109 assimilates for fruit growth such as water status or crop load. This would imply that the oil  
110 content of fruits increases proportionally to fruit dry weight from the end of pit hardening  
111 to maturity, irrespective of the fact that fruit growth rates can vary with time or among trees  
112 of the same cultivar subjected to different conditions. Under these assumptions, recurrent  
113 measurements of  $w_f$  could be easily used for tracking  $O_f$  dynamics throughout the oil  
114 accumulation period.

115 The specific goals of this study are: (i) to test in different olive cultivars that oil  
116 accumulation represents a fixed fraction of fruit growth (i.e. constant  $\beta$ ), evaluating likely  
117 genotypic differences in this trait, (ii) to test whether  $\beta$  is independent of factors affecting

118 fruit growth rates, particularly, crop load and water status, and (iii) to propose and test  
119 simple approaches derived from the conceptual model for predicting oil accumulation  
120 dynamics based on inexpensive measurements of  $w_f$ , assessing its strengths and  
121 weaknesses. A compilation of data on fruit dry weight and oil accumulation dynamics  
122 coming from four experiments with young-potted and mature field-grown olive trees are  
123 used to address these objectives.

124

## 125 **2. Materials and methods**

### 126 *2.1. Experiment I*

127 Experiment I was performed in 2017 with 2 years-old trees of five olive cultivars:  
128 ‘Arbequina’, ‘Picual’, ‘Arbosana’, ‘Frantoio’ and ‘Changlot Real’ growing outdoors in 25-  
129 L pots at the Institute for Sustainable Agriculture (IAS-CSIC, Córdoba, Spain, 37.8°N,  
130 4.8°W, 90m altitude). The substrate of the pots was composed of a mixture of sand (30 %),  
131 silt (15 %) and peat (55 %). Ten trees per cultivar were planted in the winter of 2016 and  
132 maintained under appropriate growing conditions since then by applying drip-irrigation and  
133 slow-release fertilizers. In particular, enough irrigation was supplied to cover the maximum  
134 evapotranspiration. Water requirements were established from ad hoc periodical  
135 measurements of 24-h weight loss of the tree pots. The main meteorological variables were  
136 recorded throughout the experiment with an automated weather station located 500 m apart.  
137 The climate in the area is typically Mediterranean, with 580 mm of average rainfall mainly  
138 concentrated between autumn and spring, and 1390 mm of average reference  
139 evapotranspiration ( $ET_0$ ).

140 In 2017, four trees per cultivar were selected for the experimental measurements. Samples  
141 of five fruits per tree were collected at different moments of the fruit growing cycle starting

142 on July 19<sup>th</sup> and finishing in December 15<sup>th</sup>. After collecting the samples, the fruits were  
143 weighted for determining their fresh weight and, then, oven dried for 42 h at 105 °C to  
144 obtain  $w_f$ . Their oil content was subsequently measured using a NMR oil analyser (Del Río  
145 and Romero, 1999).

146

## 147 *2.2. Experiment II*

148 Experiment II was conducted through the years 2011, 2012 and 2013 within a 22.2-ha  
149 commercial hedgerow olive (cv. ‘Arbequina’) orchard located in ‘La Harina’ farm (20 km  
150 to the southeast of Córdoba, Spain, 37.7°N, 4.6°W, 170 m altitude). The orchard was  
151 planted in 2005 with 4 × 1.5 m tree spacing over a soil of clayish texture classified as a  
152 Vertisol (López-Bernal et al., 2015).

153 Four irrigation treatments were established using a randomized complete block  
154 experimental design with four replicates. Each of the 16 plots consisted of 40 trees in four  
155 adjacent rows. The irrigation treatments included a fully irrigated control (FI) that applied  
156 enough water to satisfy the maximum ET assuming a maximum crop coefficient of 0.75.  
157 The remaining treatments consisted of two similar regulated deficit irrigation treatments  
158 (D1 and D2) differing in the timing of the imposed water deficit and in its severity (Table  
159 S1), to which we added an additional treatment mimicking the irrigation applied by the  
160 manager of the commercial orchard (MI). The annual amounts of applied irrigation are  
161 shown for each treatment and year in Table 1, along with cumulative values of rainfall and  
162  $ET_0$ . Monthly values of those variables are also presented in the Supplementary Material  
163 (Table S1). Information on how the FI, MI and D2 irrigation treatments affected tree water  
164 status, trunk growth, transpiration and assimilation is available in López-Bernal et al.



165 (2015). Crop load was high in 2011 and 2013, and low in 2012, irrespective of the  
166 irrigation treatment (Table 1).

167 The time courses of  $w_f$  and  $O_f$  were periodically monitored every year from midsummer  
168 (July-August) to the orchard harvest date (late November-early December) in randomly  
169 hand-picked samples of 72 fruits per irrigation treatment and block. Fruits were always  
170 taken from the six central trees of the plots (12 fruits per tree). Fresh and dry weight of  
171 fruits and their oil content were measured as in Experiment I.

172

### 173 *2.3. Experiment III*

174 Measurements were performed in 2014 in a 12-year old organic commercial olive (cv.  
175 ‘Cobrançosa’) orchard located at “Vilariça” Valley (Trás-os-Montes, Portugal, 41.3 °N, 7.0  
176 °W, 150 m altitude), a typical olive growing area of Northeast Portugal. The climate in the  
177 area is Mediterranean (IPMA, 2015), with an average rainfall of 520 mm concentrated from  
178 autumn to spring, and 1130 mm of average  $ET_0$ . The soil is classified as Eutric Leptosols  
179 developed on metamorphic rocks (schists), of sandy loam texture. Tree spacing was 7 x 7 m  
180 and the experimental design was a complete randomized block, replicated three times. Each  
181 plot contained four central olive trees surrounded by 14 border trees and all measurements  
182 were made on the central trees of each plot.

183 Since 2013, five irrigation treatments were imposed in the orchard:

184 - FI: fully irrigated control, for which the water applied equaled the difference between the  
185 maximum (estimated) ET and rainfall.

186 - PRD: partial root drying system applying the same irrigation dose as FI to one half of the  
187 root system, with the irrigated and drying halves of the root-zone alternating every two  
188 weeks.

189 - SD40: sustained deficit irrigation that regularly received 40% of the water applied to FI  
190 - RD75: regulated deficit irrigation that received 75% of the water applied to FI, with a  
191 midsummer deficit period from mid-July to mid-August, reducing irrigation to 15% of FI.  
192 - RD40: regulated deficit irrigation that received the same seasonal amount of irrigation as  
193 SD40 with a midsummer deficit period without irrigation from mid-July to mid-August.  
194 Measurements of  $w_f$  and  $O_f$  were performed for each treatment at three different dates in  
195 2014 (October 2<sup>nd</sup>, October 20<sup>th</sup> and November 12<sup>th</sup>), using samples of 40 fruits per tree  
196 (three trees per treatment). That year crop load was low with no noticeable differences  
197 among irrigation treatments. Determinations of oil content were based on Soxhlet  
198 extraction (Donaire et al., 1977).

