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1 **Using the dual approach of FAO-56 for partitioning ET into soil and plant**
2 **components for olive orchards in a semi-arid region**

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24 **Abstract**

25 The main goal of this research was to evaluate the potential of the dual approach of
26 **FAO-56 for estimating actual crop evapotranspiration (AET) and its components (crop**
27 **transpiration and soil evaporation) of an olive (*Olea europaea* L.) orchard in the semi arid**
28 **region of Tensift-basin (central of Morocco). Two years (2003 and 2004) of continuous**
29 **measurements of AET with the eddy covariance technique were used to test the performance**
30 **of the model.** The results showed that, by using the local values of basal crop coefficients, the
31 approach simulates reasonably well AET over two growing seasons. The Root Mean Square
32 Error (RMSE) between measured and simulated AET values during 2003 and 2004 were
33 respectively about 0.54 and 0.71 mm per day. The obtained value of basal crop coefficient
34 (K_{cb}) for the olive orchard was similar in both seasons with an average of 0.54. This value
35 was lower than that suggested by the FAO-56 (0.62). **Similarly, the single approach of FAO-**
36 **56 has been tested in the previous work (Er-Raki et al., 2008) over the same study site and it**
37 **has been shown that this approach also simulates correctly AET when using the local crop**
38 **coefficient and under no stress conditions.**

39 Since the dual approach predicts separately soil evaporation and plant transpiration, an
40 attempt was made to compare the simulated components of AET with measurements obtained
41 through a combination of eddy covariance and scaled-up sap flow measurements. The results
42 showed that the model gives an acceptable estimate of plant transpiration and soil
43 evaporation. The associated RMSE of plant transpiration and soil evaporation were 0.59 and
44 0.73 mm per day, respectively.

45 Additionally, the irrigation efficiency was investigated by comparing the irrigation
46 scheduling design used by the farmer to those recommended by the FAO model. It was found
47 that although the amount of irrigation applied by the farmer (800 mm) during the growing
48 season of olives was twice that recommended one by the FAO model (411 mm), the

49 vegetation suffered from water stress during the summer. Such behaviour can be explained by
50 inadequate distribution of irrigation. Consequently, the FAO model can be considered as a
51 potentially useful tool for planning irrigation schedules on an operational basis.

52 **Keywords:** Crop coefficient, Evapotranspiration, Eddy covariance, FAO-56 model, *Olea*
53 *europaea*, Sap flow.

54 **1. Introduction**

55 Regions classified as semi-arid or arid constitute roughly one third of the total global
56 land cover. Within these regions, water scarcity is one of the main factors limiting agricultural
57 development. The impact of such water scarcity is amplified by inefficient irrigation practices,
58 especially since the irrigation consumes more than 85% of the available water in these regions
59 (Chehbouni et al., 2008). Therefore, the first step toward sound management of the scarce
60 water resources in these regions requires an accurate estimation of the water needs and
61 consumption of irrigated agriculture. The crop water need is defined as the amount of water
62 needed to meet the **amount of** water lost to the atmosphere through evapotranspiration.

63 During the last two decades, several models have been developed to simulate crop
64 evapotranspiration (ET) and in some cases, its components (soil evaporation and plant
65 transpiration). These models ranged from complex, physically based ones such as the Simple
66 Soil Plant Atmosphere SiSPAT (Braud et al., 1995), ISBA (Noilhan and Mahfouf, 1996), to
67 more simple and conceptual ones (Sinclair and Seligman, 1996; Olioso et al., 1999) such as
68 SVAT simple (Boulet et al., 2000). Other models such as STICS (Brisson, 1998) or CERRES
69 (Ritchie, 1986) simulate ET through the combination of a water balance with a **crop** growth
70 component. All of these models need several input parameters which cannot easily be
71 obtained at the appropriate space-time scale, **and therefore difficult for operational**
72 **applications.**

73 In addition to the above models/methods, the FAO-56 approach has been extensively
74 used to derive ET and schedule irrigation on an operational basis. This approach is often
75 preferred due to its simplicity and its robustness for operational applications. It requires fewer
76 input data, and provides acceptable ET estimates when compared to heavily parameterized
77 physically-based models (Evelt et al., 1995; Kite and Droogers, 2000; Eitzinger et al., 2002),
78 to ground measurement (e.g. Paço et al., 2006; Er-Raki et al., 2007, 2008, 2009; Liu and Luo,

79 2010) or to satellite measurements (e.g. Allen, 2000; Kite and Droogers, 2000; Duchemin et
80 al., 2006; Er-Raki et al., 2006, 2010). FAO-56 is based on the concepts of reference
81 evapotranspiration ET_0 and crop coefficients K_c , which have been introduced to separate the
82 climatic demand from the plant response (Allen et al., 1998). There are two approaches to
83 estimate crop evapotranspiration: the single and the dual crop coefficients. The FAO-56 dual
84 crop coefficient approach (Allen et al., 1998) describes the relationship between maximal
85 evapotranspiration ET_c and reference evapotranspiration ET_0 by separating the single K_c
86 into the basal crop K_{cb} and soil water evaporation K_e coefficients, while in the FAO-56
87 single crop coefficient approach, the effect of both crop transpiration and soil evaporation are
88 integrated into a single crop coefficient. Many studies have focused on the application of the
89 single approach for determining olive water requirement within Mediterranean regions (e.g.
90 Palomo et al., 2002; Abid Karray et al., 2008; Martinez-Cob and Faci, 2010). In semi-arid
91 Mediterranean region of southern Morocco, Er-Raki et al. (2008) applied also the single
92 approach over the same study site of this work, and they found that the approach
93 overestimates AET by about 18% when using the crop coefficient suggested by Allen et al.
94 (1998). Knowing that the flood irrigation is the most widely used method by the majority of
95 the farmers in Morocco (as the case of our study site), which accompanied with a large
96 amount of the water lost through direct soil evaporation (Yunusa et al., 1997), it is worthwhile
97 to estimate this substantial amount. This can be achieved by partitioning ET into soil and plant
98 components, which is the main objective of this present study. For the FAO-56 dual crop
99 coefficient approach, it has been tested for many crops (bean, corn, and sugar beet) at
100 Kimberly, Idaho (Allen et al., 1996) and tomato and cotton at Fresno, California by Itenfisu
101 (1998). As reported by Allen (2000), the comparison between estimated and measured ET by
102 precision weighing lysimeters in all situations gives an error less than 10% for daily ET
103 estimates. Allen (1999) applied the FAO-56 procedure to a 200,000 ha irrigation project in

104 California to compare the estimated ET with that determined by water balance, and the results
105 showed that despite the simplicity of the FAO model, estimates of ET were reasonable with a
106 good accuracy (overestimation of 6%). Allen (2000) also applied his methodology to an
107 extensive multiple-cropped surface, and compared the estimated ET with one obtained by
108 remote sensing. Results have shown that the FAO-56 approach overestimated ET by more
109 than 20%. Recently, several studies used the FAO-56 dual crop coefficient for estimating
110 water consumptions of different crops (Allen et al., 2005 a, b; Hunsaker et al., 2003, 2005;
111 Paço et al., 2006; Er-Raki et al., 2007). Some of these studies adopted the FAO-56 dual
112 approach to use satellite based vegetation index (Hunsaker et al., 2003, 2005; Er-Raki et al.,
113 2007; González-Dugo and Mateo, 2008; Er-Raki et al., 2010). The results show that relating
114 the basal crop coefficient K_{cb} to remotely sensed vegetation index greatly improves the
115 performance of FAO-56 method. However, Er-Raki et al. (2006) showed that the performance
116 of the FAO-56 method has some limitations when there is high soil evaporation or when
117 stress occurs. To overcome this problem and then enhance the FAO-56 performances, ET
118 derived from thermal infrared (TIR) observations was assimilated into FAO-56 single source
119 model (Er-Raki et al., 2008) in order to estimate accurately the water consumption of olive
120 orchards in the semi-arid region of the Tensift basin (central of Morocco).

121 Several studies have been specifically dedicated to estimate ET over olive trees.
122 Villalobos et al. (2000) determined olive ET by the estimation of its two components, through
123 a combination of a transpiration model based on the equation of Penmann-Monteith
124 (Monteith, 1965) and a soil evaporation model similar to that of Ritchie (1972). Palomo et al.
125 (2002) used a water balance approach, to determine water consumption in olive orchards.
126 Among the several disadvantages of this approach, summarized by Fernández and Moreno
127 (1999), is the variability of the hydraulic properties of the soil profile, such as the hydraulic
128 conductivity-soil water content relationship. Recently, Testi et al. (2004) used the eddy-

129 covariance technique to obtain direct estimates of ET of young irrigated olive orchards in
130 southern Spain. In the context of SudMed project (Chehbouni et al., 2008), Williams et al.
131 (2004) used the sap flow method combined with the isotopic method to estimate plant
132 transpiration and soil evaporation over olive orchards in southern Morocco.

