

1 Seasonal and inter-annual rainfall patterns drive Mediterranean cork oak  
2 woodland carbon uptake and sequestration

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20 **Abstract**

21 Cork oak woodlands are low impact agroforestry ecosystems with high biodiversity and high  
22 social and economic value. Current climate change projections for the Mediterranean region  
23 foresee a decrease in the amount, and a change in the seasonality, of precipitation as well as an  
24 increase in temperature, threatening productivity and future carbon sequestration in these  
25 ecosystems. Analyses of long-term observations of both carbon fluxes and meteorological  
26 variables contribute to a better understanding of the ecosystem's response to climate drivers and  
27 help predict its structural and functional stability under anticipated climate change.

28 In this study, we investigated the impact of rainfall variability on seasonal and inter-annual  
29 carbon sequestration in a Mediterranean cork oak woodland across seven hydrological years.  
30 Furthermore, the impact of rainfall and other climate variables on the start and end of the  
31 growing season of the understory was evaluated as well the sensitivity of carbon fluxes to the  
32 variability of the growing season length.

33 The site is a strong carbon sink with a mean annual Net Ecosystem Exchange (NEE) of about -  
34  $338 \pm 70 \text{ g C m}^{-2}\text{y}^{-1}$  turning into a weak carbon source only occasionally in autumn when  
35 rainfall is particularly low in the previous summer. We observed a time-lag effect of up to 12  
36 months in the correlation between rainfall and NEE with the highest influence occurring at 6-  
37 month time-lag for all carbon fluxes. The length of the growing season of the understory  
38 vegetation depended on rainfall and vapor pressure deficit in periods preceding the start and end  
39 of the growing season but also on winter rainfall and explained about 70% of inter-annual  
40 variability of summer gross primary productivity (GPP) and ecosystem respiration (Reco).

41 Analyses of inter-annual and seasonal rainfall patterns, as well as of associated lagged effects on  
42 carbon dynamics, are not only relevant for our understanding of ecosystem functioning under  
43 future climate change but also for planning forest management actions that maximize carbon  
44 sequestration in Mediterranean oak woodlands .

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46 **Keywords:** Mediterranean region; cork oak woodlands; CO<sub>2</sub> fluxes; eddy covariance; drought;  
47 precipitation

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## 48 **1. Introduction**

49

50 Cork oak woodlands are low impact agroforestry ecosystems with high socio-economic and  
51 conservation value, widely distributed in the western Mediterranean region. They are  
52 characterized by a sparse tree cover of cork oak trees (*Quercus suber* L) and a diverse  
53 understory vegetation, ranging from shrub formations to grasslands (Bugalho et al., 2011).  
54 Portugal has the largest cork oak habitat worldwide representing 23% of the national forested  
55 area (ICNF, 2019). In addition to the economically important cork production and pasture, cork  
56 oak woodlands provide several ecosystem services, such as microclimate regulation, soil  
57 retention, air quality improvement, nutrient regulation, biodiversity conservation and carbon  
58 sequestration (Bugalho et al., 2011).

59 About 70% of Portuguese cork oak woodlands are concentrated in the southern part of the  
60 country (ICNF, 2019), a region characterized by a Mediterranean climate, with scarce water  
61 resources and a long dry summer season, usually coupled with high temperatures and high  
62 radiation (Faria et al., 1996). Climate change scenarios for the region anticipate increased  
63 vulnerability resulting from the combination of a decrease in annual precipitation, changes in  
64 precipitation seasonality and an increase in temperature (Dubrovský et al., 2014; Giorgi and  
65 Lionello, 2008). An increase in the intensity and frequency of extreme events is also expected  
66 (IPCC, 2021).

67 Trees and understory species characteristic of Mediterranean oak woodlands are well adapted to  
68 seasonally dry climate conditions typical of this region. However, the stability of these  
69 ecosystems under prolonged and exacerbated droughts is largely unknown and could  
70 compromise many of their ecological functions including their carbon sequestration capacity  
71 (Mori et al., 2017). Even though individually Mediterranean regions provide only a modest  
72 contribution to the overall biosphere-atmosphere carbon budget, as part of semi-arid ecosystems  
73 they contribute considerably to the degree of uncertainty in global CO<sub>2</sub> fluxes variability, thus  
74 hindering predictions of the global carbon cycle (Ahlström et al., 2015). This underlines the

75 importance of understanding the environmental controls on carbon exchange fluxes in these  
76 regions.

77 A key feature to diagnose the complex responses of ecosystems to climate drivers has been the  
78 analysis of carbon and water exchange between the ecosystem and the atmosphere. Eddy  
79 covariance measurements provide time series of carbon and water fluxes and meteorological  
80 variables largely contributing to the understanding of the biophysical and climate factors  
81 determining the carbon exchange between the biosphere and the atmosphere from hourly to  
82 multi-annual time scales (Baldocchi, 2020). Canopy-scale photosynthesis (gross primary  
83 productivity; GPP) and ecosystem respiration (Reco) are the two processes determining the  
84 magnitude of the net ecosystem CO<sub>2</sub> exchange (NEE) between the biosphere and the  
85 atmosphere over time in terrestrial ecosystems. In spite of a different sensitivity to climate  
86 drivers (Baldocchi, 2008), GPP and Reco are generally strictly related since photosynthetic  
87 assimilated carbon constitutes the substrate for autotrophic respiration (Collalti et al., 2020) and  
88 also stimulates soil (Davidson et al., 2006) and heterotrophic respiration (Migliavacca et al.,  
89 2015, 2011). However, several studies have reported a decoupling of the two processes  
90 (Baldocchi and Penuelas, 2019; El-Madany et al., 2020; Ma et al., 2019; Migliavacca et al.,  
91 2015; Reichstein et al., 2007), mainly resulting from differences in the response to climate  
92 drivers or substrate availability. For example, in an oak-grass savanna in California, a  
93 decoupling was observed at the beginning of the autumn season when the presence of litter  
94 stimulates the heterotrophic respiration, before the understorey was sufficiently developed to  
95 contribute to the overall GPP (Ma 2016).

96 Temperature, precipitation and radiation are well known drivers of the variability of CO<sub>2</sub>  
97 exchange in different ecosystems (e.g. Fu et al., 2019; Jung et al., 2017; Marcolla et al., 2011;  
98 Wang et al., 2014). Likewise, vegetation traits such as biodiversity and ecosystem structure, are  
99 known to influence NEE variability (Anderegg et al., 2018; Musavi et al., 2017;  
100 RICHARDSON et al., 2007). Additionally, phenology can have a large impact on both  
101 photosynthesis and respiration, hence the analysis of phenological patterns may help explain  
102 observed trends in GPP and Reco. In cork oak woodlands, only the herbaceous understorey

103 shows marked seasonal patterns while evergreen trees maintain similar canopy cover all year  
104 round (Cerasoli et al., 2016) and the impact of understory phenology on carbon fluxes is  
105 difficult to ascertain and rarely considered (Correia et al., 2016; Ma et al., 2007).