199

#### 200 *2.4. Experiment IV*

201 Experiment IV was conducted in a 10-year old commercial olive (cv. ‘Cobrançosa’)  
202 orchard located at Vilarica Valley (Trás-os-Montes, Portugal, 41.3 °N, 7.0 °W, 240 m  
203 altitude), in the same area as the previous experiment. The soil is classified as Eutric  
204 Leptosols developed on metamorphic rocks (schists), of sandy loam texture. Tree spacing  
205 was 6 x 6 m. The design of the experimental plot consisted of three adjacent blocks, each of  
206 these made of four rows with twenty olives trees, where only the six central trees were used  
207 for sampling. Three irrigation treatments were imposed during three consecutive seasons  
208 starting in 2004: full irrigation (FI), that received a seasonal water equivalent to 100%  
209 estimated crop evapotranspiration; sustained deficit irrigation (SD30), that received a  
210 volume of water equivalent to 30% of FI; and a rainfed treatment (RF).

211 In 2006, samples of 40 fruits per tree in 4 trees per treatment were collected periodically  
212 from September to December to monitor the dynamics of  $w_f$  and  $O_f$ . The latter was

213 determined by Soxhlet extraction. That year crop load was the highest of the three  
214 experimental seasons, with FI and RF showing the highest and lowest fruit numbers,  
215 respectively (Fernandes-Silva et al., 2010).

216 Further information describing the orchard characteristics, the climatic conditions during  
217 the experiment, the irrigation amounts applied to each treatment and their impacts on the  
218 water status and productivity of the trees is provided in Fernandes-Silva et al. (2010).

219

## 220 *2.5. Hypothesis testing*

221 Linear regression analyses of  $O_f$  versus  $w_f$  were performed to test whether  $\beta$  can be  
222 assumed both constant during the ripening period (i) and independent of the carbon  
223 availability for fruit growth (ii). As these assumptions refer to the oil accumulation phase  
224 only, data with oil concentrations below 5 % on a dry matter basis were, whenever present,  
225 excluded from the analyses. According to the conceptual model (Eqn. 1),  $\beta$  was estimated  
226 from the slope of the linear fit and  $w_{f0}$  was deduced from the intercept (as it should equal  
227 the product of  $\beta$  and  $w_{f0}$ ).

228 In Experiment I, regressions were performed for each cultivar independently, allowing us to  
229 compare the differences in the resulting linear models. Water stress and crop load  
230 presumably affect the availability of assimilates for fruit growth, so separate regressions  
231 were conducted for each irrigation treatment in Experiments III and IV, and for each  
232 combination of “irrigation treatment” x “year” in Experiment II. Finally, the regression  
233 lines were compared experiment by experiment, evaluating the statistical significance of the  
234 differences in the slopes and intercepts among the linear fits with the software Statistix 10  
235 for Windows (Analytical Software, Tallahassee, FL, USA).

236 An additional quantitative assessment of the sensitivity of model parameters to carbon  
237 availability was performed in Experiment II. Estimates of tree assimilation (López-Bernal  
238 et al., 2015) and records of crop load (Table 1) were used to calculate the cumulative values  
239 of assimilation per fruit ( $A_f$ , g C fruit<sup>-1</sup>) for periods preceding (June 18<sup>th</sup> to July 18<sup>th</sup>,  $A_{f1}$ )  
240 and following (August 2<sup>nd</sup> to September 26<sup>th</sup>,  $A_{f2}$ ) the onset of the oil accumulation phase.  
241 The dependency of  $w_{f0}$  and  $\beta$  on carbon availability was assessed from plots of their  
242 apparent values (obtained from the linear fits) versus  $A_{f1}$  and  $A_{f2}$ , respectively. The choice  
243 of the starting and ending dates of the two periods was constrained by both the availability  
244 of assimilation records for the three years and the uncertainty regarding the timing of the  
245 onset of oil accumulation. We left a gap between the two periods on purpose because,  
246 under the conditions of Southern Spain, the start of the oil accumulation phase has been  
247 reported to start 10-12 weeks after full bloom (Beltrán et al., 2017; García and Mancha,  
248 1992), with the average flowering date for ‘Arbequina’ in Córdoba being May 10th (De  
249 Melo-Abreu et al., 2004) (unfortunately, the actual dates of full bloom were not recorded in  
250 Experiment II).

251

## 252 *2.6. Testing model's predictive power in practice*

253 If the conceptual model presented in Eqn. 1 is sound in practice, then oil accumulation  
254 dynamics could be easily predicted from routinely measurements of  $w_f$ . Any increase in  $w_f$   
255 over time can be translated into an increase in  $O_f$  by multiplying by  $\beta$ . Furthermore,  
256 absolute values of  $O_f$  can also be theoretically estimated if a) oil concentration is measured  
257 once on a representative sample of fruits at any time during the oil accumulation phase or,  
258 b) if  $w_f$  is sampled around the date at which the oil accumulation phase starts in midsummer

259 (when  $w_f = w_{f0}$ ). These two simple approaches (hereafter referred as ‘Approach A’ or  
260 ‘Approach B’) were tested for the cultivar ‘Arbequina’.

261 The dataset of Experiment I was used for calibrating the value of  $\beta$ , while that of  
262 Experiment II was selected for validation. In Approach A, the intercept of the model ( $O_{f0}$ )  
263 is calibrated, for each combination of “irrigation treatment” x “year”, as:

$$264 \quad O_{f0} = O_{fj} - \beta w_{fj} \quad (2)$$

265 where  $w_{fj}$  and  $O_{fj}$  are the average dry weight and oil content of a representative sample of  
266 fruits taken on day ‘j’. For testing purposes,  $O_{f0}$  was calculated from the measured values of  
267  $w_{fj}$  and  $O_{fj}$  that were the closest to October 1<sup>st</sup> each year (‘j’ was October 5<sup>th</sup> in 2011,  
268 October 1<sup>st</sup> in 2012 and September 24<sup>th</sup> in 2013).

269 In Approach B,  $O_{f0}$  was determined from the product of  $\beta$  and  $w_{f0}$ , the latter estimated for  
270 each combination of “irrigation treatment” x “year” from the time course of  $w_f$ , assuming  
271 three fixed-date scenarios for the onset of the oil accumulation phase: July 20<sup>th</sup>, August 1<sup>st</sup>  
272 and August 10<sup>th</sup>. We selected these dates due to the aforementioned uncertainty regarding  
273 the onset of oil accumulation.