133 Estimating total ET over olive orchards has been investigated by many studies in semi
134 arid regions (e.g. Fernández et al., 2001; Testi et al., 2004; Ezzahar et al., 2007; Er-Raki et al.,
135 2008), but there is little information on the partitioning of the relevant components of ET in
136 such areas; this is one of the most important ecohydrological challenges in understanding
137 water exchange and vegetation dynamics in arid and semi arid ecosystems (Reynolds et al.,
138 2000; Huxman et al., 2005). Partitioning ET is possible by using a combination of micro-
139 meteorological measurements (e.g. Bowen ratio, eddy covariance system), and eco-
140 physiological techniques (e.g. sap flow, stable isotopes) (Williams et al., 2004; Yepez et al.,
141 2005; Scott et al., 2006). However, these methods are expensive and difficult to deploy and
142 maintain in both time and space. Recently, Moran et al. (2009) developed an operational
143 approach for partitioning ET with a minimal cost and suitable for operation over long time
144 periods. This approach is based on the difference between the mid-afternoon and pre-dawn
145 soil surface temperature, which is considered as the indicator of the soil evaporation. In the
146 same context, the present study aims to use an operational model of FAO-56 dual crop
147 coefficient approach for partitioning ET. We first compared actual ET derived from this
148 approach to that measured using an eddy covariance device, then the issue of the ability of
149 this approach to provide accurate estimates of components of AET through a comparison of
150 field data obtained from a combination of eddy covariance based AET measurements and
151 spatialized or scaled-up sap flow measurements of transpiration. In this context the objectives
152 of this study were:

- 153 1. to analyze the ability of the FAO-56 dual crop coefficient model to reproduce the
154 temporal evolution of evapotranspiration and its components: plant transpiration
155 and soil evaporation. Actual plant transpiration was compared with scaled-up sap
156 flow measurements.
- 157 2. to estimate the most adequate water quantity needed for the olive and to determine
158 the best timing of irrigation by using the FAO-dual approach.

159 This paper is organized as follows. Section 2 presents a description of study site and
160 data collected during the 2003 and 2004 growing season. In Section 3, we provide a brief
161 theoretical overview of the FAO-56 dual crop coefficient approach. Section 4 presents an
162 application of this approach over an olive site. Also in this section, we analyze the ability of
163 this approach to reproduce the temporal evolution of both components of evapotranspiration
164 (plant transpiration and soil evaporation), and to derive the soil water stress coefficient K_s of
165 olives orchards during growing season in order to determine the best timing of irrigation. In
166 the final section, summary and conclusion are provided.

167 **2. Experimental Data**

168 The experimental data used for this research are similar to those used in the previous
169 paper (Er-Raki et al., 2008). Here, we presented briefly the site description, climatic and eddy
170 covariance measurements. However, as the main objective of this work is the partitioning of
171 actual evapotranspiration (AET) into plant transpiration and soil evaporation through a
172 combination of eddy covariance and scaled-up sap flow measurements, more detailed
173 presentation of sap flow measurements and its scaled-up to stand level transpiration, which is
174 not used in a previous paper (Er-Raki et al., 2008), is required.

175 **2.1. Site description**

176 This study was carried out during 2003 and 2004, in the Agdal olive (*Olea europaea* L.)
177 orchard located in the semi-arid region of the Tensift basin, south-east of Marrakech,

178 Morocco (31.601 N, 7.974 W). This area has a mean total annual precipitation of 240mm and
179 a corresponding mean annual reference evapotranspiration of 1600 mm. The average annual
180 temperature is about 22°C, rising to 38 °C in the summer (July-August) and going down to 5
181 °C in winter (December-January). Mean seasonal wind speed was about 1.2 m/s. The
182 experimental field is almost flat, planted with 240-year old olive trees, grown in an orchard of
183 about 275 ha. The density of olive trees was about 225 trees per hectare, which provides an
184 area of about 45 m² occupied by each tree. The soil type is homogeneous, with silt clay loamy
185 texture (30% clay, 25% silt, and 44% sand). The groundwater depth is approximately 40 m.
186 The soil surface was partly covered (\approx 20%) by natural grass (under story) consisting mainly
187 of short weeds during most of the year. More details about the site description are given in
188 Williams et al. (2004) and Er-Raki et al. (2008).

189 ***2.2. Data description***

190 ***2.2.1. Meteorological data and eddy covariance measurements***

191 All climatic parameters (solar radiation, air temperature, relative humidity and wind
192 speed), needed for estimate daily reference evapotranspiration (ET_0) by the FAO-Penman
193 Monteith (Equation 6 in FAO-56, Allen et al., 1998) are measured. More details about the
194 instruments (type, position) used for measurements of these parameters collected over our
195 study site are provided in Er-Raki et al. (2008).

196 Figure 1 reports the daily pattern of ET_0 calculated by the FAO-Penman-Monteith
197 equation for the two olive growing seasons (2003 and 2004). The temporal evolution of ET_0
198 during the year is typically that of a semi-arid continental climate. It is characterised by a high
199 climatic demand, with the lowest during rainy periods (winter) and the highest values
200 occurred in the sunny days (summer). Precipitation temporal patterns over the growing season
201 of olive trees were characterized by low and irregular rainfall events, with a total precipitation
202 amount of about 280 mm (Figure 1). The amount and timing of irrigations applied by the

203 farmer are presented also in this figure. It was about 800 mm for each season (2003 and 2004)
204 of olives with around 100 mm in each supply. This irrigation scheme used by the farmer has
205 been evaluated in this study.

206 **In addition to climatic measurements, an eddy covariance system, constituted with a**
207 **3D sonic anemometer (CSAT3, Campbell Scientific Ltd.) and an open-path infrared gas**
208 **analyzer (Li7500, Licor Inc.), was installed over olive tree to provide continuous**
209 **measurements of vertical fluxes of heat and water vapour at 9.2 m. A detailed description of**
210 **eddy covariance measurements can be found in Ezzahar et al. (2007) and Er-Raki et al.**
211 **(2008). As reported also in the same papers, the approximate fetch (spatial scale) of**
212 **evapotranspiration measurement is about 40m in the** northwestern direction. It might be
213 considered adequate as it contributed 90% of the measured sensible heat flux (Hoedjes et al.,
214 2007) and it includes the trees where the sap flow measurements were taken. Data sets of
215 latent heat and sensible heat fluxes have been available during 2003 and 2004 growing
216 seasons of olive orchards. Missing data in some days is associated to the collapse of power
217 supply.

218 **The evaluation of the flux measurements is undertaken through the analyzing the**
219 **energy balance closure. By ignoring the term of canopy heat storage and the radiative energy**
220 **used in photosynthesis (Testi et al., 2004; Baldocchi et al., 2000), the energy balance closure**
221 **is defined as:**

$$222 \quad R_n - G = H_{EC} + L_v E_{EC} \quad (1)$$

223 Where R_n is the net radiation; G is the soil heat flux; H_{EC} and $L_v E_{EC}$ are respectively
224 the sensible heat flux and the latent heat flux measured by eddy covariance system. Figure 2
225 shows how well the available energy ($R_n - G$) was balanced by ($H_{EC} + L_v E_{EC}$) at daily time
226 scale for the 2003 and 2004 growing season of olive orchards. The slope of the regression
227 forced through the origin was 1.06 in 2003 and 1.07 in 2004, indicating an underestimation of

228 the flux ($H_{EC} + L_v E_{EC}$) was less than 10% of the available energy ($R_n - G$), with the Root
229 Mean Square Error (RMSE) being about 17 w.m^{-2} in 2003 and 19 w.m^{-2} in 2004 (the equation
230 used to calculate *RMSE* is presented in Appendix). These results indicate --at least at the daily
231 time scale-- a good closure of the energy balance, which is in agreement with other studies
232 (Testi et al., 2004; Baldocchi et al., 2000; Twine et al., 2000).