106 Water availability is a major environmental factor constraining NEE fluxes in Mediterranean  
107 forests, controlling photosynthetic carbon assimilation rate and dampening respiration fluxes  
108 during dry hot periods (Pereira et al., 2007). Given the proneness of Mediterranean ecosystems  
109 to long dry periods, a better understanding of climate-ecosystem interactions demands a  
110 corresponding understanding of the impact of drought on GPP and Reco. Drought events can  
111 have a strong influence on the inter-annual variability of NEE as demonstrated in studies  
112 considering diverse terrestrial ecosystems and climatic zones (Baldocchi et al., 2018;  
113 Zscheischler et al., 2014).

114 Total annual precipitation contributes to the replenishing of soil moisture and groundwater  
115 resources allowing deep-rooted oak trees to survive the dry season (David et al., 2016). As such,  
116 understanding the response of Mediterranean ecosystems to rainfall variability at annual scales  
117 is essential to anticipate the potential effects of future climate change and increasingly frequent  
118 extreme events (Reichstein et al., 2013).

119 Changes in rainfall timing also have an impact on the growing season length (GSL) of the  
120 herbaceous layer. While a delay in autumn rains postpones the start of the season (Luo et al.,  
121 2018; Nogueira et al., 2017), in spring rainfall scarcity and high temperatures accelerates the  
122 dry-down period (Cerasoli et al., 2018). Hence, in addition to total rainfall, rainfall timing  
123 largely influences the carbon fluxes in cork oak woodlands, in particular those that have a  
124 considerable contribution from shrubs and grasses to the overall carbon balance (Correia et al.,  
125 2016; Dubbert et al., 2014; El-Madany et al., 2018; Luo et al., 2020; Piayda et al., 2014).

126 Furthermore, under expected conditions of high temperature and low soil moisture, transpiration  
127 of trees and grass is constrained and vapor pressure deficit (VPD) increases, further  
128 exacerbating water stress (Seneviratne et al., 2010). Hence, together with rainfall, VPD can also  
129 influence the length of the growing season of the herbaceous vegetation.

130 Examining the inter-annual variability of the start and end of the understory growing season is  
131 therefore crucial to our understanding of the impact of expected exacerbated water scarcity on  
132 the ecosystem carbon balance.

133 Carbon flux dynamics are not controlled only by their instantaneous dependency on climate and  
134 vegetation but also on time lagged effects (Besnard et al., 2019). However, in most of the  
135 studies only the direct effect of climate variables is considered, while the time lagged effect is  
136 generally ignored (Zhang et al., 2015). Considering the importance of time lagged effects, can  
137 contribute for our understanding of long term ecosystem responses to climate disturbances  
138 (Marcolla et al., 2011). The analysis of the inter-annual variability of carbon and respiration  
139 fluxes should therefore consider altogether climate drivers, vegetation traits and possible legacy  
140 effects to improve our understanding of factors controlling carbon exchange. Long-term datasets  
141 are necessary to observe processes at different time scales and to reveal short-term dependencies  
142 and legacy effects. Notwithstanding, few studies analysing the inter-annual variability of carbon  
143 fluxes have focused on Mediterranean oak woodlands (Costa-e-Silva et al., 2015; El-Madany et  
144 al., 2020, 2018; Ma et al., 2016).

145 In this work, we analyse seven years (2010–2017) of continuous eddy covariance (EC) flux  
146 measurements at a Mediterranean cork oak woodland to: i) evaluate the impact of rainfall timing  
147 and magnitude on carbon fluxes at inter-annual and seasonal scales; ii) identify any potential  
148 legacy effects, and quantify the corresponding time lags, between rainfall and carbon fluxes  
149 dynamics; and to iii) investigate the sensitivity of ecosystem carbon fluxes to the understory  
150 GSL and if such sensitivity is mediated by precipitation and VPD. The hypothesis tested is that  
151 the duration of the understory GSL has a significant impact on the annual fluxes and that it is  
152 affected not only by water scarcity close to the emergence and senescence of the understory  
153 but also by winter precipitation.

154

155

## 156 **2. Materials and methods**

157

158 *2.1 Study site*

159 Our study was conducted in a 50-year-old cork oak woodland (*Quercus suber* L.) at Herdade da  
160 Machoqueira do Grou, located in Central Portugal (39° 08' 20.9" N, 9° 19' 57.7" W, 165m  
161 altitude). The climate is Mediterranean, with mild, wet winters and hot, dry summers. Long  
162 term (1981–2010) mean annual rainfall is 653 mm and mean annual temperature is  $16.2 \pm 4.7$   
163 °C (IPMA ( Instituto Portugêus do mar e da Atmosfera), n.d.). The soil is a cambisol (FAO-  
164 UNESCO, 1974), with 81% sand, 5% clay and 14% silt. Tree density is 177 tree ha<sup>-1</sup> and tree  
165 leaf area index (LAI) is 1.6 m<sup>2</sup> m<sup>-2</sup> (Correia et al., 2014). At the beginning of the measurement  
166 period the mean height below the canopy and total height of trees was 3.1 and 7.9 m,  
167 respectively and the mean diameter at breast height was 24.7 cm (Costa-e-Silva et al., 2015).  
168 Cork oak canopy cover at the site is 36% and the understory layer is composed by a mixture of  
169 shrubs and herbaceous species. The site was ploughed in 2009 (Correia et al., 2016) and 2013,  
170 hence the shrubs cover fraction changed across years. A field survey in 2017 measured a 18%  
171 coverage of shrubs and 41% of herbaceous species, with litter and bare soil covering the  
172 remaining area (Heuschmidt et al., 2020). *Cistus salviifolius* and *Ulex airensis* are the most  
173 represented shrubs species, while grasses (44.5%) and legumes (28.7%) dominate the  
174 herbaceous layer (Xavier Lecomte, personal communication).

175

176 *2.2 Climatic drivers and ecosystem carbon fluxes*

177 Carbon dioxide (CO<sub>2</sub>) and water vapor fluxes were measured continuously by eddy covariance  
178 from September 2010 to August 2017. The equipment is positioned at the top of a 22 m tower  
179 and consists of a 3D sonic anemometer (R3, Gill Instruments Ltd., Lymington, England) and a  
180 closed-path infrared gas analyser (IRGA, Li 7000, Li-Cor Inc., Lincoln, NE, USA).  
181 Precipitation (ARG100, Environmental Measurements Ltd., Gateshead, UK), solar radiation  
182 (BF2, Delta-T Devices Ltd., Cambridge, UK), humidity and air temperature (CS215, Campbell  
183 Scientific, Inc., Logan, UT, US) were collected by a data logger (CR1000, Campbell Scientific,  
184 Inc., Logan, UT, USA) and averaged at a 30-minutes time step (Costa-e-Silva et al., 2015).



185 Flux data were processed with the EddyPro software (version 6.2.0, LI-COR Biosciences Inc.,  
186 Lincoln, NE, USA). Raw data were first despiked according to Vickers and Mart (1997). Time  
187 lags were then determined by a procedure accounting for the dependency of water vapor lags to  
188 relative humidity. Coordinates rotation followed the planar fit method (Wilczak et al., 2001) for  
189 the main wind directions between 15°-90° and 200°-330°, while the double rotation was applied  
190 for wind directions comprised between 90°-200° and between 330°-15°. With this approach we  
191 tested the influence of the potential flux contribution of a rain-water-harvesting lake and an  
192 eucalypt area in the 80% iso-line of the footprint climatology (Kljun et al., 2015). The estimate  
193 of the  $u^*$  threshold (Papale et al., 2006), gap filling and flux partitioning were performed using  
194 the REddyProc R package (Wutzler et al., 2018). The average  $u^*$  threshold was 0.32 ( $\pm 0.030$ ).  
195 After correction of CO<sub>2</sub> fluxes for storage (using the one-point storage correction), the NEE was  
196 partitioned into GPP and Reco according to Reichstein et al. (2005). We follow the convention  
197 that negative NEE represents a carbon sink and positive NEE represents a carbon source at the  
198 surface, thus,  $GPP = -NEE + Reco$ . Further details on eddy covariance data processing and the  
199 dataset used are openly available (Cerasoli et al., 2020).

