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Campo García, ADD.; González Sanchis, MDC.; Molina Herrera, A.; Garcia-Prats, A.; Ceacero, CJ.; Bautista, I. (2019). Effectiveness of water-oriented thinning in two semiarid forests: The redistribution of increased net rainfall into soil water, drainage and runoff. *Forest Ecology and Management*. 438:163-175. <https://doi.org/10.1016/j.foreco.2019.02.020>



The final publication is available at

<http://doi.org/10.1016/j.foreco.2019.02.020>

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Additional Information

1 **Effectiveness of water-oriented thinning in two semiarid forests: the redistribution**
2 **of increased net rainfall into soil water, drainage and runoff**

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15 **Key words:** adaptive silviculture, forest hydrology, Aleppo pine *Pinus halepensis*,

16 Holm oak *Quercus ilex*, boosted regression trees, elasticity analysis.

17

18 **Abstract**

19 Water is the key element that modulates the provision of goods and services together with
20 global/climate stressors affecting semiarid forests. In this sense, there is a need to improve
21 the understanding and quantification of forest and water relationships as affected by forest
22 management. This work addresses this issue by comparing net rainfall (Pn) redistribution
23 into different belowground hydrological processes (BHP) in two forest types after a
24 thinning treatment: a holm oak coppice (HU) and a post-fire Aleppo pine regeneration
25 (CAL). The relative contribution (RI) of forest structure, antecedent soil moisture (θ_{st}),
26 rainfall and meteorological conditions on the BHP was assessed through boosted
27 regression trees models. In both sites, the RI of the forest structure itself was limited
28 (<10%). However, θ_{st} , which clearly increased significantly with thinning, received an
29 average RI of 29%. Surface and subsurface lateral flows showed values <1% of gross
30 rainfall (Pg) in either site and were not significantly affected by thinning. On the other
31 hand, soil moisture and drainage were affected by the thinning treatment, although with
32 different extent depending on the site: in the drier site (CAL), the increased Pn in the
33 thinning was mainly allocated into increased soil water content, with very limited
34 improvement in drainage (<10 mm/year); in contrast, in the wetter continental site of HU,
35 drainage to deeper soil layers was the most remarkable effect of thinning (50 mm/year
36 higher than in control), given the higher θ_{st} and hence the lower soil water storage
37 available. Thinning also improved the response of BHP during drought, making these
38 processes more elastic and less vulnerable to climatic extremes. The results presented
39 here complement those previously reported on rainfall partitioning in these sites and all
40 together provide a comprehensive understanding of the short-term effect (3-4 years) of
41 water-oriented silviculture *Quercus ilex* and *Pinus halepensis* low-biomass semiarid

42 forests. Questions such as the long-term effects of thinning remain open for these
43 ecosystems.

44

45 **1. Introduction**

46 Forests affect all components of the water cycle from global to local scales (Ellison et al.,
47 2017), which has prompted for different water-oriented management strategies along the
48 last decades (Hibbert et al., 1982, Sahin and Hall, 1996; Troendle et al., 2001; Ganatsios
49 et al., 2010). More recently, the focus of these strategies has been put on increasing soil
50 water and aquifer recharge rather than increasing runoff and/or streamflow (Creedy and
51 Wurzbacher, 2001; Grant et al., 2013; Ungar et al., 2013; Klein et al., 2013; del Campo
52 et al., 2014; Ilstedt et al., 2016). This is an eco-hydrological approach that considers
53 improving forest resilience through a watering effect on the trees remaining after
54 silvicultural intervention, and it has emerged especially for semiarid forests where the
55 provision of goods and services is threatened by global changes and the related more
56 intense droughts (Allen et al., 2010; Lindner et al., 2014). In any case, water is the key
57 element that modulates either the provision of goods and services or the global and
58 climate stressors affecting semiarid forests, such as increased risk of fire (Hurteau et al.,
59 2008; García-Prats et al., 2015), blue/green water impairment (Grant et al., 2013;
60 González-Sanchis et al., 2015), growth stagnation, higher tree-climate sensitivity, drought
61 stress and mortality (López et al., 2009; García de la Serrana et al., 2015; Fernandes et
62 al., 2016). In the context of *difficult* hydrology (Grey and Sadoff, 2007) characterized by
63 water scarcity, marked seasonality with torrential events followed by long dry seasons,
64 combination of high intra and inter-annual variability, etc., water-oriented forest
65 management should necessarily pay especial attention to both soil water storage and

66 groundwater recharge (Ilstedt et al., 2016; García-Prats et al., 2016), given their key roles
67 in maintaining watershed and ecosystem resilience.