274 Model performances in reproducing measured oil dynamics were assessed using mean  
275 absolute error (MAE; from 0 to  $+\infty$ , optimum 0), root mean square error (RMSE; from 0 to  
276  $+\infty$ , optimum 0) and coefficient of residual mass (CRM, from  $-\infty$  to  $+\infty$ , optimum 0):

$$277 \quad \text{MAE} = \sum_i^n |S_i - M_i|/n \quad (3)$$

$$278 \quad \text{RMSE} = \sqrt{\sum_i^n (S_i - M_i)^2/n} \quad (4)$$

$$279 \quad \text{CRM} = 1 - \sum_i^n S_i / \sum_i^n M_i \quad (5)$$

280 Where  $M_i$  is the  $i$ th measured oil,  $S_i$  is the  $i$ th simulated *oil* and  $n$  is the number of  $O_f$   
281 measurements.

282

### 283 **3. Results**

#### 284 *3.1. Model's proof of concept*

285 The linear regression fits performed for each and every independent dataset of  $O_f - w_f$  were  
286 always highly significant ( $P < 0.001$ ), with the determination coefficient ranging from 0.72  
287 (Experiment IV, FI) to 0.999 (Experiment II, MI in 2013) and averaging 0.940 (Table 2).

288

#### 289 *3.2. Carbon availability effects*

290 The time courses of  $w_f$  and  $O_f$  in Experiment II (cv. 'Arbequina') exhibited considerable  
291 differences among years and irrigation treatments (Figure 2). On the one hand, the year of  
292 low crop load (2012, Table 1) always led to fruits of higher weight and oil content than its  
293 high crop load counterparts. On the other, FI showed higher values of  $w_f$  and  $O_f$  in 2012  
294 and 2013 than D1 and D2, although slight differences were noticed among treatments in  
295 2011. MI presented similar patterns of  $w_f$  and  $O_f$  to those of FI in 2011 and 2012, but it was  
296 the treatment with the lowest values in 2013. These differences in the patterns of fruit  
297 growth and oil accumulation among treatments were in consonance with the differences in  
298 water status and assimilation rates reported by López-Bernal et al. (2015) in the same  
299 experiment. In any case, estimates of  $A_f$  revealed that inter-annual differences in crop load  
300 had a higher weight on the carbon availability per fruit than the differences in water status  
301 among treatments (Fig. 3).

302 The values of  $\beta$ , estimated as the slope of the linear fits of  $O_f$  versus  $w_f$ , averaged 0.79 g oil  
303  $g^{-1}$  and ranged from 0.70 to 0.87 g oil  $g^{-1}$  (Table 2). All treatments averaged similar  $\beta$ , and  
304 no significant differences among them were found when they were compared within each  
305 year (Table S2). By contrast, the tests revealed statistically lower  $\beta$  for 2013 in relation to

306 2011 and 2012 in most cases (Table S2) and a slight direct relationship was found between  
307 this parameter and  $A_{f2}$  (Fig. 3A). The slope resulting from the linear regression between  $\beta$   
308 and  $A_{f2}$  was significant ( $P < 0.02$ ), although its value was low. No single combination of  
309 “irrigation treatment” x “year” showed significant differences in  $\beta$  in relation to the value  
310 obtained in Experiment I for the same cultivar (Table 2).

311 The intercept of the set-specific linear fits ranged from -0.34 to -0.15 g oil fruit<sup>-1</sup> (average -  
312 0.25 g oil fruit<sup>-1</sup>) (Table 2). Significant differences were usually found when comparing the  
313 same treatment among years and when comparing the treatments in each year, except for  
314 2011 (Table S2). The high variability in the intercepts was mainly driven by large  
315 differences in  $w_{f0}$ . In this regard, its apparent values ranged from 0.21 to 0.39 g fruit<sup>-1</sup>  
316 (Table 2, average 0.31 g fruit<sup>-1</sup>). The highest  $w_{f0}$  were observed in the low crop load year  
317 (irrespective of the treatment), and deficit irrigation treatments resulted in lower values than  
318 FI in 2012 and 2013. Moreover, the apparent estimates of  $w_{f0}$  presented a robust correlation  
319 with  $A_{f1}$  ( $r^2 = 0.84$ ,  $P < 0.001$ , Fig. 3B).

320 ‘Cobrançosa’ datasets generally showed no statistical differences when the slope or the  
321 intercept of the linear fits were compared among either irrigation treatments or experiments  
322 (Table S3, Fig. 4). Even if non-significant, differences in the estimates of  $w_{f0}$  between  
323 experiments were considerable, averaging 0.33 g fruit<sup>-1</sup> in Experiment III and 0.57 g fruit<sup>-1</sup>  
324 in Experiment IV (Table 2). Slightly higher values of  $\beta$  were also found in Experiment IV,  
325 irrespective of the treatment. In the FI treatments,  $\beta$  yielded 0.49 g oil g<sup>-1</sup> in Experiment III  
326 and 0.62 g oil g<sup>-1</sup> in Experiment IV.

327

328 *3.3. Cultivar effects*

329 No statistical differences were found among cultivars for  $\beta$  in Experiment I (Table S4), its  
330 values ranging from 0.75 ('Arbequina') to 0.82 ('Arbosana') g oil g<sup>-1</sup> (Table 2). The linear  
331 fits of  $O_f$  versus  $w_f$  were parallels, evidencing clear differences in their intercepts (Fig. 5,  
332 Table S4). In this regard,  $w_{f0}$  ranged from 0.26 ('Arbosana') to 0.60 ('Picual') g fruit<sup>-1</sup>.  
333 With the exception of the treatments FI and SD30 in Experiment IV, the slope of the fits  
334 obtained for 'Cobrançosa' was always significantly lower than those observed for the five  
335 cultivars tested in Experiment I (Fig. 5, Table 2).

336

#### 337 *3.4. Performance of Approach A in predicting oil accumulation dynamics*

338 Using the dataset of Experiment I for calibrating the slope of the model for 'Arbequina' led  
339 to  $\beta = 0.75$  g oil g<sup>-1</sup> (Table 1). As the intercept of the model is considered to be affected by  
340 carbon availability per fruit, it was obtained from pair measurements of  $w$  and  $oil$  around  
341 October 1<sup>st</sup> for each set in Experiment II (Eqn. 2). Its values were the lowest in 2012 and  
342 the highest in 2013, ranging from -0.28 to -0.17 g oil fruit<sup>-1</sup>.

343 Using the routine measurements of  $w_f$  during the oil accumulation phase to feed the model,  
344  $O_f$  predictions agreed very closely with observations irrespective of the year and irrigation  
345 treatment, as shown by the vicinity of the plots to the 1:1 line in Fig. 6. The satisfactory  
346 performance of Approach A for reproducing  $O_f$  dynamics is also supported by the low  
347 values of MAE (0.008 g oil fruit<sup>-1</sup>), RMSE (0.013 g oil fruit<sup>-1</sup>) and CRM (0.01), the latter  
348 indicating a negligible bias.