200 The soil water content (SWS) was calculated at daily time step adopting the one-layer bucket  
201 concept (Manabe, 1969). At each day ( $t$ ),

$$202 \text{SWS}(t) = \text{SWS}(t-1) + R(t) - E(t);$$

203 Being the recharge  $R(t) = \min[P(t), \text{SWS}_{\max} - \text{SWS}(t-1)]$

204

205 and the daily evapotranspiration  $E(t)$  calculated from latent heat values.

206 The maximum water content ( $\text{SWS}_{\max}$ ) or field capacity was estimated to be 200mm and the  
207 permanent wilting point, the lower value of SWS that still permits water extraction by plants,  
208 was fixed at 7.5mm with basis on 10cm depth soil measurements done in previous studies  
209 carried out at the site (Jongen et al., 2013).

210

### 211 *2.3 Extraction of seasonality parameters*

212 Phenological parameters were estimated from remote-sensed time-series data, Here, we used  
213 the TIMESAT software package (Jönsson and Eklundh, 2002, 2004) to analyse a time-series of

214 NDVI (Tucker, 1979) data of the study area derived from quality-controlled MODIS bands. The  
215 daily MODIS collection MCD43A4, at 500m spatial resolution (Schaaf and Wang, 2015a), was  
216 downloaded from the Google Earth Engine (GEE) platform (Gorelick et al., 2017)  
217 (<https://earthengine.google.com>) for a 2km radius circular area centred around the tower  
218 location. The MCD43A2 product (Schaaf and Wang, 2015b) was used to control for the quality  
219 of the BRDF inversion for each surface reflectance band and filter out observations with clouds  
220 and water coverage. A mean value of the good quality pixels was then obtained for each  
221 MODIS band used to calculate NDVI (Supplement material, fig. S1).  
222 The program first fits smooth functions to the data, in this study the Savitzky-Golay filter was  
223 selected (Savitzky and Golay, 1964), and once the fitting has been achieved, it computes  
224 seasonality parameters such as the start (SOS), the end (EOS) and the length of the growing  
225 season length (GSL). An anticipation of the SOS and/or a delay in the EOS result in an increase  
226 of GSL. These parameters give important information about short- and long-term vegetation  
227 changes, and their dependency on climate, providing insights on the functional and structural  
228 characteristics of the ecosystem. Despite the complex multilayer composition of vegetation at  
229 the site, we are confident that the temporal NDVI trend realistically represents the phenology of  
230 the herbaceous understory.  
231 Indeed, spectral measurements of the different vegetation types, collected *in-situ* and for the  
232 same period considered in this study, showed a marked NDVI seasonal trend of the herbaceous  
233 layer and very constant values for the evergreen cork oak canopy (Soares et al., 2022).  
234 In addition, a simulation analysis, revealed that even when the tree density is high, the  
235 variability of MODIS NDVI is mainly driven by the herbaceous layer (Häusler et al., 2016).

236

#### 237 2.4 Data analysis

238 Meteorology and carbon flux data from 2010 to 2017 were analysed and hereafter reported by  
239 hydrological year, considered running from September 1 to August 31 based on the rainfall  
240 pattern in the region. For seasonal analyses, we considered the following periods: autumn  
241 (September, October, November), winter (December, January, February), spring (March, April,

242 May) and summer (June, July August). A 30-day moving average was used to visualize  
243 temporal patterns of carbon fluxes, daily mean temperature and global radiation.

244 The occurrence of time lags between rainfall and carbon fluxes (GPP, Reco and NEE) was  
245 addressed by computing the correlation between 12 months cumulative values of the variables  
246 differing one month from each other. The multiple “yearly” values of carbon fluxes obtained  
247 were correlated with those of rainfall considering different potential time lags by using a  
248 moving time window ranging from 0 to 12 months and shifted three months at a time (Marcolla  
249 et al., 2011).

250 All statistical analyses were performed using R (R CoreTeam, 2016). One-way ANOVAs were  
251 performed to evaluate the significance of variance between seasons and years of GPP, Reco and  
252 NEE measurements. Conditions of homoscedasticity and normality were always verified by  
253 visual inspection of residuals. To estimate which climate variable controls SOS and EOS of the  
254 growing season of the understory vegetation we correlated total precipitation and the average  
255 maximum vapor pressure deficit to each phenological date using both linear and quadratic  
256 models. Models with the greatest coefficient of determination (adjusted  $R^2$ ) were selected. The  
257 time span considered for correlation with the SOS was September-October since data showed  
258 that SOS was always comprised between the end of September and the beginning of October.  
259 For EOS both seasonal and bi-monthly time spans were used. Only regression with  $p \leq 0.05$  were  
260 considered. The inter-annual anomalies of GSL ( $\Delta$ GSL) have been calculated as the difference  
261 between values estimated from a regression line fitted to all years and the observed annual  
262 values. Linear regression models were applied to estimate the impact of  $\Delta$ GSL on carbon fluxes  
263 (GPP, Reco and NEE).