233 **2.2.2. Sap flow measurements**

234 Heat-pulse sap flow sensors (Heat Ratio Method, HRM, Burgess et al., 2001) were used to
235 measure xylem sap flux on eight olive trees. The HRM method has been described in detail in
236 Burgess et al. (1998) and Burgess et al. (2001). Briefly, the HRM is a modification of the
237 Heat Pulse Method (HPM) and it employs temperature probes inserted into the active xylem
238 at equal distances down- and upstream from a heat source. This method improves on the HPM
239 by its precision at very slow flow rates and even reverses sap flow can be measured.
240 Reliability of this technique for determining transpiration has been demonstrated by several
241 studies (Burgess et al., 2001; Fernández et al., 2001 and Williams et al., 2004). The heat pulse
242 sensors were installed on four large single-stemmed and on four large multi-stemmed trees
243 adjacent to the eddy covariance tower during the summer of 2003 and 2004. Sap flow
244 measurements were conducted from the 14th of June (DOY 165) through the 30th of July
245 (DOY 211) during 2003 and from the 9th of May (DOY 130) through the 28th of September
246 (DOY 272) during 2004. Missing data in some days (from DOY 183 to DOY 194 and from
247 DOY 222 to DOY 224) is due to problems with the power supply. The same measurements of
248 sap flow, with the same sensors and over the same field have been made by Williams et al.
249 (2004) except in other climatic conditions (during the winter). A detailed description of HRM
250 technique and the principle of measurements can be found in Williams et al. (2004).
251 Volumetric sap flow (L day^{-1}) was scaled to tree transpiration (mm day^{-1}) using a survey of
252 the average ground area of each tree (45 m^2). After the scaling the measured sap flow to the

253 single tree transpiration, we extrapolated this latter to the stand level transpiration, which is
254 representative for the whole field. The allometric method is the most one used for this up
255 scaling (e.g. Kumagai et al., 2005; Ford et al., 2007). However, this method is destructive and
256 very time-consuming. Nevertheless, in this study the extrapolation of the tree transpiration to
257 the stand level transpiration has been performed based on the measurements of eddy
258 covariance. This strategy was also proposed by Williams et al. (2004) and Oishi et al. (2008)
259 in the context of scaling ecosystem-level transpiration from sap flux measurements based on
260 the measured AET by eddy covariance system. In the same context, we tried to find the
261 relationship between the scaled measured sap flow and measured AET by eddy covariance,
262 equivalent to tree level transpiration, during the dry conditions (when the soil evaporation is
263 negligible). In order to select the dry period, daily evolution of $\Delta\theta$ was plotted from DOY 195
264 to DOY 219 (Figure 3), when $\Delta\theta$ means the difference between the soil moisture at 5 cm
265 depth on day i and on day i-1. One assumes that $\Delta\theta$ is proportional to the soil evaporation flux
266 at least several days after a major irrigation, i.e. when the excess water has been redistributed
267 within the soil moisture profile. By analysing this figure, one can see that $\Delta\theta$ increased in
268 absolute value from DOY 195 to DOY 202, and after it decreased until DOY 212. After this
269 day, $\Delta\theta$ is almost constant and close to zero. Therefore, the soil moisture ($0.13 \text{ m}^3/\text{m}^3$)
270 correspond to DOY 212 was considered as the threshold which can be the indicator of the
271 presence of soil evaporation. When the soil moisture at 5 cm is lower than this threshold, the
272 soil evaporation is considered negligible regarding the large values of ET_0 during the
273 summer. After the selection of the dry conditions, measured daily AET by eddy covariance
274 system was plotted against the daily scaled sap flow measurements (data not presented here).
275 Then, the obtained linear regression between daily stand level transpiration and daily scaled
276 sap flow is:

$$277 \text{ Stand level transpiration} = 1.25 * (\text{scaled sap flow}), \quad R^2 = 0.74 \quad (2)$$

278 This model was applied also for the remaining days (wetting days) of sap flow
279 measurements for calculating the stand level transpiration. This may create some errors in
280 estimating stand level transpiration as reported by Oishi et al. (2008) when they found that the
281 agreement between the estimates of components AET is greatly affected by the scaling
282 procedure.

283 3. Theoretical overview of the FAO-56 dual approach

284 Detailed descriptions of the FAO-56 dual crop coefficient approach are available from
285 Allen et al. (1998). In this section, we will briefly present this approach. The actual crop
286 evapotranspiration (AET) estimated by this approach is given by the following equation:

$$287 \quad AET = (K_s K_{cb} + K_e) ET_0 \quad (3)$$

288 Where K_{cb} , K_e and K_s are basal crop coefficient, soil evaporation and water stress
289 coefficient, respectively. The required equations for deriving these three parameters are
290 presented henceforth.

291 3.1. Calculation of K_{cb}

292 The methodology adopted here to compute basal crop coefficient (K_{cb}) follows
293 strictly the same method displayed in Er-Raki et al. (2008) for calculating crop coefficient
294 (K_c). As mentioned previously in the introduction, two different crops (olives and under
295 story) grew up together in the same field. A simple formula was used to calculate the
296 equivalent basal crop coefficient (Allen et al., 1998):

$$297 \quad K_{cbfield} = f_{c-olives} K_{cbngc} + (1 - f_{c-olives}) K_{cbcov er} \quad (4)$$

298 Where K_{cbngc} and $K_{cbcov er}$ are basal crop coefficients for olive and under story,
299 respectively. The average seasonal value of basal crop coefficient of olive orchards (K_{cbngc})
300 was derived based on sap flow measurements. For the basal crop coefficient of under story

301 ($K_{cb\ cover}$), it was assumed equal to the difference between the ratio of measured AET and ET_0

302 ($K_c = \frac{AET}{ET_0}$) and K_{cbngc} during the dry conditions. $f_{c-olives}$ is the fraction of soil surface

303 covered by olive trees, calculated as $f_{c-olives} = \frac{\pi D^2 N}{40000}$ where D (m) is the average diameter of

304 the canopy and N is the number of trees per hectare. It was found to be equal to about 0.60.

305 **3.2. Calculation of K_e**

306 This coefficient is calculated based on daily computation of the water balance for the

307 surface soil evaporation layer Z_e (equation 71 in FAO-56 paper). The calculation procedure

308 requires input of soil parameters such as the soil moisture at field capacity (θ_{fc}) and at the

309 wilting point (θ_{wp}), the total evaporable water (TEW), the readily evaporable water (REW)

310 and the depth of Z_e . An average value of $0.32\ m^3\ m^{-3}$ for θ_{fc} and $0.19\ m^3\ m^{-3}$ for θ_{wp} were

311 obtained (Er-Raki et al., 2008). For the depth of the soil surface evaporation layer Z_e (m),

312 Allen et al. (1998) suggested values ranging between 0.10 and 0.15 m. Because the soil is

313 covered with herbs and is shaded by trees, the value of $Z_e = 0.15\ m$ is adopted in this study. A

314 typical REW value for a silt clay loamy soil of 9 mm (FAO-56, table19) was used in the

315 calculations. Another parameter is needed for the calculation of K_e . This parameter is named

316 the exposed and wetted soil fraction (f_{ew}). It was derived following the suggestions of Allen

317 et al. (1998) (equation 75 in FAO-56 paper). In fact, following irrigation (flooding technique)

318 or rainfall, the soil was completely wetted except the area covered by the stem ground tree.

319 This area represents about 5% of the whole area occupied by each tree. Then, the fraction of

320 soil surface wetted by irrigation or precipitation f_w was about 0.95 and f_{ew} was equal to $(1-$

321 $f_c)$. In the absence of the rainfall and irrigation, it was equal to $f_w \cdot f_c$. f_c represents the fraction

322 of soil surface covered by olive trees and understory, which it was set at 0.80.

323 3.3. Derivation of K_s

324 When the dual crop coefficient approach is adapted, the effects of soil water stress on
325 crop AET are accounted by multiplying the basal crop coefficient by the water stress
326 coefficient, K_s . Mean water content of the root zone is expressed by the root zone depletion,
327 D_r . At field capacity, the root zone depletion is zero ($D_r = 0$). Water stress occurs when D_r
328 becomes greater than RAW , the depth of readily available water in the root zone. For $D_r >$
329 RAW , K_s is given by:

$$330 K_s = \frac{TAW - D_r}{TAW - RAW} = \frac{TAW - D_r}{(1-p)TAW} \quad (5)$$

331 Where K_s is a dimensionless transpiration reduction factor dependent on available
332 soil water [0–1], D_r is root zone depletion [mm], TAW is total available soil water in the root
333 zone [mm], and p is the fraction of TAW that a crop can extract from the root zone without
334 suffering water stress. When $D_r \leq RAW$; $K_s = 1$.

335 TAW is estimated as the difference between the water content at field capacity and
336 wilting point:

$$337 TAW = 1000 \left(\theta_{fc} - \theta_{wp} \right) Z_r \quad (6)$$

338 Where Z_r is the effective rooting depth [m].

339 Readily available soil water of the root zone is estimated as:

$$340 RAW = pTAW \quad (7)$$

341 The required soil parameters for calculation of K_s are taken from Er-Raki et al.
342 (2008). TAW and RAW values calculated from equation (6) and (7) are 208 and 135.2 mm,
343 respectively. The rooting depth Z_r of olive trees and the depletion fraction p were set at 1.60
344 m and 0.65 respectively, according to FAO-56 values (table 22).