349

#### 350 *3.5. Performance of Approach B in predicting oil accumulation dynamics*

351 In Approach B, the model intercept is calibrated from the product of the slope (0.75 g oil g<sup>-1</sup>  
352 <sup>1</sup>, obtained from the independent set of 'Arbequina' in Experiment I) and  $w_{f0}$ , the latter



353 being estimated for three hypothetical date scenarios for the start of the oil accumulation  
354 phase. Using this procedure, the intercept averaged -0.21, -0.24 and -0.26 g oil fruit<sup>-1</sup> for  
355 the date scenarios July 20<sup>th</sup>, August 1<sup>st</sup> and August 10<sup>th</sup>, respectively. Regardless of the date  
356 scenario, the intercepts were always the lowest in 2012 and the highest in 2013.

357 Model performance was the best in overall terms assuming August 1<sup>st</sup> as the date for the  
358 onset of oil accumulation, with MAE, RMSE and CRM being 0.021 g oil fruit<sup>-1</sup>, 0.027 g oil  
359 fruit<sup>-1</sup> and 0.09, respectively (Table 3). However, the best date scenario was different when  
360 each year was analyzed independently. For instance, the model made the best predictions of  
361 oil for the year 2011 under the date scenario of July 20<sup>th</sup>, while August the 10<sup>th</sup> was the best  
362 for reproducing oil accumulation dynamics in the year 2013 (Fig. 7).

363

#### 364 **4. Discussion**

365 This study presents a simple conceptual model in which oil accumulation is linearly related  
366 to fruit growth on a dry matter basis. The two model parameters can be associated with  
367 physiologically relevant traits: the slope ( $\beta$ ) is the fraction of dry weight growth that  
368 accumulates in the fruit as oil, while the intercept is given by the product of  $\beta$  and fruit dry  
369 weight at the start of the oil accumulation phase ( $w_{f0}$ ). All the plots of  $O_f$  versus  $w_f$   
370 compiled in the four experiments of this article exhibited satisfactory linear fits ( $P < 0.001$ )  
371 with high determination coefficients (Table 2), which demonstrates the applicability of the  
372 model.

373 Understanding the factors that affect model parameters is a pivotal step to assess how it can  
374 be used in practice for predicting oil accumulation dynamics. In Experiment II  
375 ('Arbequina'),  $\beta$  was significantly lower in 2013 than in 2011 and 2012 in most cases  
376 (Table S2). This result might have been related with a lower assimilate availability per fruit

377 in 2013 (Fig. 3), contrary to our starting hypothesis, but the likely effect of carbon  
378 availability on  $\beta$  is of limited importance actually (or at least it was so for the range of  $A_f$   
379 covered in Experiment II). This is evidenced by the excellent performance of Approach A  
380 in reproducing oil accumulation dynamics for all the independent datasets (Fig. 6, Table 3)  
381 even if a fixed and independent value of  $\beta$  was always used. The lack of statistical  
382 differences among irrigation treatments in Experiments III and IV also support the premise  
383 of a negligible effect of carbon availability on  $\beta$  for ‘Cobrançosa’ (Fig. 4, Table S3). We  
384 must acknowledge, however, that assessing differences in  $A_f$  among treatments or  
385 experiments in those datasets was not possible with the available experimental information.  
386 Despite slight and non-significant differences among cultivars being noticed in Experiment  
387 I, the values of  $\beta$  obtained for ‘Cobrançosa’ in Experiments III and IV revealed  
388 substantially lower values. Consequently,  $\beta$  might be genotypically controlled, which  
389 would imply that this parameter requires cultivar-specific calibration.

390 Significant cultivar variability was also observed for  $w_{f0}$  in Experiment I, which was  
391 somehow expected as differences in fruit size among cultivars are usually evident from a  
392 few weeks after flowering (Beltrán et al., 2017; Lavee and Wodner, 1991). This fact  
393 originates, mainly, from genotypic differences in the rates of cell division (Hammami et al.,  
394 2011). Besides cultivar variability,  $w_{f0}$  also seems to be significantly determined by carbon  
395 availability as evidenced by results in Experiment II, where high crop load and water  
396 deficits led to lower values (Fig. 3, Table 2). Both the high rates of cell division and  
397 expansion in the first weeks following flowering and the production of lignin during pit  
398 hardening are metabolically expensive processes (Hammami et al., 2011, 2013; Rapoport et  
399 al., 2017), which explains why any limitation in the availability of assimilates is expected

400 to reduce  $w_{f0}$ . As a corollary,  $w_{f0}$  can vary every season even for the same orchard and  
401 cultivar, so it is a parameter that requires both year- and orchard-specific calibration.

402 Two approaches for calibrating  $w_{f0}$  are implicitly proposed in this paper assuming that  $\beta$  is  
403 both independent of carbon availability and genotypically-controlled. If the cultivar-  
404 specific value of  $\beta$  is available, the first approach (Approach A) just requires a single  
405 measurement of  $w_f$  and  $O_f$  from a representative fruit sample at any time during the oil  
406 accumulation phase to calibrate  $w_{f0}$ . The second (Approach B) prevents the need for oil  
407 determinations requiring, instead, measuring  $w_f$  at the date at which the oil accumulation  
408 phase begins. Both approaches can potentially yield excellent results, as demonstrated by  
409 the model performance tests conducted for Experiment II (Table 3, Fig. 6, Fig. 7).  
410 However, it must be noted that choosing the date of the onset of oil accumulation is rather  
411 challenging, as it varies from year to year, which translates into substantial bias in  
412 subsequent model predictions (Fig. 7, Table 3). The economic advantage of Approach B  
413 (no single  $O_f$  determination is needed) comes, therefore, at the cost of limited reliability in  
414 relation to Approach A when an absolute estimate of  $O_f$  is required. Nevertheless, we must  
415 note also that both approaches will yield equally reliable estimates of the rate of oil  
416 accumulation in the period between consecutive measurements of  $w_f$ .

417 The development of simple methods to predict the onset of oil accumulation seems a  
418 desirable target for future research. So far, measurements of pit breaking resistance with  
419 penetrometer devices (Rapoport et al., 2013) might provide a good indication of the ideal  
420 date for measuring  $w_{f0}$ , as oil accumulation is likely to start when pit hardening (and its  
421 competition for assimilates against the mesocarp) is reaching an end (Beltrán et al., 2017;  
422 Rapoport et al., 2017), but such measurements might be too laborious to be applied by  
423 farmers. On the other hand, a simple model for predicting the onset of oil accumulation

424 based on thermal time has been proposed for several cultivars in the arid environment of  
425 Mendoza, Argentina (Trentacoste et al., 2012). Unfortunately, the model has not been  
426 validated in the Mediterranean area and remains empirical, as acknowledged by their  
427 developers. In this regard, a simple thermal time approach might not be entirely satisfactory  
428 for predicting the onset of oil accumulation, as there are evidences pointing that the  
429 duration of pit hardening is affected by water stress (Hammami et al., 2013).