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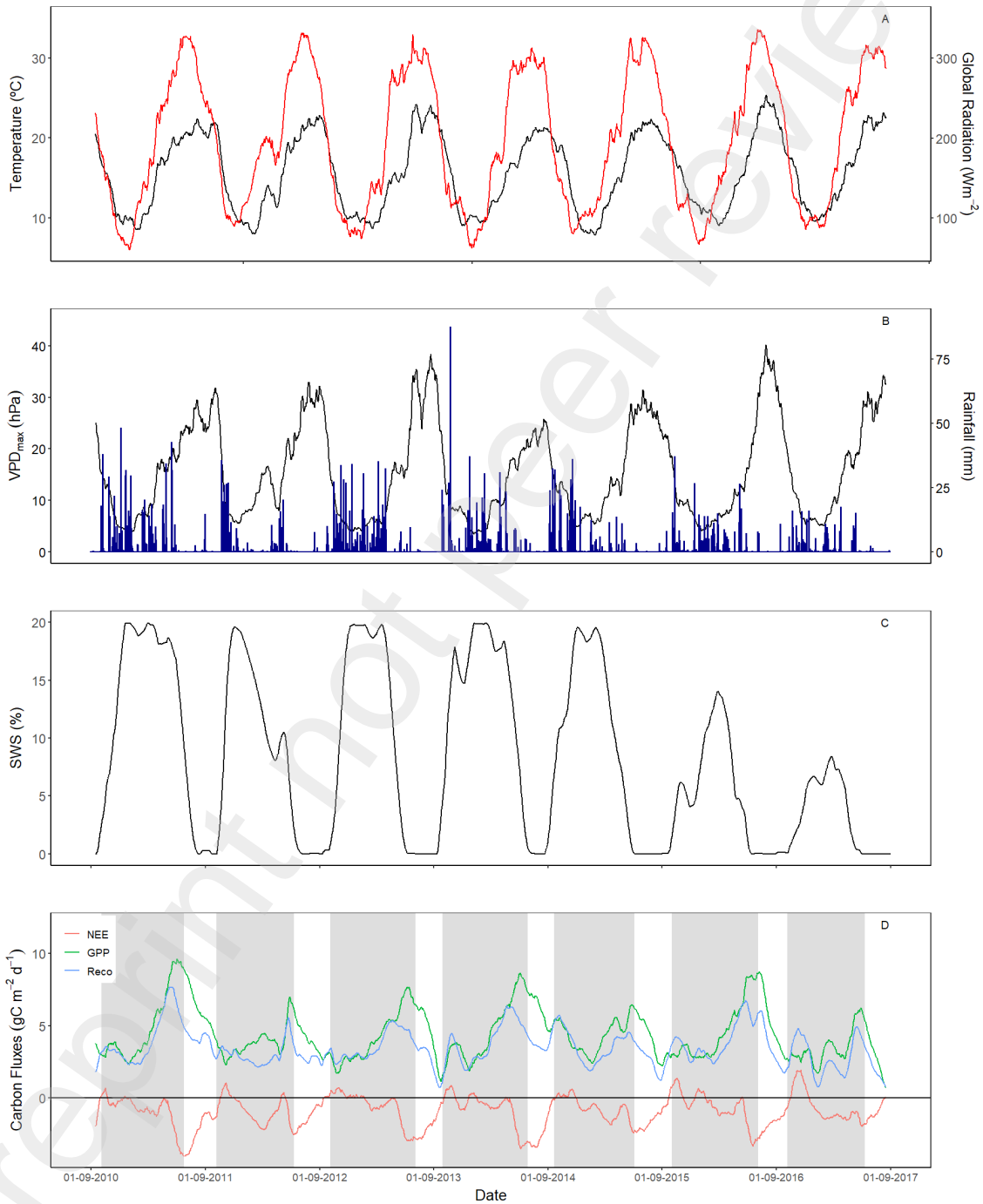
269 **3. Results**

270

271 *3.1 Time series of CO<sub>2</sub> fluxes and meteorological variables*

272 In spite of a moderate difference among records of average daily temperature within the

273 analysed period ( $T_{min} = 10.9 \pm 4.5$  °C;  $T_{max} = 21.4 \pm 7.2$ °C;  $T_{mean} = 15.7 \pm 5.5$  °C for 2010-



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275

276 **Fig.1** Daily values of average temperature (black line, °C), average global radiation (red line, W  
277 m<sup>-2</sup>) (A), maximum VPD (black line, hPa) and total Rainfall (blue bars, mm) (B), soil water  
278 storage (%) (C) and Carbon Fluxes (g C m<sup>-2</sup> d<sup>-1</sup>) (D). With the exception of rainfall all variables  
279 are represented as 30-days moving averages. Shaded areas (D) represent the growing seasons  
280 length (GSL) of the herbaceous understory. Data shown have been recorded at the site between  
281 September 2010 and August 2017.

282

283 2017) a marked seasonality was observed (Fig. 1A). The maximum and the minimum daily  
284 average temperatures were always recorded in summer and winter, respectively. Daily average  
285 temperature and global radiation (Fig. 1A) were highly correlated ( $R^2=0.71$ ;  $p$ -value  $<0.001$ ).  
286 Total rainfall was generally concentrated in autumn and winter (Fig. 1B). On the opposite, the  
287 maximum vapor pressure deficit ( $VPD_{max}$ ) was higher during the summer period (Fig. 1B), in  
288 agreement with the temperature trend observed. The Soil Water Storage (SWS, Fig.1C)  
289 followed roughly seasonal rainfall patterns, being higher in winter and decreasing sharply in  
290 summer to recover only after the occurrence of first rains in September. The extent of the dry  
291 summer period (consecutive days with  $SWS < 7.5\%$ ) varied among years, ranging from 80 days  
292 in 2014 to 149 in 2016. The impact of low winter rainfall on the SWS is particularly evident in  
293 the last two years of the study when the SWS never reach its maximum.

294 Ecosystem carbon fluxes also showed a marked seasonality (Fig 1D), typical of Mediterranean  
295 oak woodlands. Each year, a tendency for higher NEE with the onset of spring was observed,  
296 while GPP and Reco showed a positive trend. At the end of the herbaceous understory growing  
297 season (shaded areas, Fig. 1D), the NEE trend was generally inverted as a result of a drop in  
298 GPP and Reco towards summer and beginning of autumn. During these periods, Reco flux rates  
299 were occasionally higher than GPP causing NEE to approach zero or to become slightly positive  
300 and the ecosystem to temporarily function as a carbon source.

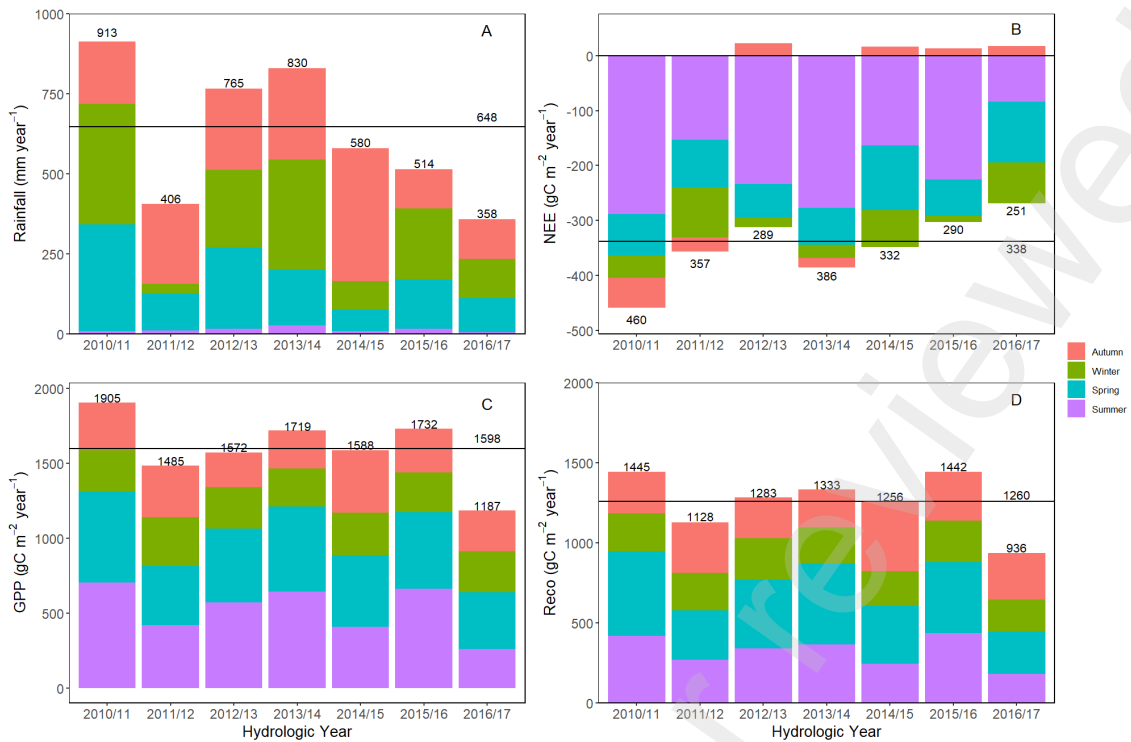
301

302 *3.2 Inter-annual and seasonal variability*

303 A large inter-annual variability was observed for rainfall ( $p$ -value $<0.001$ ), with autumn, winter  
304 and spring rainfall being highly variable between years (Fig. 2A). Total annual rainfall during  
305 the study period was 43%, 18% and 28% higher than the long-term average of 623 mm in  
306 2010–2011 (928 mm), 2012-2013 (765 mm) and 2013-2014 (830 mm), respectively, and 37%,  
307 22%, 21% and 44% lower in 2011–2012 (408 mm), 2014-2015 (580 mm), 2015-2016 (511 mm)  
308 and 2016-2017 (360 mm), respectively.