345 **4. Results and discussions**

346 ***4.1. Estimating AET by the FAO-56 dual crop coefficient approach***

347 **We calculated the variation over time of** actual evapotranspiration (AET) using the
348 FAO-56 dual crop coefficient approach over olive trees during two consecutive growing
349 seasons (2003 and 2004). The simulation was performed from March, 1st (DOY 60) to
350 November, 25th (DOY 329) for 2003, and from the March, 1st (DOY 61) to November, 7th
351 (DOY 312) for 2004. The average seasonal value of basal crop coefficient of olive orchards
352 used for the simulation was about 0.54. This value was derived based on sap flow
353 measurements during 2003. It was expressed as the ratio of measured transpiration by sap
354 flow to the reference evapotranspiration ET_0 . When sap flow measurements are not available,
355 basal crop coefficient has been derived based on minimum values of measured crop
356 coefficient (ratio of measured AET to ET_0) according to Teixeira et al. (2008). The
357 comparison between measured and predicted AET (figure 4) shows good agreement between
358 observed and simulated AET values. The Root Mean Square Error (RMSE) between
359 measured and simulated AET values during 2003 and 2004 were respectively about 0.54 and
360 0.71 mm per day. **Some discrepancies between measured and simulated AET can be clearly**
361 **observed around wetting events (irrigation and rainfall) and stress period. For wetting events,**
362 **the difference between measured and simulated AET values may be due to the deep**
363 **percolation and the rainfall interception, which not taken into account by the model. As**
364 **reported by Gomez et al. (2001), rainfall interception plays an important component of the**
365 **water balance, and they showed that the water intercepted by the rainfed olive orchards was**
366 **about 8% for a heavy rainfall. Another factor that may partly explain some of the difference**
367 **between measured and simulated AET values is the flux source area measured by eddy**
368 **covariance which depends to the wind direction. In fact, the eddy covariance system measures**
369 **the evapotranspiration over a relatively large area (wet and dry) whereas the model simulates**

370 it locally (wet or dry). Moreover, the dual approach tends to overestimate the crop
371 evapotranspiration at the peak values (wetting events). This is corroborated by Liu and Luo
372 (2010) when they found that the dual approach of FAO-56 is appropriate for simulating the
373 total quantity of evapotranspiration but inaccuracy in simulating the peak value after
374 precipitation or irrigation (Peng et al., 2007). This necessitates a more profound study for
375 improving some parameterisation used in the FAO-56 method. For the periods of hydric stress
376 (ex. from DOY 212 to DOY 238 during 2004), the approach leads to higher values of AET in
377 comparison with that measured by the eddy covariance system. This may be due to an
378 overestimation of soil water stress coefficient. The overestimation of soil water stress can be
379 related to the overestimation of the rooting depth (1.6 m). This misrepresentation of the
380 rooting depth influences directly the ability of the plant to extract water. A similar result has
381 been found when using the single approach for estimating water consumption for the same
382 olive orchard (Er-Raki et al., 2008). To illustrate more clearly the effect of the rooting depth
383 (Z_r) on evapotranspiration, it is essential to perform a sensitivity analysis of the FAO model to
384 this parameter. The impact of rooting depth on AET (data not shown here) showed that the
385 values of simulated AET increase with increasing of Z_r . In fact, an increase in Z_r causes an
386 increase of Total Available Water (TAW) within the root zone and leads to an increase in the
387 soil water stress coefficient K_s which is straightforward according to the equations (6) and
388 (7). An increase in Z_r of 33.33%, 66.66%, and 166% (from 1.2 to 1.6, from 1.2 to 2 from 1.2
389 to 3.2 m, respectively) leads to an increase of AET by about 34%, 47% and 49%, respectively.
390 Note that the effect of rooting depth on AET is negligible when the water is not limiting
391 within the root zone.

392 According to this study, the FAO-56 dual crop coefficient approach simulates
393 reasonably well the total evapotranspiration. The question addressed after is how efficiently

394 this approach simulates the two components individually: plant transpiration and soil
395 evaporation.

396

397 ***4.2. Performance of the dual crop coefficient approach for Partitioning of*** 398 ***evaporation and transpiration***

399 The FAO-56 dual crop coefficient approach computes separately soil evaporation and
400 plant transpiration; it is of interest to investigate how well these individual components are
401 simulated. To achieve this objective, we combined eddy covariance based measurement of
402 AET with scaled-up sap flow measurements to estimate soil evaporation and plant
403 transpiration against which the simulated components will be compared. As stated above, sap
404 flow measurements were conducted from DOY 130 to DOY 272 during 2004. The sum of
405 measured soil evaporation and the under story transpiration was computed as the difference
406 between AET measured by the eddy covariance and the olive transpiration measured by the
407 sap flow method. For practical reasons, the heat pulse sensors were inserted only into the
408 active xylem of olive trees. Therefore, we compared the simulated soil evaporation with the
409 sum of measured soil evaporation and the under story transpiration. Also, because the FAO-
410 dual approach predicts separately only soil evaporation and plant transpiration (olives + under
411 story) and does not discriminate between the three components (soil evaporation, olive
412 transpiration and under story transpiration), we compared the simulated plant transpiration
413 with the measured olive transpiration only.

414 Figure 5a presents a comparison between the measured and simulated transpiration for
415 the 2004 growing season. Daily patterns are very similar and respond to that of ET_0 (see
416 Figure1-bottom). The RMSE between measured and simulated olive transpiration was about
417 0.59 mm per day. The cumulated values of the entire experimental period (128 days) are 362
418 mm for the transpiration measured by the HRM method and 387 mm for the model, leading to

419 a difference of 7%. This difference is consistent with the fact that the model simulates the
420 olive transpiration and the under story transpiration while the measurement estimates only the
421 olive transpiration. Another factor that may partly explain this difference is that the scaling
422 approach (Eq. 2) is not error-free (Fernández et al., 2001; Williams et al., 2004).

423 Regarding soil evaporation, Figure 5b shows that in dry conditions (absence of
424 irrigation and rainfall) both simulated and measured soil evaporation are almost zero. It
425 should be noted that the observed negative values of measured soil evaporation are considered
426 as an artefact and set to zero. After irrigation or rainfall, as expected, both measured and
427 estimated soil evaporation increased. However the increasing magnitude is different. This
428 error is likely due to the fact that:

429 1) the model simulate only soil evaporation, but the measurement gives the soil
430 evaporation and under story transpiration;

431 2) the scaling approach of measured transpiration from the eddy covariance (Eq. 2)
432 may not be valid in humid conditions and tends to underestimate the measured plant
433 transpiration and thus overestimate soil evaporation, also this error can be related also to the
434 under story contribution;

435 3) the impact of the difference in the footprint of the eddy covariance and the area
436 where the measured sap flow was taken during the irrigation events. This may be
437 underestimates or overestimates measured evaporation.

438 Despite those discrepancies between the measured and simulated soil evaporation, the
439 model gives acceptable results in estimating soil evaporation. The RMSE between simulated
440 and measured soil evaporation (+ under story transpiration) was 0.73 mm per day.

441 The ratio of plant transpiration and soil evaporation to total evapotranspiration (data
442 not shown here) showed that before irrigation, plant transpiration represents 100% of total
443 evapotranspiration and decrease to about 65% after irrigation. An amount of 35% of water

444 was lost by direct soil evaporation due to flooding irrigation and the olive trees are widely
445 spaced. Another study was done over the same field by Williams et al. (2003, 2004) in winter.
446 It showed that after the irrigation, the soil evaporation represents about 14–28% of the total
447 evapotranspiration. Yunusa et al. (1997) also studied the partitioning of seasonal
448 evapotranspiration from a commercial furrow-irrigated Sultana vineyard, and they found that
449 substantial amounts of the soil water (about 49% of total ET) were lost through soil
450 evaporation. In order to prevent the water losses by evaporation from the ground, it is very
451 important to wet the maximum volume of root zone and the minimum soil surface. Therefore,
452 a localized irrigation system is usually the most appropriate.

453 Despite the simplicity of the water balance model used in the FAO-56 formulation, the
454 obtained results showed that the FAO-56 dual crop coefficient can simulate correctly crop
455 evapotranspiration and gives also an encouraging result for partitioning of evapotranspiration
456 into soil evaporation and plant transpiration. Because this approach is very simple and
457 designed to schedule irrigation on an operational basis, it is of interest to check the irrigation
458 planning practiced by the farmer over the study site. This can be achieved by applying the
459 FAO model in order to determine when to irrigate and how much water to apply.