430 Beyond detailed technical examinations of parameter calibration, the model presented in  
431 this study is of the greatest relevance for the olive growing sector, as most farmers still lack  
432 inexpensive methods for following oil accumulation dynamics and rational criteria to  
433 decide the most appropriate harvest date accordingly. The best approach for monitoring oil  
434 accumulation dynamics to date depends on periodical determinations of oil concentration,  
435 which can be expensive for small growers. Our results suggest that recurrent measurements  
436 of  $w_f$  might be enough to predict oil accumulation dynamics reducing the number of  
437 determinations of oil concentration to a minimum (i.e. to “one” single determination if  
438 Approach A is used). Moreover, even if oil concentration cannot be determined, the model  
439 is able to predict oil accumulation rates (any increase in  $w_f$  can be easily converted into  $O_f$   
440 multiplying by  $\beta$ ) and theoretically yields approximate estimates of  $O_f$  if  $w_f$  is measured  
441 around the date at which the oil accumulation starts (Approach B). Another implicit point  
442 in our conceptual model is that the growth of  $w_f$  should stop once the fruit reaches its  
443 maximum  $O_f$ . From the practical point of view, this implies that the maximum oil content  
444 could be determined when  $w_f$  reaches a plateau. However, we must acknowledge that the  
445 absence of late harvests in our experiments prevented us to probe that point thoroughly.

446 Despite our results being promising, the conceptual model and its derived practical  
447 applications require further testing under contrasting environmental and agronomical

448 conditions including different cultivars in order to better assess their reliability. In this  
449 regard, many studies report that oil accumulation rates decrease under high temperatures  
450 (Lavee et al., 2012; García-Inza et al., 2014; Rondanini et al., 2014; Benlloch-González et  
451 al., 2019; Nissim et al., 2020). These observations might suggest that  $\beta$  is reduced at high  
452 temperatures, but the evidences available are not fully conclusive because the reduced oil  
453 accumulation rates may as well be the result of a decrease in fruit dry weight accumulation.  
454 In any case, we are confident that many researchers could easily contribute to the testing of  
455 the model under contrasting temperatures, environments or agronomical management  
456 conditions using already collected datasets of  $w_f$  and  $O_f$ . Finally, the model may also be  
457 used within a process-based model of olive orchards like OliveCan, which currently lacks a  
458 mechanistic simulation of oil production (López-Bernal et al., 2018).

459

## 460 **5. Conclusion**

461 This paper presents a conceptual model that estimates the oil content of olive fruits ( $O_f$ ) as a  
462 fixed proportion ( $\beta$ ) of their dry weight increase since the onset of the oil accumulation  
463 phase ( $w_f - w_{f0}$ ). A compilation of datasets of paired oil content and weight determinations  
464 from experiments with different cultivars and conditions of water status and crop load  
465 supports the validity of the model. The two parameters of the conceptual model ( $\beta$  and  $w_{f0}$ )  
466 are physiologically-relevant traits and can be obtained from the slope and intercept of linear  
467 regressions of  $O_f$  on  $w_f$ . Our results indicate that  $\beta$  could be cultivar-specific but remains  
468 fairly unaffected by factors modulating the availability of carbon per fruit, such as crop  
469 load or water stress. On the contrary, the fruit dry weight at the onset of oil accumulation  
470 ( $w_{f0}$ ) is both genotypically-controlled and dependent on crop load and photosynthesis  
471 during the earlier stages of fruit growth, which implies that it requires orchard- and year-

472 specific calibration. Fortunately, the model allows for easily determining  $w_{f0}$  from a single  
473 determination of  $O_f$  and  $w_f$  at any date during the oil accumulation phase provided that a  
474 cultivar-specific value of  $\beta$  is available (Approach A), or, alternatively, it can be measured  
475 directly if the date of the onset of oil accumulation can be estimated (Approach B). Overall,  
476 these model features indicate that oil accumulation rates could be estimated reliably from  
477 inexpensive measurements of  $w_f$  during autumn. This opens the door for providing olive  
478 growers with simple affordable methods to estimate  $O_f$ , which is a critical indicator for  
479 establishing optimal harvesting periods. Prior to that, further research testing the validity of  
480 our findings for different environmental conditions and/or new cultivars would be highly  
481 desirable.

482

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500

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632

633 **Tables**

634 **Table 1.** Annual values of rainfall, reference evapotranspiration (ET<sub>0</sub>) and applied  
 635 irrigation and fruit number for each irrigation treatment in Experiment II (FI, full irrigation;  
 636 MI, manager irrigation; D1, regulated deficit irrigation 1; D2, regulated deficit irrigation 2).

Year	Rain (mm)	ET <sub>0</sub> (mm)	Applied irrigation (mm)				Fruit number (fruits m <sup>-2</sup> )			
			FI	MI	D1	D2	FI	MI	D1	D2
2011	514	1229	465	326	306	243	859	745	779	772
2012	660	1266	591	376	471	240	307	391	387	406
2013	770	1178	536	144	277	219	1123	1116	1063	1055

637

638 **Table 2.** Results from independent linear regression analyses of fruit oil content ( $O_f$ , g fruit<sup>-1</sup>) versus fruit dry weight ( $w_f$ , g fruit<sup>-1</sup>) for each combination of “irrigation treatment” x  
639 “year” x “cultivar” in the four experiments. Note that the slopes of the linear regression  
640 lines are equivalent to  $\beta$ . The apparent dry weight at the start of oil accumulation ( $w_{f0}$ ) is  
641 calculated from the slopes and intercepts of linear fits. The last two columns ( $P_{intercept}$  and  
642  $P_{slope}$ ) show whether the slopes and intercepts differ statistically from the apparent value  
643 obtained for the olive cultivar ‘Arbequina’ in Experiment I.

Experiment	Year	Cultivar	Treatment	n	Intercept (g fruit <sup>-1</sup> )	Slope ( $\beta$ ) (g oil g <sup>-1</sup> )	r <sup>2</sup>	w <sub>f0</sub> (g)	P <sub>intercept</sub>	P <sub>slope</sub>
I	2017	Arbequina	FI	52	-0.22	0.75	0.88	0.29		
I	2017	Picual	FI	52	-0.48	0.80	0.80	0.60	***	n.s.
I	2017	Arbosana	FI	52	-0.21	0.82	0.92	0.26	***	n.s.
I	2017	Frantoio	FI	52	-0.33	0.74	0.96	0.45	***	n.s.
I	2017	Changlot	FI	52	-0.36	0.79	0.89	0.45	***	n.s.
II	2011	Arbequina	FI	5	-0.24	0.79	0.99	0.30	n.s.	n.s.
II	2011	Arbequina	MI	5	-0.26	0.84	1.00	0.31	n.s.	n.s.
II	2011	Arbequina	D1	5	-0.26	0.84	1.00	0.30	n.s.	n.s.
II	2011	Arbequina	D2	5	-0.26	0.84	1.00	0.31	n.s.	n.s.
II	2012	Arbequina	FI	7	-0.32	0.83	1.00	0.38	***	n.s.
II	2012	Arbequina	MI	7	-0.34	0.87	0.99	0.39	***	n.s.
II	2012	Arbequina	D1	7	-0.28	0.79	0.99	0.35	***	n.s.
II	2012	Arbequina	D2	7	-0.30	0.75	0.99	0.39	**	n.s.
II	2013	Arbequina	FI	8	-0.21	0.75	0.99	0.28	n.s.	n.s.
II	2013	Arbequina	MI	8	-0.15	0.72	1.00	0.21	***	n.s.
II	2013	Arbequina	D1	8	-0.18	0.72	1.00	0.25	n.s.	n.s.
II	2013	Arbequina	D2	8	-0.16	0.70	1.00	0.23	*	n.s.
III	2014	Cobrançosa	FI	9	-0.15	0.49	0.99	0.31	***	***
III	2014	Cobrançosa	PRD	9	-0.18	0.50	0.85	0.36	***	***
III	2014	Cobrançosa	RD75	9	-0.17	0.50	0.98	0.35	***	***