309 A high inter-annual variability was also observed in carbon fluxes. The average annual NEE  
310 was  $-338 \pm 70 \text{ g C m}^{-2}\text{y}^{-1}$ , ranging from  $-460 \text{ gCm}^{-2}\text{y}^{-1}$  to  $-251 \text{ gCm}^{-2}\text{y}^{-1}$  (Fig. 2B). There were  
311 significant differences among years ( $p$ -value $<0.001$ ), but the ecosystem acted as a carbon sink  
312 (i.e. annual NEE $<0$ ) in all the seven years of the study.

313 The annual GPP was on average  $1598 \pm 227 \text{ gCm}^{-2}\text{y}^{-1}$  ranging from  $1187 \text{ gCm}^{-2}\text{y}^{-1}$  to  $1905$   
314  $\text{gCm}^{-2}\text{y}^{-1}$ , while average annual Reco was  $1260 \pm 181 \text{ gCm}^{-2}\text{y}^{-1}$ , ranging from  $936 \text{ gCm}^{-2}\text{y}^{-1}$  to  
315  $1445 \text{ gCm}^{-2}\text{y}^{-1}$ . The maximum and minimum values of NEE, GPP and Reco were observed in  
316 2010/2011 and 2016/2017, respectively. Annual GPP showed significant inter-annual variability  
317 ( $p$ -value $<0.001$ ), with some years being more productive (2010/11, 2013/14 and 2015/16) than  
318 the average and a clear tendency of summer and spring to be the most productive seasons of the  
319 year (Fig. 2C).



320

321 **Fig.2** Cumulative values of Rainfall (A, mm year<sup>-1</sup>), NEE (B, gC m<sup>-2</sup> year<sup>-1</sup>), GPP, (C, gC m<sup>-2</sup>  
 322 year<sup>-1</sup>) and Reco (D, gC m<sup>-2</sup> year<sup>-1</sup>) All variables are grouped by hydrological year and  
 323 partitioned into seasons. Black line represents the average over the study period.

324

325 At the annual scale, Reco showed patterns similar to those observed for GPP (Fig. 2D) and the  
 326 two fluxes were strictly correlated ( $r=0.97$ ). However, the degree of correlation between daily  
 327 sums of GPP and Reco were different among seasons. A tight GPP-Reco relationship was  
 328 observed in summer ( $r=0.89$ ) and spring ( $r=0.76$ ), while it was weaker in autumn ( $r=0.49$ ) and  
 329 winter ( $r=0.40$ ), suggesting a potential GPP-Reco decoupling in these seasons. We are aware  
 330 that the GPP and Reco relationship might be partly spurious because it is derived from the flux  
 331 partitioning methods. Therefore, we used night-time NEE in place of Reco and midday NEE in  
 332 place of GPP as an alternative proxy to test for the relationship between the two biological  
 333 processes. The obtained results showed a less tight but consistent correlation than the ones  
 334 reported between GPP and Reco fluxes ( $r=0.69, 0.31, 0.19$  and  $0.005$  for summer, spring,  
 335 autumn and winter respectively), giving us confidence that the relationship between

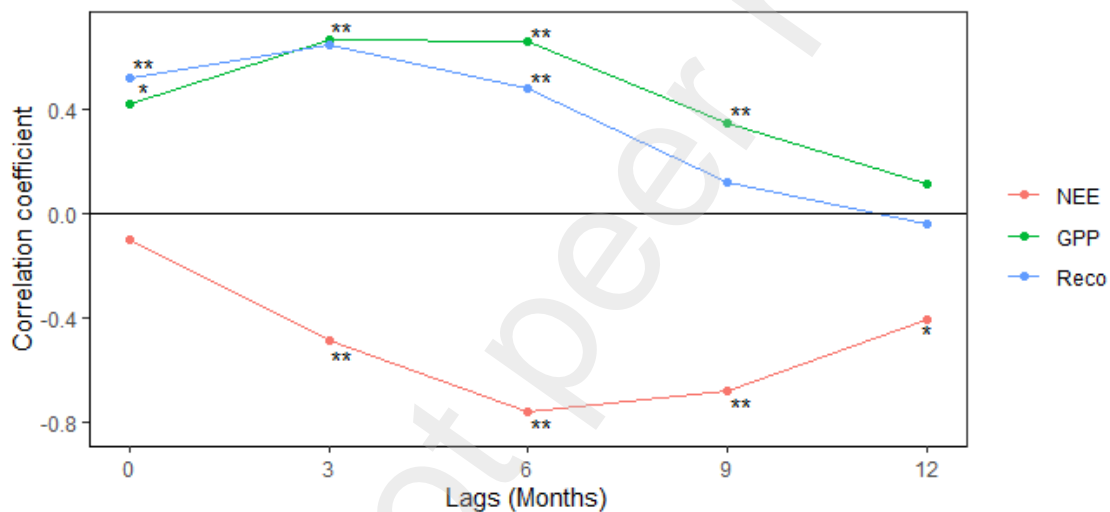
336 photosynthesis and ecosystem respiration found is not impacted by the potential spurious  
337 correlation between GPP and Reco.

338 Spring and summer contributed more to the annual NEE, while autumn showed the largest  
339 variability, with the ecosystem acting as a carbon source in autumn of 2012/2013, 2014/2015,  
340 2015/2016 and 2016/2017.

341

### 342 3.2 Rainfall and carbon fluxes

343 The sign, magnitude and delay of the response of carbon fluxes to rainfall were investigated by  
344 the analysis of time-lagged correlations between NEE, GPP, Reco and rainfall (Fig. 3).



345

346 **Fig.3** Correlation coefficients observed between annual carbon fluxes and annual rainfall  
347 repeated at different time lags in which rainfall precedes carbon fluxes of 0 to 12 months

348 \*Correlations significant at the 95 %; \*\*correlations significant at 99 %.

349

350 At 0 and 3 months' time-lag, GPP and Reco correlations with rainfall were significant and of  
351 similar magnitudes. At longer time lags, the Reco-rainfall correlation was weaker than that  
352 observed for GPP-rainfall but still significant at 6 months' time-lag. For GPP the degree of  
353 correlation decreased only after 6 months' time-lag onward and was still significant at a 9  
354 months' time-lag. As a result differences were observed between GPP-Rainfall and Reco-  
355 Rainfall correlations. The association between NEE and Rainfall was not significant at 0 time-



356 lag but increased progressively at 3 and 6 months' time lag, when reached a peak and then  
357 decreased at 9 and 12 months' time-lag but maintaining a significant degree of correlation.