460

461 ***4.3. Assessment of irrigation planning***

462 In order to assess the efficiency of the irrigation planning over the study site, we
463 calculated the soil water stress coefficient K_s by the FAO-56 dual approach (Figure 6a). The
464 soil water stress coefficient, K_s , for olive orchards ranges from 0 to 1 according to Equation
465 (5), and it shows how the soil water depletion, D_r , changes to limit crop evapotranspiration.
466 The value of K_s depends on the soil water depletion linked to water supply (rainfall or
467 irrigation). The soil water stress is equal to 1 when the soil water depletion is less than the
468 readily available water of the root zone (*RAW*). The absence of irrigation and rainfall (from

469 DOY 191 to DOY 237) results in an increase in the root zone depletion that exceeds *RAW* and
470 generates stress (K_s below to 1). The increase in soil water depletion is due to the removal of
471 water by evapotranspiration and percolation losses that induces water stress conditions. The
472 information obtained from the soil water stress coefficient can be used in an irrigation
473 scheduling program for deciding when and how much to irrigate. For this purpose, we used
474 the FAO-56 dual approach to simulate the amount and the frequency of irrigation needed
475 (figure 6a) for olive orchards in 2004 to avoid water stress (i.e. so that $K_s=1$ at all times).
476 Following the FAO-56 procedure, irrigation is required when rainfall is insufficient to
477 compensate the water lost by evapotranspiration. By calculating the soil water balance of the
478 root zone on a daily basis (Equation 85 in FAO-56), the timing and the depth of the irrigations
479 can be planned (Figure 6a). According to this figure, the average amount of irrigation in each
480 supply was about 136 mm which corresponds to the value of *RAW*. The total irrigation
481 recommended by the model was about 411 mm which it is half that given by the farmer (800
482 mm). It can be noticed also in this figure, that although the amount of irrigation given by the
483 farmer was greater than the one simulated by the FAO model, the vegetation suffered from
484 water stress during the summer (between DOY 228 and DOY 237). Such behaviour can be
485 explained by inadequate distribution of irrigation: the wrong quantity is applied at the wrong
486 moment. In fact, the farmer didn't apply the irrigation in this period while the model
487 recommends the irrigation in the same period. It can be seen also that the farmer applied a
488 large amount relatively to the required quantity given by the model. This is due to some
489 unnecessary irrigation events during the rainfall period (e.g. DOY 169). The results revealed
490 that the method of irrigation applied by the farmer was not appropriate for the olive orchards
491 in the Tensift plain.

492 As we found above that an important amount of water ($\approx 35\%$) was lost by direct soil
493 evaporation during the wetting event, it is of interest to quantify the amount of water needed

494 to fulfill this water depleted from the topsoil layer (Z_e). A similar parameter to K_s named soil
495 evaporation reduction coefficient (K_r) was deployed to schedule irrigation in the topsoil layer
496 (Allen et al., 1998). Similarly to K_s , the estimation of K_r requires a daily water balance
497 computation but for the surface soil evaporation layer Z_e . Figure 6b presents the evolution of
498 K_r with the timing and the amount of irrigation required. According to this figure, K_r varied
499 between 0 and 1 depending to the amount of water available in the topsoil. Following heavy
500 rainfall or irrigation, the evolution of K_r or soil evaporation rates can be described as two-
501 stage process: an energy limiting stage, and a falling rate stage. In the first stage, the soil
502 surface remains wet, the amount of water depleted by evaporation is equal to 0 and K_r is equal
503 to 1. When the water content was reduced due to the depletion of water by evaporation
504 (second stage) with the absence of rainfall and irrigation, K_r decreases and reaches 0 when the
505 total evaporable water (TEW) was totally depleted. In this second stage, the soil evaporation
506 rate decreases depending to the amount of water remaining in the surface soil layer and the
507 soil hydraulic properties that determine the transfer of liquid and vaporized water to the
508 surface. Ritchie (1972) funded that in the second stage the evaporation rate decreases as a
509 function of the square root of time after wetting event. Based on the calculation of K_r by using
510 the water balance equation (equations 74 and 77 in FAO 56), the timing and the amount of
511 irrigations needed in the topsoil layer can be planned (Figure 6b) in order to compensate the
512 water depleted from the topsoil layer (Z_e). According to this figure, the irrigation is required
513 by FAO when the soil water depletion is higher than the readily evaporable water (REW) and
514 no irrigation otherwise. The average value of irrigation recommended by FAO model in each
515 supply was about 9 mm which corresponds to the value of REW except in the beginning of the
516 calculation where it would be assumed as the total evaporable water (TEW=33.55 mm). The
517 total amount of irrigation recommended by the FAO model for maintaining soil wet was about
518 186 mm, which presents about 45% of the irrigation needed (411 mm) for avoiding olive

519 water stress. The cumulated values of total irrigation in the topsoil and in the root zone are
520 about 597 mm for the simulated irrigation requirement by the FAO model and 800 mm for
521 that given by the farmer, with a difference of 25%. This difference may be related to the
522 inadequate amount and planning of irrigation by the farmer. Water amounts and timing are
523 planned only by the understanding and perception of the farmer without using any guideline
524 for scheduling the amount and timing of irrigation water applications. **Consequently, some**
525 **irrigations are missing or unnecessary. For example on DOY 201 (July, 19), the model**
526 **recommends the irrigation while the farmer didn't apply it in this day. Effectively, in this**
527 **period, the irrigation is needed because the climatic demand was very higher in the summer.**
528 **During the rainfall period (e.g. DOY 169), the irrigation is not necessary, but the farmer apply**
529 **one. The results revealed that the method of irrigation applied by the farmer was not**
530 **appropriate for the olive orchards in the Tensift plain.** It would be advisable to improve the
531 irrigation management and to recommend the farmer to use the simple FAO model, which can
532 be considered as a potentially useful tool for planning irrigation schedules on an operational
533 basis.

534 **5. Conclusions**

535 The main objective of this study was to investigate the potential of the FAO-56 dual
536 crop coefficient approach to provide accurate estimates of actual evapotranspiration (AET)
537 and its components of the olive orchard in semi-arid region. Model simulations of evaporation
538 and transpiration were compared to data obtained from a combination of eddy covariance
539 based AET measurements and scaled-up sap flow measurements of transpiration.

540 The results showed that, by using the local values of basal crop coefficients derived
541 from sap flow measurements, the approach simulates reasonably well AET over two growing
542 seasons. The Root Mean Square Error (RMSE) between measured and simulated AET values
543 during 2003 and 2004 were respectively about 0.54 and 0.71 mm per day. The value of basal

544 crop coefficient for the olive orchard used in this study was about 0.54. This value was lower
545 than that suggested by the FAO-56 (0.62).

546 Since the FAO-56 dual crop coefficient approach predicts separately soil evaporation
547 and plant transpiration, an attempt of comparison of the simulated components of
548 evapotranspiration (soil evaporation and plant transpiration) with the measurements showed
549 that the model gives an acceptable estimate of plant transpiration and soil evaporation. The
550 RMSE between measured and simulated transpiration and soil evaporation (resp) were 0.59
551 and 0.73 mm per day. In conclusion, it can be stated that the FAO-56 dual crop coefficient
552 gives an encouraging result for partitioning of evaporation and transpiration despite the
553 simplicity of the formulation used to derive such partition. Further effort will be necessary if
554 more accurate simulation of ET partitioning is needed by this approach. This can be achieved
555 by subdividing the plant transpiration into olive transpiration and understory transpiration and
556 so two water balance are necessary for calculation stress coefficient K_s .

557 Additionally, the results of this study revealed that the irrigation scheme used by the
558 farmer was not appropriate for the olive orchards in the Tensift plain. It was found that
559 although the amount of irrigation applied by the farmer (800 mm) during the growing season
560 of olive orchards was greater than the simulated one by the FAO model (411 mm), the
561 vegetation suffered from water stress especially during the summer. Such behaviour can be
562 explained by inadequate distribution of irrigation. It would be advisable to improve the
563 irrigation management and to recommend the farmer to use the simple and operational FAO
564 model for planning irrigation schedules.

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571 **7. Appendix**

572 Three statistics were used for analyzing the data: 1) the Mean Bias Error (MBE),
573 which indicates the average deviation of the predicted values from the measured values; 2) the
574 Root Mean Square Error (RMSE), which measures the discrepancy of predicted values around
575 observed values; 3) the efficiency (E), which judges the performances of simulation data.

$$576 \quad MBE = \bar{y}_{mod} - \bar{y}_{obs}$$

$$577 \quad RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_{i\ mod} - y_{i\ obs})^2}$$

$$578 \quad E = 1 - \frac{\sum_{i=1}^n (y_{i\ mod} - y_{i\ obs})^2}{\sum_{i=1}^n (y_{i\ obs} - \bar{y}_{obs})^2}$$

579 Where \bar{y}_{mod} and \bar{y}_{obs} are the averages of simulations and observations, n is the number of
580 available observations, $y_{i\ mod}$ and $y_{i\ obs}$ are daily values of modeled and observed variables
581 respectively.