III	2014	Cobrançosa	RD40	9	-0.14	0.49	0.97	0.29	***	***
III	2014	Cobrançosa	SD40	9	-0.18	0.52	0.93	0.35	***	***
IV	2006	Cobrançosa	FI	20	-0.38	0.61	0.72	0.62	***	n.s.
IV	2006	Cobrançosa	SD30	20	-0.40	0.67	0.96	0.60	***	n.s.
IV	2006	Cobrançosa	RF	20	-0.28	0.57	0.91	0.49	***	**

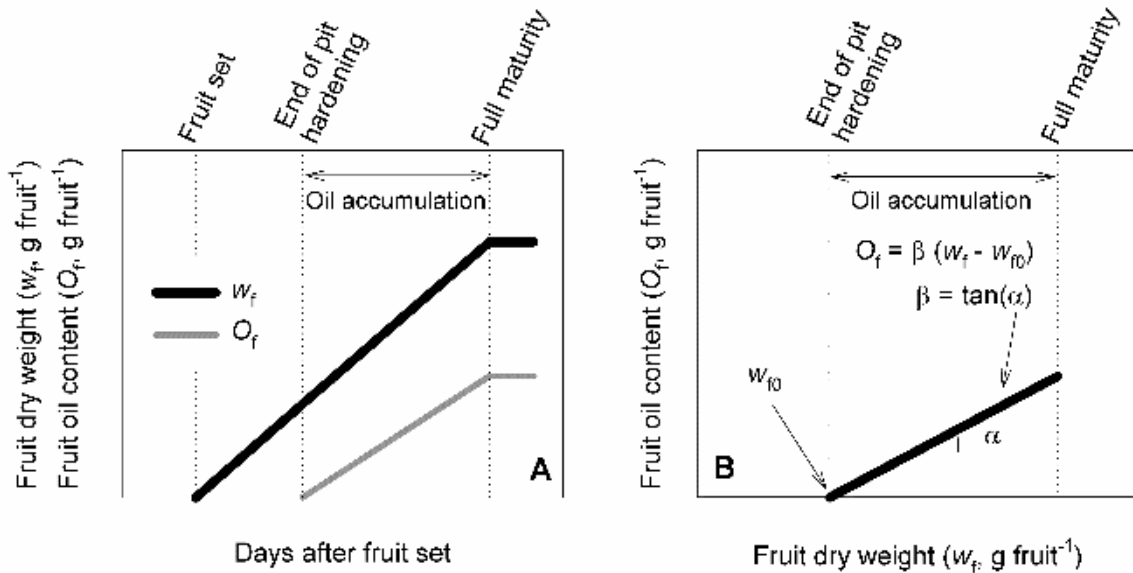


646 **Table 3.** Performance of Approach A and Approach B (for three scenarios for the onset of  
 647 oil accumulation) in reproducing fruit oil content ( $O_t$ ) dynamics. MAE is mean absolute  
 648 error, RMSE is root mean square error and CRM is coefficient of residual mass.

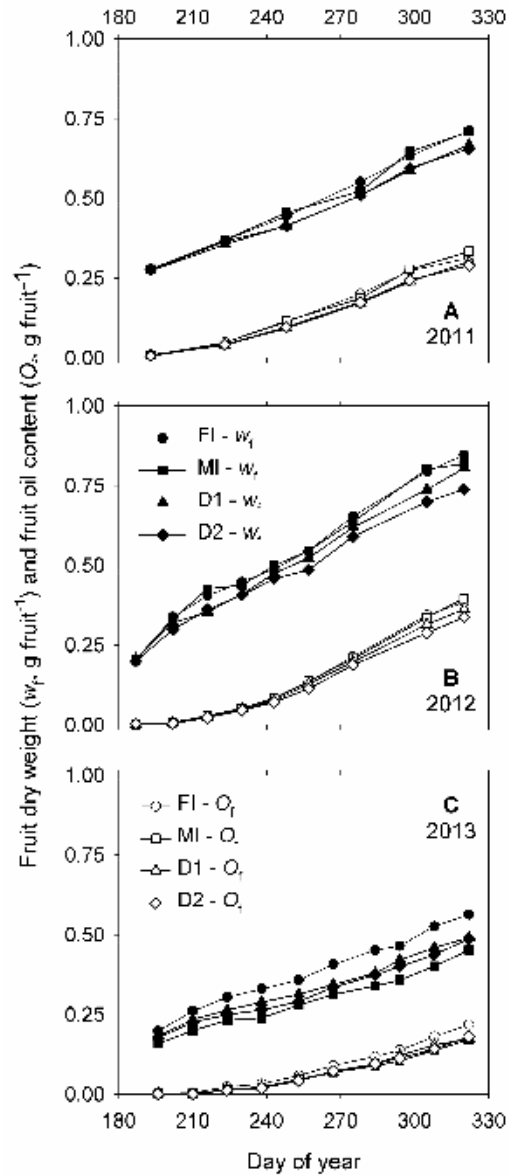
Parameter	Approach A	Approach B		
		July 20 <sup>th</sup>	August 1 <sup>st</sup>	August 10 <sup>th</sup>
n	80	80	80	80
MAE (g oil fruit <sup>-1</sup> )	0.008	0.027	0.021	0.032
RMSE (g oil fruit <sup>-1</sup> )	0.013	0.031	0.027	0.040
CRM	0.01	-0.15	0.09	0.23