358

### 359 3.4 Growing Season length

360 Growing season start (SOS), end (EOS), and length (GSL) obtained by the analysis of NDVI  
361 time series are represented as grey shades in Figure 1B and summarized in Table 1. We  
362 observed high inter-annual variability in all the parameters, with differences in the SOS of about  
363 16 days varying from the day of the year (DOY) 263 in 2014/2015 to the DOY 279 in  
364 2011/2012 and 2016/2017. In 2014/2015 the earlier SOS coincided with an anomalous high  
365 precipitation recorded in September 2014. The variability of the EOS was 31 days, larger than  
366 that of the SOS ranging from DOY 154 in 2014/2015 to DOY 185 in 2015/2016. Similarly,  
367 GSL showed an inter-annual variability of 29 days ranging from 247 days (2011/2012 and  
368 2016/2017) to 276 days (2015/2016).

369

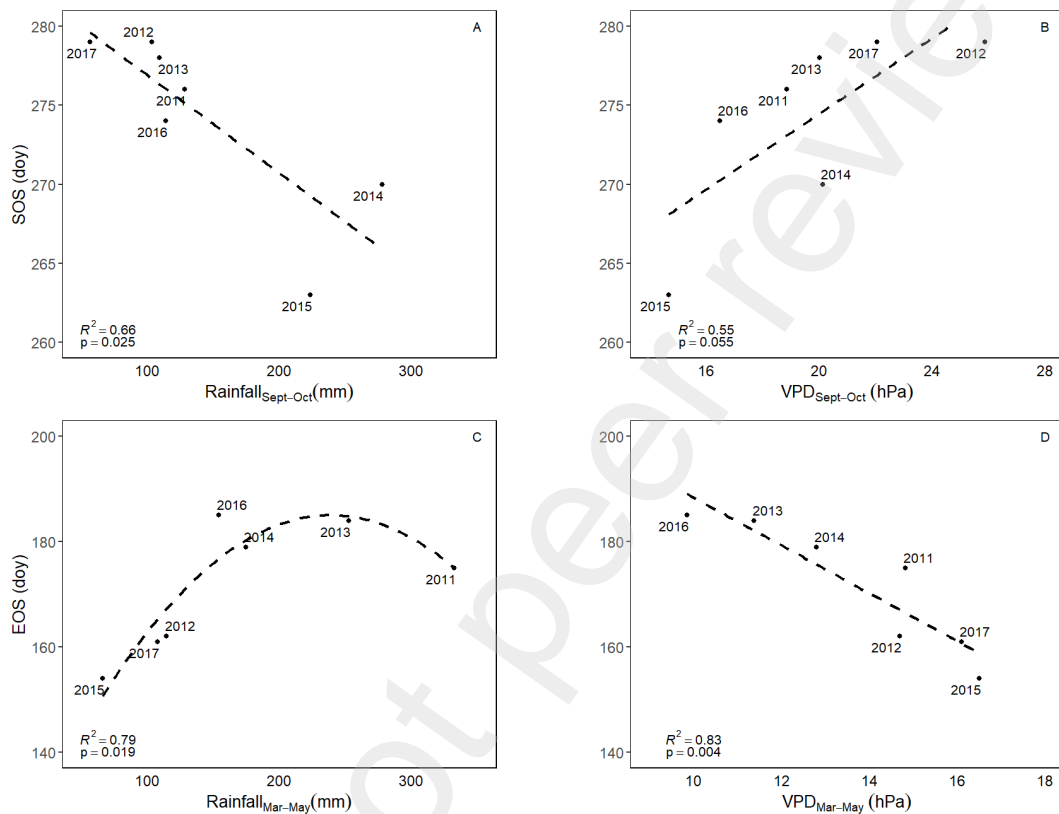
370 **Table 1** Seasonality parameters extracted from TIMESAT analysis of NDVI time series: start of  
371 the growing season (SOS, doy), end of the growing season (EOS, doy) and the growing season  
372 length (GSL, n° of days).

Hydrologic year	GSL (n° of days)	SOS (doy)	EOS (doy)
2010/2011	264	276	175
2011/2012	247	279	162
2012/2013	271	278	184
2013/2014	274	270	179
2014/2015	256	263	154
2015/2016	276	274	185
2016/2017	247	279	161

373

374 The dependency of SOS and EOS on rainfall and vapor pressure deficit ( $VPD_{max}$ ) was tested by  
375 regression analyses for different time periods preceding the occurrence of the event. Over the

376 study period, the SOS was clearly influenced by rainfall at the beginning of the hydrologic year  
 377 (September-October) (figure 4, A). being delayed when it was scarce and anticipated under  
 378 opposite conditions. On the contrary, the daily average of the maximum vapor pressure deficit  
 379 over the same period (VPDmax) showed negative trend with SOS, resulting in a delay of the  
 380 SOS as VPD increases (figure4, B).



381  
 382 **Fig. 4** Relationships between the Start of the season (SOS) and September –October total  
 383 rainfall (A), and vapor pressure deficit (B); relationship between the End of the Season (EOS)  
 384 and March-May total rainfall (C) and vapor pressure deficit (D).

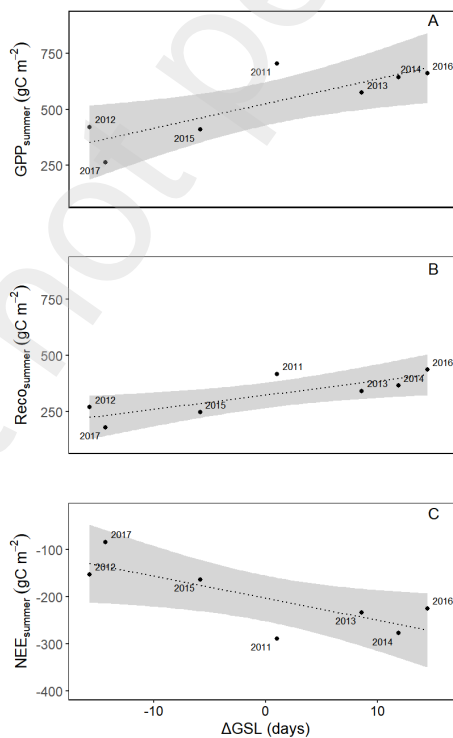
385  
 386 Opposite trends to the observed for the SOS were observed assessing the sensitivity of EOS to  
 387 climate variables. The EOS was found significantly sensitive to Rainfall and VPD in Spring  
 388 (March-May) (Fig.4, C and D) just before the onset of senescence. The quadratic regression  
 389 observed indicates that the dependency of EOS on rainfall, is particularly strong for rainfall  
 390 values up to about 230 mm when the fitted curve reached its maximum while at higher rainfall  
 391 values the sensitivity seems to decrease (Fig. 4, C). Instead, the sensitivity of the EOS to VPD

392 in spring was linear, showing a sharp anticipation in the EOS as VPD increases. EOS was  
 393 particularly anticipated in 2015 which exhibited the lower rainfall and higher VPD in Spring.  
 394 Noticeably a significant regression was also verified between EOS and total Rainfall recorded in  
 395 January and February ( $y=0.125R+155$ ,  $R^2=0.642$ ,  $p=0.03$ ). Any other time period did not result  
 396 in any significant correlation.