582

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761 **Figure captions**

762 **Figure 1.** Daily reference evapotranspiration ET_0 calculated following the FAO-Penman-
763 Monteith equation during 2003 (top) and 2004 (bottom) growing seasons. Rainfall and
764 irrigation events are shown in the same figures.

765 **Figure 2.** Assessment of energy balance closure. Daily average fluxes of net radiation (R_n)
766 minus the soil heat flux (G) are compared against the sums of sensible (H_{EC}) and latent heat
767 ($L_v E_{EC}$) measured by the eddy covariance system.

768 **Figure 3.** Daily evolution of the difference between the soil moisture at 5 cm depth on day i
769 and on day $i-1$ ($\Delta\theta$). The vertical dotted line shows the date on which the soil evaporation was
770 negligible.

771 **Figure 4.** Time course of observed (triangles on dotted line) and simulated (solid line) actual
772 evapotranspiration using the **FAO-56 dual crop coefficient approach** for 2003 (top) and 2004
773 (bottom) growing seasons of olives orchard in Tensift Alhaouz, Marrakech Morocco.

774 **Figure 5.** Time course of observed (triangles on dotted line) and simulated (solid line) two
775 components of evapotranspiration using the FAO-56 dual crop coefficient approach of olives
776 orchard in Tensift Alhaouz, Marrakech Morocco during the 2004 growing season (from
777 DOY130 to DOY272): a) plant transpiration, b) soil evaporation. The measured soil
778 evaporation is computed as the difference between measured evapotranspiration by eddy
779 covariance system and measured transpiration by sap flow. Note that the negative values of
780 measured soil evaporation obtained when the measured transpiration is higher than measured
781 (AET) by eddy covariance system, were taken equal zero.

782 **Figure 6.** Estimated daily soil water stress coefficient K_s (figure a) and daily soil evaporation
783 reduction coefficient K_r (figure b) by the FAO-56 dual crop coefficient approach for the olives
784 orchard in Tensift Alhouz, Marrakech Morocco during 2004 growing season. Amount and

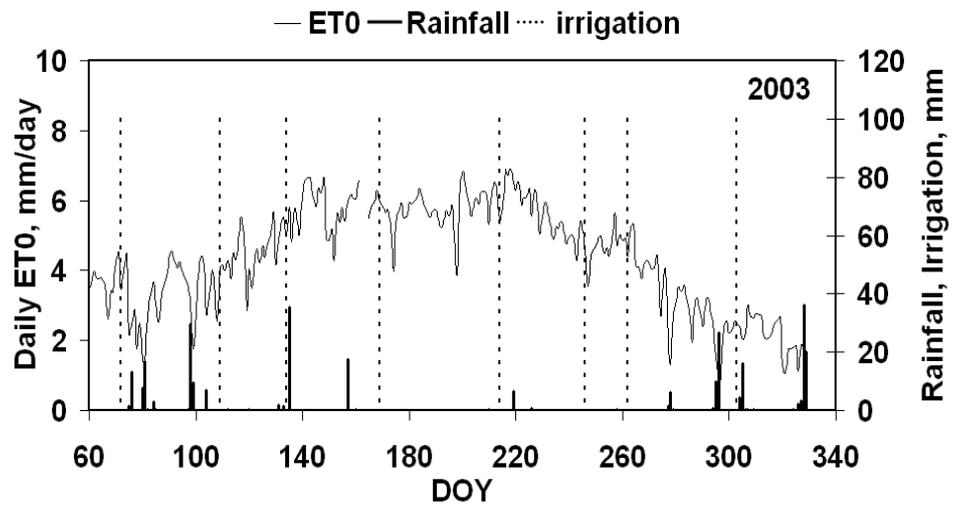
785 frequency of irrigation given by the farmer and recommended by the FAO model are shown
786 in the same figures.

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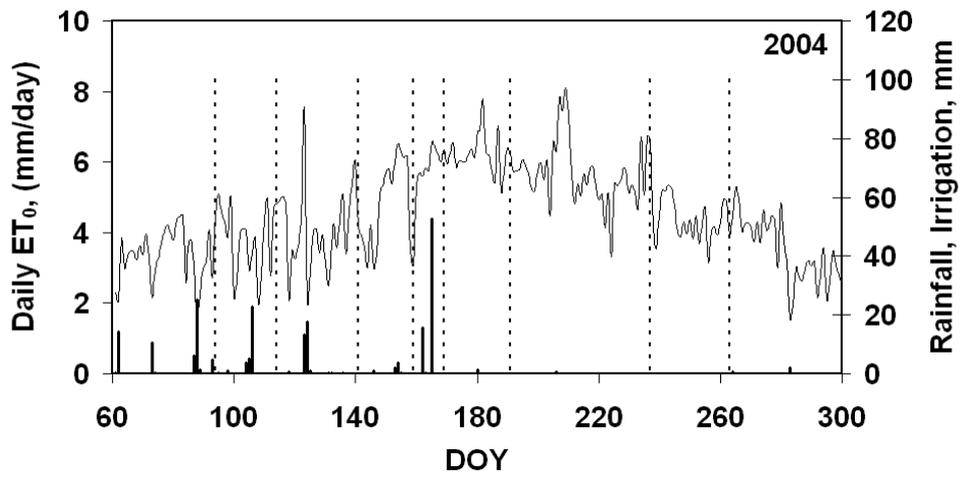
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Figure 1



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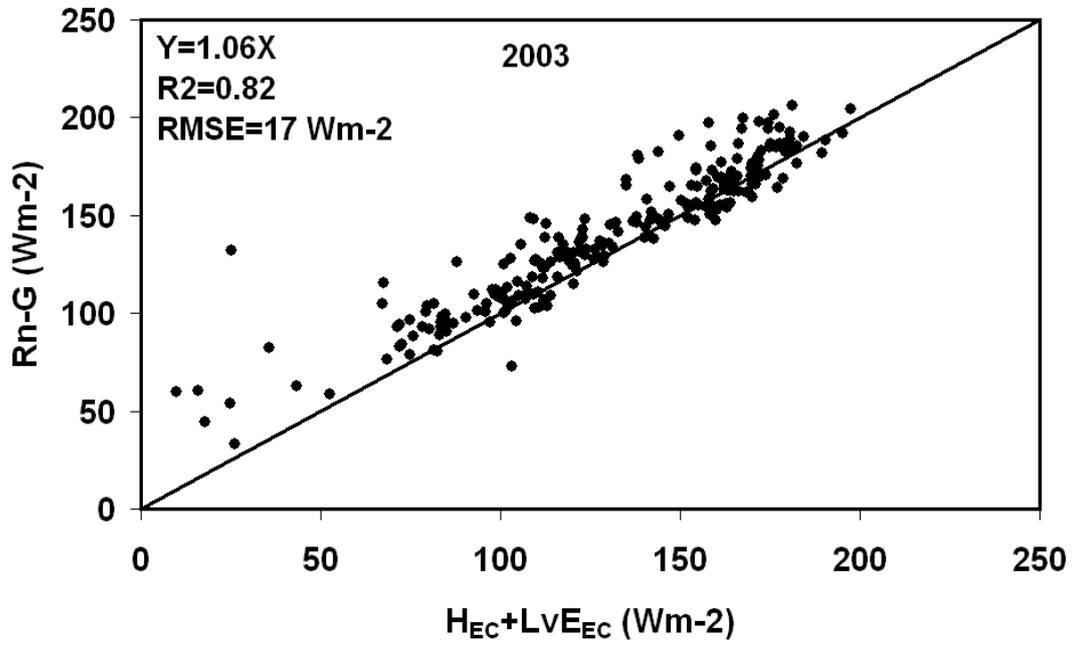
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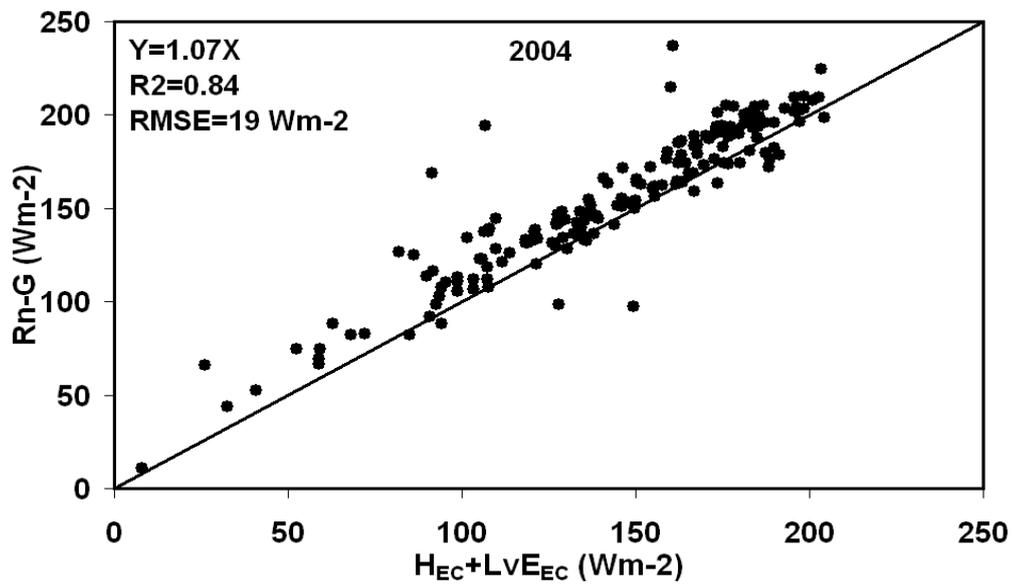
Figure 2