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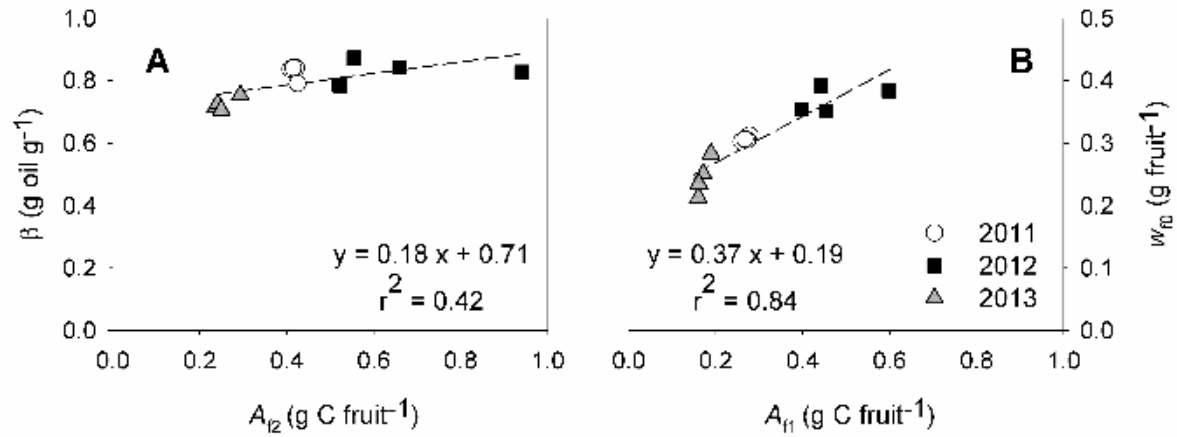
652  
 653 **Fig. 1.** Graphical description of the main features of the conceptual model. Panel A shows a  
 654 simplified time course of fruit dry weight ( $w_f$ ) and oil content ( $O_f$ ) from fruit set to full  
 655 maturity. While the former increases throughout this period, oil accumulation only starts  
 656 when the metabolically expensive pit hardening process has been completed. Panel B  
 657 shows the plot of  $O_f$  versus  $w_f$  assuming that the amount of oil accumulated per unit of dry  
 658 weight increase ( $\beta$ ) is constant during oil accumulation. Under these conditions,  $O_f$  is  
 659 linearly related to  $w_f$  from the end of pit hardening to full maturity. The slope of the  $O_f - w_f$   
 660 relationship during this period is indeed the parameter  $\beta$  while the intercept with the X-axis  
 661 represents the fruit dry weight at the onset of oil accumulation ( $w_{f0}$ ). Thus, both  $\beta$  and  $w_{f0}$   
 662 are physiologically relevant parameters that can be used to formulate a linear model to  
 663 estimate  $O_f$  dynamics from those of  $w_f$  during the oil accumulation period.



665

666 **Fig. 2.** Time course of fruit dry weight ( $w_f$ , closed circles) and fruit oil content ( $O_f$ , open  
 667 circles) in 2011 (A), 2012 (B) and 2013 (C) in Experiment II. Each type of symbol  
 668 corresponds to a different irrigation treatment: circles, squares, triangles and diamonds for  
 669 FI (full irrigation), MI (management irrigation), D1 (deficit irrigation 1) and D2 (deficit  
 670 irrigation 2), respectively.

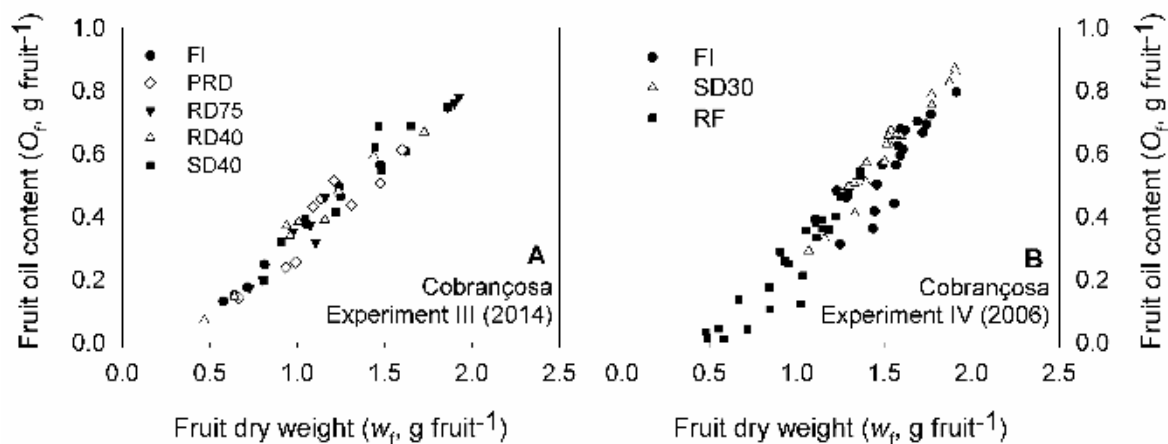
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672

673 **Fig. 3.** Relationships between the fraction of fruit dry weight increase allocated to oil  
 674 accumulation ( $\beta$ ) and cumulative assimilation per fruit from August 2<sup>nd</sup> to September 26<sup>th</sup>  
 675 ( $A_{f2}$ , A) and between fruit dry weight at the onset of oil accumulation ( $w_{f0}$ ) and cumulative  
 676 assimilation per fruit from June 18<sup>th</sup> July 18<sup>th</sup> ( $A_{f1}$ , B). Data are grouped in years (circles,  
 677 squares and triangles for 2011, 2012 and 2013, respectively).