397 **Table 3.** Regressions between GPP, Reco and NEE and the inter-annual variability of the  
 398 growing season length ( $\Delta$ GSL). Fluxes are accumulated values by hydrologic year (year) and by  
 399 summer.

	year	summer
GPP	$12.3\Delta$ GSL+1598; $R^2=0.44$ , $p=0.10$	$11.04\Delta$ GSL+526; $R^2=0.69$ , $p=0.021$
Reco	$11.7\Delta$ GSL+1260; $R^2=0.63$ , $p=0.033$	$6.34\Delta$ GSL+322; $R^2=0.68$ , $p=0.022$
NEE	$-0.62\Delta$ GSL-338; $R^2=0.01$ , $p=0.818$	$-4.70\Delta$ GSL-204; $R^2=0.66$ , $p=0.036$

400



401

402 **Fig. 5** Relationship observed between GPP (a), Reco (b) and NEE (c) (gCm<sup>-2</sup>) accumulated  
 403 during the summer period and the anomalies of the growing season length ( $\Delta$ GSL). The labels

404 refer to the year. Equations are reported in table 3. The shaded areas represent the 95%  
405 confidence interval.

406

407 Differences in the SOS and EOS resulted in inter-annual differences of the GSL. We verified  
408 the impact of the inter-annual anomalies of the growing season length ( $\Delta$ GSL) on carbon fluxes  
409 at annual and seasonal time scale. Carbon fluxes accumulated during the summer period were  
410 significantly impacted by  $\Delta$ GSL . Each additional day of the GSL resulted in an additional  
411 uptake of  $11\text{gC m}^{-2}$  (Table 3, Fig.5A.), an extra respiratory loss of  $6.3\text{gC m}^{-2}$  (Table 3, Fig.5B)  
412 and thus an additional net carbon sequestration of  $4.70\text{gC m}^{-2}$  (Table 3, Fig.5C).

413 At annual scale, the impact of the  $\Delta$ GSL on GPP and Reco was similar, but Reco was almost  
414 twice than the observed in summer, indicating differences in the GPP/NEE ratio along the year.  
415 As a consequence of the GPP-Reco decoupling, GSL has no influence on the annual NEE.

416

#### 417 **4. Discussion**

418

##### 419 *4.1 Inter-annual and seasonal variability*

420 The study site showed a considerable ability to act as carbon sink. The observed average annual  
421 NEE ( $-338\text{gCm}^{-2}\text{y}^{-1}$ ) was in agreement with what was previously observed at the same site  
422 (Costa-e-Silva et al., 2015) but higher than values recorded at other similar ecosystems. For  
423 example, in a mixed cork (*Quercus suber* L.) and holm oak (*Quercus ilex* spp. rotundifolia) in a  
424 southern Portuguese site, NEE ranged between  $-140$  and  $-28\text{gCm}^{-2}\text{y}^{-1}$  (Pereira et al., 2007). In  
425 two other holm-oak woodland sites in the nearby region of Extremadura (Spain) the average  
426 annual NEE recorded was  $75$  and  $15\text{gCm}^{-2}\text{y}^{-1}$  (El Madany, 2020) and in a Californian site, NEE  
427 was about  $-110\text{gCm}^{-2}\text{y}^{-1}$  (Ma et al., 2016).

428 Differences in climate and vegetation among sites make it difficult to discern reasons for the  
429 higher carbon sequestration observed in this study as compared to other Mediterranean oak  
430 woodlands. A possible distinguishing feature could be the higher evergreen tree canopy cover

431 (around 40% in this study, against 19-24% in the other mentioned studies) and consequently a  
432 higher contribution of tree photosynthesis to the overall GPP. Cork oak trees are drought  
433 avoiding species, relying on deep roots and a tight stomatal control to escape summer drought  
434 conditions (David et al., 2016). This efficient strategy allows trees to recover rapidly  
435 transpiration and carbon assimilation rates when drought conditions are alleviated (Caldeira et  
436 al., 2015). Hence, a higher tree density can contribute to maintain a positive carbon  
437 sequestration (negative NEE), as compared to sites where the understory, dependent on shallow  
438 soil water, represents a larger fraction of the overall GPP. The presence of semi-deciduous and  
439 evergreen shrubs in the understory could also have contributed to maintain GPP during summer  
440 dry periods (Correia et al., 2014). However, shrubs were regularly removed (every 3-4 years)  
441 and their contribution varied across time.

442 Despite the consistent annual carbon sequestration, we observed significant inter-annual and  
443 seasonal differences in NEE, GPP and Reco. Spring and summer carbon assimilation  
444 contributed most to the annual GPP, similar to what was observed in a Californian savanna-type  
445 ecosystem (Ma et al., 2016). Summer was also the season in which we observed the largest  
446 inter-annual variability in NEE and a tight GPP-Reco coupling, suggesting that Reco was  
447 strictly dependent on GPP during that season and that the observed variability in summer NEE  
448 was related mainly to constraints in photosynthetic carbon assimilation.

449 In four of the seven years analysed, NEE became positive in autumn and the ecosystem  
450 switched then to a carbon source. A positive NEE was observed in years with below average  
451 summer rainfall (2012/2013, 2015/2016 and 2016/2017), suggesting that the critical depletion of  
452 soil moisture was responsible for the observed decrease of GPP and in a lesser extent of Reco at  
453 the beginning of the hydrological year.

454 A positive NEE in the autumn season was also observed under opposite conditions, more than  
455 average rainy summer and autumn (2014/2015). In this year, both GPP and Reco were above  
456 the long-term average but the proportion was slightly higher for Reco than GPP (see section  
457 3.2). The increased proportion of Reco could be partly ascribed to a high heterotrophic  
458 respiration at the beginning of the autumn season (El-Madany, 2020). A similar pattern was also

459 observed under low water availability conditions, particularly under isolated rain pulses (Fraser  
460 et al., 2016; Jarvis et al., 2007).

461 A low degree of correlation between GPP and Reco was also observed in winter. In winter  
462 months, when usually soil water is not a limiting factor, GPP can be constrained by low  
463 radiation and temperature. On the opposite, Reco is driven mainly by temperature and hence a  
464 decoupling between GPP and Reco can occur.

465

#### 466 *4.2 The long-term impact of Rainfall*

467 The importance of rainfall for ecosystem survival in semi-arid ecosystems is unquestionable and  
468 several studies have highlighted that carbon fluxes in Mediterranean ecosystems are strongly  
469 affected by the erratic seasonal and inter-annual distribution of rainfall events (Almagro et al.,  
470 2009; Jongen et al., 2011; Pereira et al., 2007; Poulter et al., 2014; Reichstein et al., 2002; Ross  
471 et al., 2012). Our results show that rainfall has a time-lagged impact on NEE that extend up to a  
472 12 months delay. As observed in other studies (Marcolla et al., 2011), the time-lagged effect  
473 was stronger than the concurrent effect. The strongest correlation between rainfall and carbon  
474 fluxes was observed at 3 and 6 -month time lags. The reason behind this long time-lagged effect  
475 can be partly explained by the existence of a water table reservoir accessible by deep oak roots  
476 that likely promotes inter-seasonal rainfall mixture (Costa-e-Silva et al., 2015). In addition to  
477 the ability to extract water from deep soil layers (Costa-e-Silva et al., 2015; Pinto et al., 2014),  
478 cork oak trees can also redistribute it to the soil surface through the hydraulic lift (Pinto et al.,  
479 2014). This process alleviates summer stress conditions and can decrease the ecosystem  
480 dependency on recent rainfall (David et al., 2016). Indeed, in woodlands with a lower tree cover  
481 fraction, GPP was found largely dependent on short term rainfall (Luo et al., 2020, 2018). An  
482 analysis of past conditions' influence on ecosystem carbon exchange across different ecosystem  
483 types, revealed that about 32% of the variation in NEE was explained by previous conditions  
484 and that environmental memory was primarily driven by soil moisture (Liu et al., 2019). Such  
485 time-lagged effects could enhance the ecosystem's resistance to more frequent climate extremes  
486 and deserves a better consideration in the evaluation of ecosystem stability. A deeper knowledge

487 of lag effects could advance our understanding of the mechanisms responsible for climate  
488 change effects on ecosystem carbon fluxes.