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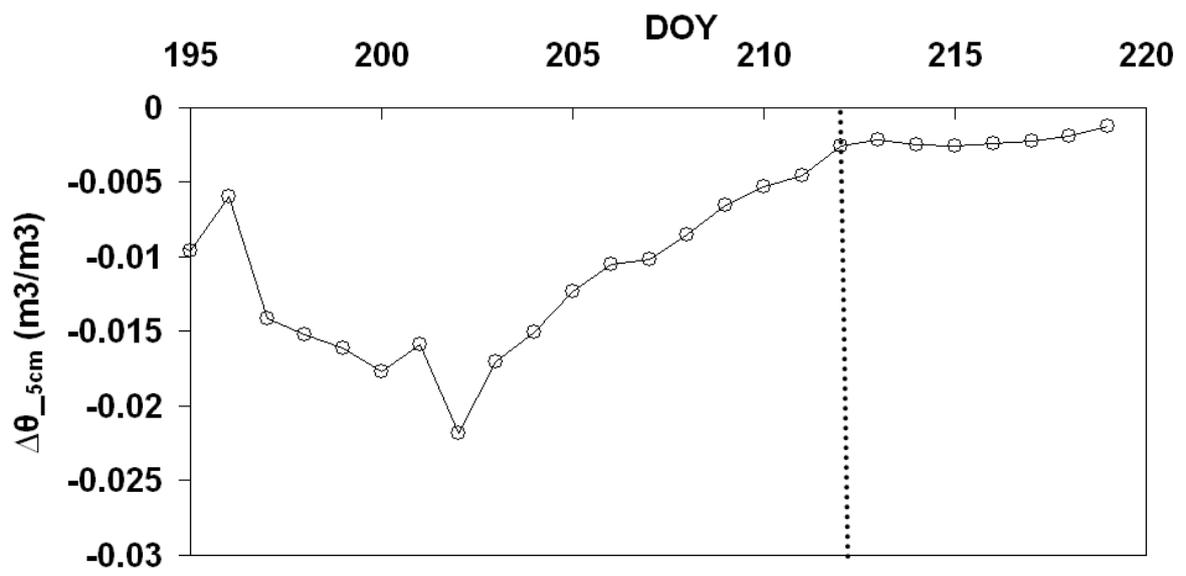
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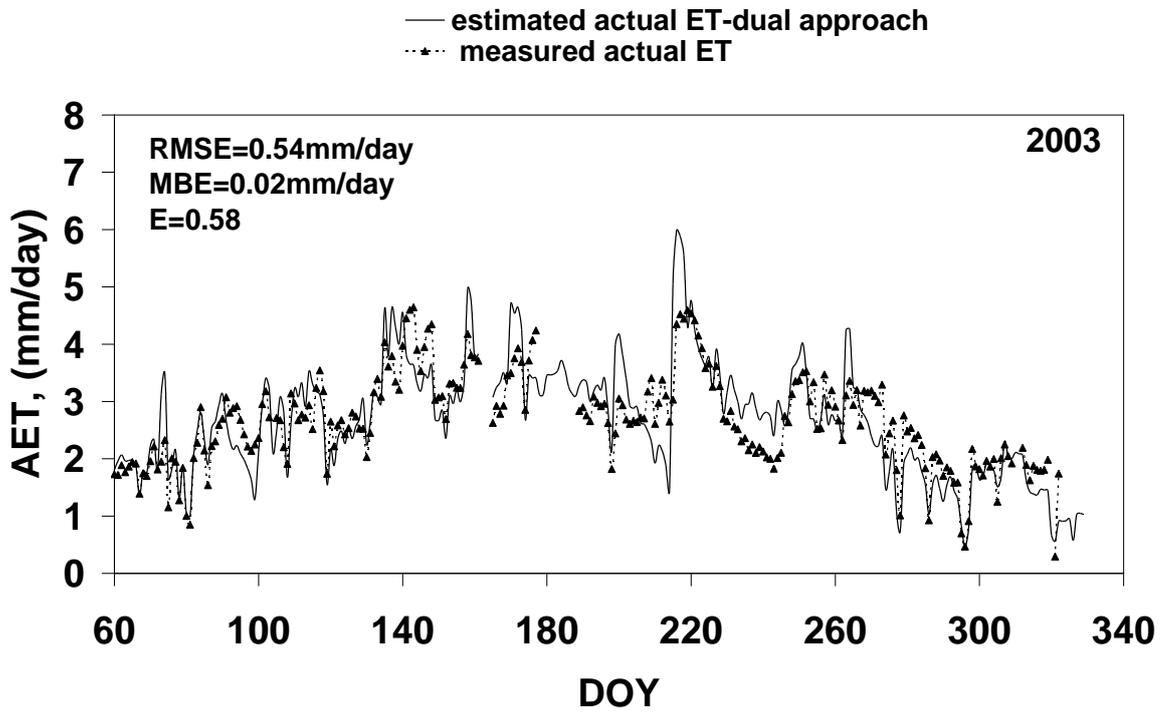
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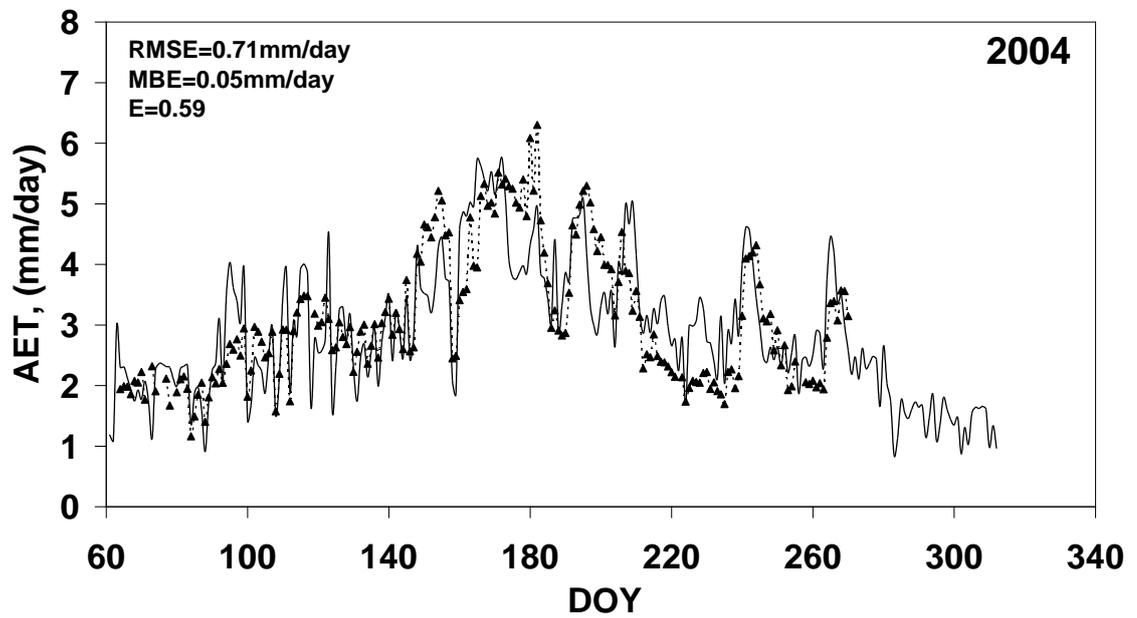
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Figure 4