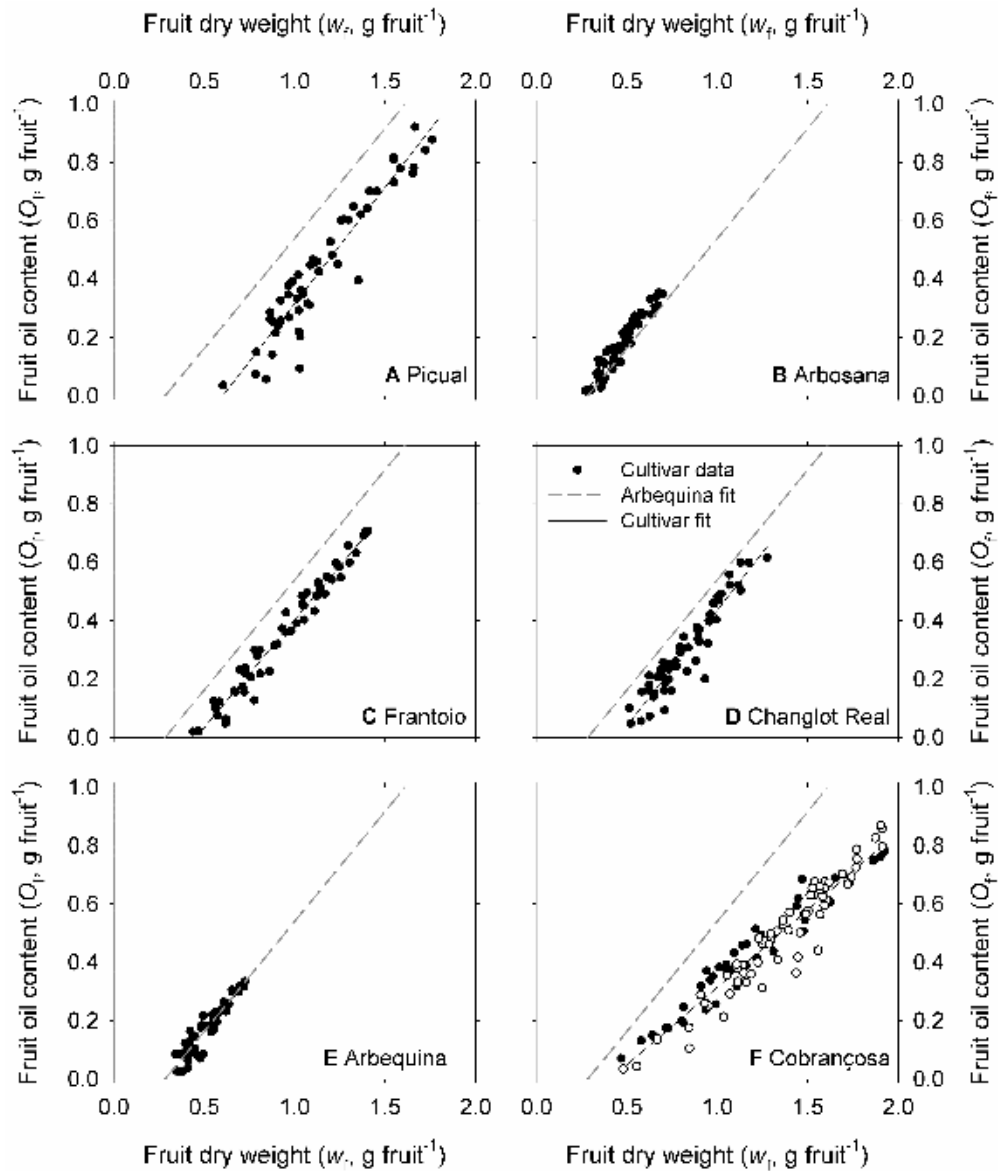
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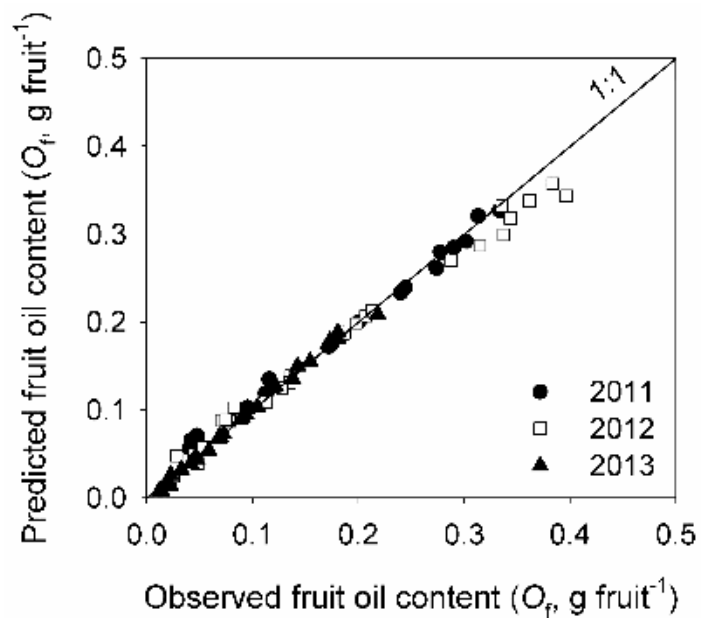
680 **Fig. 4.** Plots of fruit oil content ( $O_f$ ) versus fruit dry weight ( $w_f$ ) for the olive cultivar  
 681 ‘Cobraçosa’ in Experiments III (A) and IV (B). Each type of symbol corresponds to a  
 682 different irrigation treatment (FI, full irrigation; PRD, partial root drying; RD75, regulated  
 683 deficit irrigation applying 75 % of FI; RD40, regulated deficit irrigation applying 40 % of  
 684 FI; SD40, sustained deficit irrigation applying 40 % of FI; SD30, sustained deficit irrigation  
 685 applying 30 % of FI; RF, rainfed).

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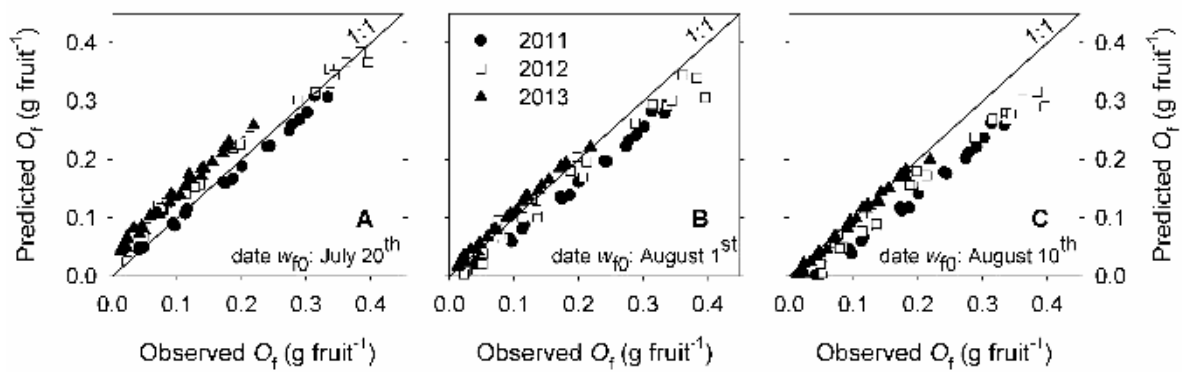
688 **Fig. 5.** Plots of fruit oil content ( $O_f$ ) versus fruit dry weight ( $w_f$ ) for the olive cultivars  
 689 ‘Picual’ (A), ‘Arbosana’ (B), ‘Frantoio’ (C), ‘Changlot Real’ (D), ‘Arbequina’ (E) and  
 690 ‘Cobrançosa’ (F). Data for ‘Picual’, ‘Arbosana’, ‘Frantoio’, ‘Changlot Real’ and  
 691 ‘Arbequina’ come from Experiment I. Data for ‘Cobrançosa’ comes from Experiment III  
 692 (closed symbols) and Experiment IV (open symbols), mixing all irrigation treatments. The  
 693 linear fit obtained for ‘Arbequina’ in Experiment I is shown in all panels with a grey dashed  
 694 line to serve as a reference.



695

696 **Fig. 6.** Validation of Approach A for the olive cultivar ‘Arbequina’ on the dataset of  
 697 Experiment II. Predicted versus observed plots of fruit oil content ( $O_f$ ) in relation to the 1:1  
 698 line. Data are grouped in years (circles, squares and triangles for 2011, 2012 and 2013,  
 699 respectively).

700



701

702 **Fig. 7.** Validation of Approach B for the olive cultivar ‘Arbequina’ on the dataset of  
 703 Experiment II using three scenarios for the onset of oil accumulation: July 20<sup>th</sup> (A), August  
 704 1<sup>st</sup> (B) and August 10<sup>th</sup> (C). Predicted versus observed plots of fruit oil content ( $O_f$ ) in  
 705 relation to the 1:1 line. Data are grouped in years (circles, squares and triangles for 2011,  
 706 2012 and 2013, respectively).