489 We observed an almost parallel trend in GPP and Reco correlation with rainfall as the time lag  
490 increased (Fig.4). However, while at short time lags (0 and 3 months) the correlation  
491 coefficients were very similar for the two fluxes, beyond a 6-month lag the correlation was  
492 consistently higher for GPP. The presence of the water table may also explain this observation  
493 by acting as a long-term reservoir of rainfall and allowing higher GPP rates while Reco, which  
494 is partly dependent on heterotrophic respiration, would be more strongly related with short term  
495 rainfall (0 and 3 months). These different dependencies are in agreement with the observed  
496 GPP-Reco decoupling.

497

#### 498 *4.3 The Relationship between rainfall, growing season length and carbon fluxes*

499 In this study, remotely-sensed data were used to retrieve the start (SOS), the end (EOS) and the  
500 length (GSL) of the growing season of the understory vegetation. Even though multi-layer  
501 vegetation strata are present at the site, we are confident that the extracted seasonality  
502 parameters represent changes in the annual growing seasonal cycle of the herbaceous vegetation  
503 and semi-deciduous shrubs. In fact, previous analyses of in-situ spectral retrievals data collected  
504 at the same site, showed a fairly constant NDVI trend of cork oak canopy across seasons and  
505 years but a large NDVI inter-annual variability for the herbaceous layer, in agreement with the  
506 variability of the phenological parameters observed in the present study (Cerasoli et al., 2016;  
507 Soares et al., 2022). The GSL observed at our site was also in line with an average GSL of 200-  
508 300 days for savanna ecosystems reported in a global synthesis from Baldocchi et al. (2008).

509 Both rainfall and VPD recorded in the months previous to the phenological events showed  
510 clearly to influence both the start (SOS) and the end of the season (EOS). The dependency of  
511 the SOS on rainfall observed at the beginning of the hydrological year (September- October)  
512 was already reported in other studies in similar ecosystems and climate conditions (Luo et al.,  
513 2020; Nogueira et al., 2017). Equally, the dependency of the EOS on spring rainfall and VPD  
514 was reported in previous studies (Marchin et al., 2018).

515 In our study, a larger variability of the end of the season (EOS) than the start of the season  
516 (SOS) dates was observed, emphasizing the high sensitivity of the understory to summer  
517 drought conditions. Ma and colleagues (2007) reported that the lack of precipitation in the  
518 spring (April, May and June) was one of the predominant factors affecting the length of the  
519 growing season in a oak-grass savanna in California.

520 Similarly to what reported in other studies (Marchin et al., 2018), our results also evidence the  
521 role of VPD to predict the timing of grass green-up (SOS) and yellowing (EOS). The VPD is a  
522 measure of the atmosphere humidity which is known to be related to evapotranspiration and  
523 hence also to the physiological response of vegetation to soil water deficit. Here, the strong and  
524 linear correlation observed between EOS and VPD evidence the sensitivity of the herbaceous  
525 layer to atmospheric humidity and temperature and their influence on the timing of phenological  
526 events.

527 In agreement with previous studies at the same site (Correia et al., 2016; Piayda et al., 2014),  
528 the EOS was also found dependent on winter rainfall suggesting a delay in the impact of  
529 precipitation also for the growing season length of the herbaceous understory

530 The relevance of winter rainfall is probably a consequence of the annual rainfall distribution,  
531 mainly concentrated in winter, and hence contributing most to the overall annual rainfall. On the  
532 contrary, summer precipitation represents only 1-3% of the total annual rainfall. Hence, the  
533 strong EOS-rainfall correlation observed in this period highlights the importance of  
534 precipitation timing for the onset of senescence of the understory vegetation.

535 The tested hypothesis that the duration of the GSL could significantly impact annual carbon  
536 fluxes was only partially confirmed. The increase of the GSL observed at our site leads to  
537 positive anomalies of the annual gross fluxes (GPP and Reco) but not of NEE. In fact, the  
538 relationship NEE- $\Delta$ GSL at the annual scale was not significant neither strong ( $0.6 \text{ gC m}^{-2}$ , Table  
539 3). This is in contrast with previous studies that showed a significant relationship between GSL  
540 and NEE for savanna and forest ecosystems (Baldocchi, 2008; Richardson et al., 2010).



541 However, the impact of  $\Delta$ GSL was significant on summer fluxes. These results confirm the  
542 importance of precipitation timing, indicating that a small amount of precipitation can delay the  
543 onset of senescence resulting in a considerable gain in ecosystem GPP and NEE.

544

## 545 **5. Conclusions**

546 In this study we evaluate the long-term impact of rainfall on a cork oak woodland carbon fluxes.

547 This knowledge can contribute to foresee the ecosystem's response to climate change conditions  
548 and support forest management actions in order to maximize productivity and ecosystem  
549 services. We argue that the long time lag between rainfall and carbon dynamics was determined  
550 mainly by the high tree density, which may confer long-term stability to the ecosystem.

551 However, progressive drought conditions reported in recent years have been responsible for an  
552 increase in tree mortality and decline (Camilo-Alves et al., 2017) and further studies should  
553 focus on the ability of trees to overcome expected exacerbated droughts.

554 Our results also highlight the sensitivity of the understory to the rainfall regime and its  
555 associated impact on GPP and Reco. Other studies, employing stable isotopes, demonstrated  
556 that the understory can contribute considerably (50%) to the whole GPP and, in spite of a high  
557 sensitivity to water shortage, can have a beneficial effect on the resilience of the ecosystem to  
558 drought, favouring, for example, soil water infiltration (Dubbert et al., 2014). This shows the  
559 importance of considering the interactions between over- and understory to maximize oak  
560 woodlands productivity and carbon sequestration.

561 Overall, our results emphasize the need for a better understanding of the sensitivity of trees and  
562 understory vegetation to changes in rainfall regime and underline the importance of considering  
563 legacy effects on carbon fluxes. The multi-temporal scale approach appears essential in  
564 understanding climate drivers and improving our ability to foresee the impacts of forthcoming  
565 climate change on ecosystem carbon exchange.

566

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580

#### 581 **Declaration of Competing Interest**

582 The authors declare that they have no known competing financial interests or personal  
583 relationships that could have appeared to influence the work reported in this paper.

584

#### 585 **Database**

586 Publicly available at <https://zenodo.org/record/3727798#.Xqh9dJnOXIV>

587

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