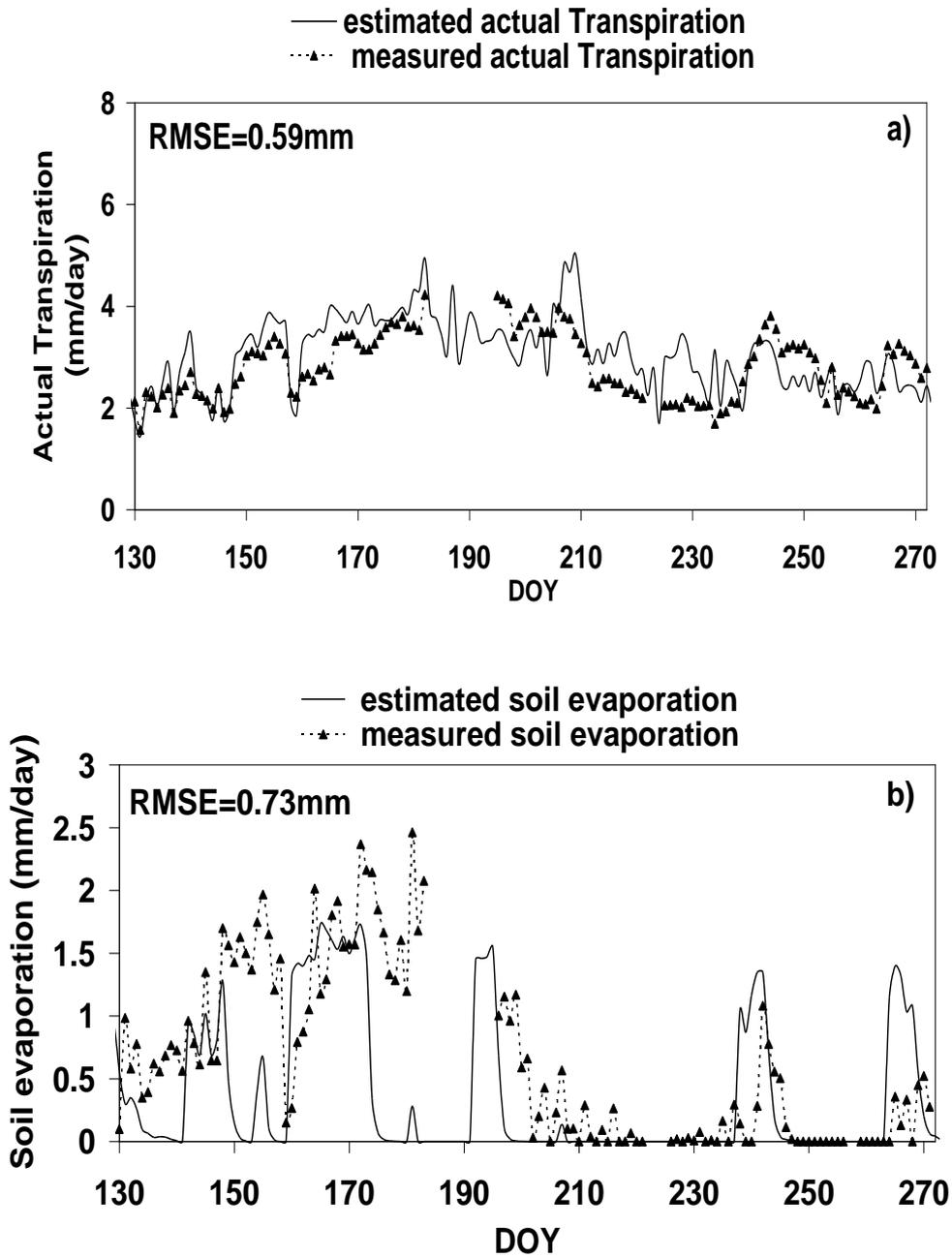
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Figure 5



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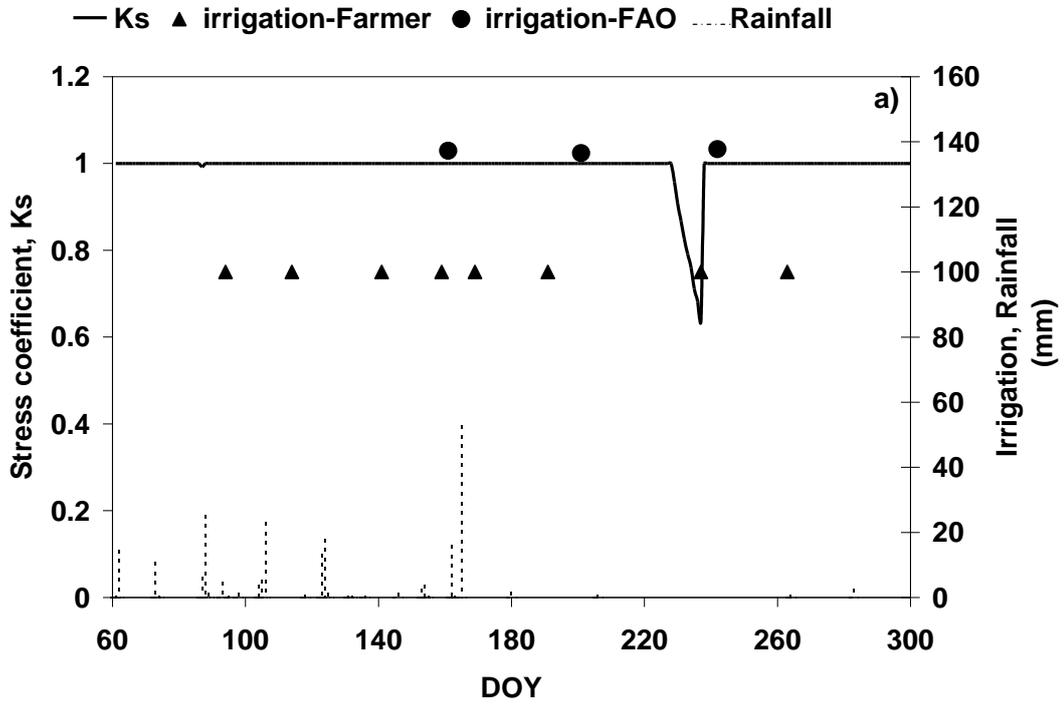
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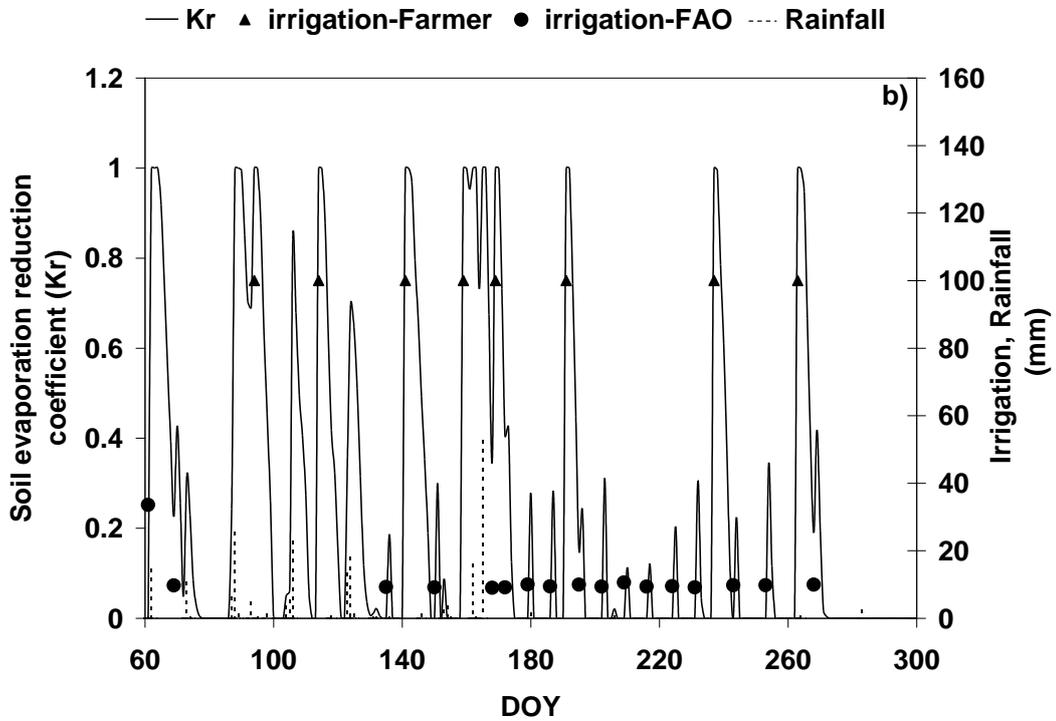
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Figure 6



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