68 In the topic of forest-water relationships and their affection through forest management,
69 the belowground hydrological processes (BHP) have received little attention as compared
70 to the aboveground processes, in spite of the role that forest structure (above and
71 belowground) plays on the dynamics of soil hydrology (Devitt and Smith, 2002; Lin and
72 Zhou, 2008; Coenders-Gerrit, 2012; Bachmair et al., 2012). The manipulation of
73 vegetation has a primary impact on net precipitation at the short term, as increased
74 throughfall and stemflow may enhance soil moisture and the related flows (Taniguchi et
75 al., 1996; Liang et al., 2011; Molina et al., 2019). Thereafter, when biomass is removed,
76 the net rainfall partitioned into soil water replenishment and vertical water flow may
77 locally change, thus affecting lateral water flow (Dung et al., 2012). However, this simple
78 conceptualization becomes more complex in the reality, as the horizontal and vertical
79 heterogeneity of the soil properties in forested hillslopes make predicting soil water
80 content difficult, but also the soil water flows given that the macropores created by plant
81 roots and burrowing animals can be laterally and/or vertically connected over several
82 meters (Lin and Zhou, 2008; Beven and Germann, 2013). In addition, other factors such
83 as antecedent soil moisture, rainfall characteristics, meteorology during rainfall,
84 topography and slope length also play fundamental roles in explaining soil hydrology
85 variability for a particular forest (Gómez-Plaza et al., 2001; Moreno de las Heras et al.,
86 2010; Rosenbaum et al., 2012).

87 These factors make the water redistribution in the soil profile during the infiltration
88 process and the flow generation mechanisms to change in time and space (Calvo-Cases
89 et al., 2003), and they can lead to contrasting results when accomplishing water-oriented
90 forest treatments. Integral assessments of both above- and belowground hydrological

91 processes following vegetation management are really scarce, especially in low-biomass
92 semiarid forests. Such studies are needed in order to understand the relative importance
93 of the driving factors and hence the efficiency and effectiveness of forest treatments in
94 increasing soil drainage and deep-water storage under different climate and soil
95 conditions.

96 Unmanaged high-density forests with low aboveground biomass (e.g. oak coppices and
97 shrublands regenerated after wildfires) are common land-use covers in most dry regions
98 such as eastern Spain, where they occupy about half of the forest area (DOGV, 2013).
99 These areas are prone to climate-related disturbances (López et al., 2009, Doblas-Miranda
100 et al., 2017), stressing the need to better define adaptive treatments for increasing
101 watershed resilience. Likewise, in these water-scarce regions, aquifer recharge and other
102 water-related issues have been identified as environmental services that should be
103 targeted in forest planning and management (DOGV, 2013). Since forest treatments have
104 already been proved to significantly increase net rainfall in these semiarid ecosystems
105 (del Campo et al., 2018), it remains untested how this increase is actually allocated into
106 soil water storage and other BHP. The experimental hypothesis is that the increased net
107 precipitation through thinning/clearing of the low-biomass forests will enhance soil water
108 replenishment, drainage and runoff at the plot scale. The major goal of this work is to
109 study the effectiveness or the degree to which water-oriented forest treatments are
110 successful in improving soil moisture and other BHP. This is assessed by comparing a
111 thinning treatment with an un-thinned one in two forest types growing under contrasting
112 environmental conditions but similar slope. To reach the major goal we address the
113 following questions: i) What are the rainfall characteristics associated with different BHP
114 as well as their relative contribution in explaining these processes together with that
115 related to forest structure and antecedent soil water content?; ii) What is the net impact of

116 forest treatments on the magnitude and rate (as % of gross rainfall) of soil moisture,
117 drainage and run-off?; iii) What are the short to mid-term trends and the cumulated effect
118 in soil water, vertical and lateral flows following silvicultural interventions in the two
119 contrasted forests? These results will help us to define properly forest management for
120 improving water resources in these ecosystems.

121 **2. Materials and methods**

122 *2.1. Study sites and treatments application*

123 The study was carried out in two environmentally contrasted sites with marked
124 differences in their climate-forest-soil aggrupation, although in both cases a low-biomass
125 forest type (Pan et al., 2013) with high tree density and competition (overstocked) is
126 present. Both sites are located in the Valencia province (E Spain) and distant about 100
127 km one to the other. The Calderona site (CAL, 39°42' N 0°27' W, 790 m a.s.l.), located
128 within the Natural Park of “Sierra Calderona”, has a marked influence of the
129 Mediterranean Sea (25 km away), and the vegetation consists of an Aleppo pine (*Pinus*
130 *halepensis* Mill.) stand of saplings with sparse shrubs, which was regenerated after a
131 wildfire occurred in 1992. The south-westernmost site (HU, 39°04' N 1°14' W, 1090 m
132 a.s.l.) is located in “La Hunde” public forest. This site has a more pronounced continental
133 climate and it is occupied by a marginal oak coppice (*Quercus ilex* subsp. *ballota* (Desf.)
134 Samp.). Both sites have been previously defined in terms of vegetation, climate, soils and
135 other bio-geographical traits (del Campo et al., 2018, 2019). Table 1 summarizes key
136 information according to the objectives of this work.

137 In each study site, detailed information about the depth and features of the soil-rock
138 stratum (depth-to-rock, bedrock weathering and shallow groundwater routing) was
139 gathered through a 2D electrical resistivity tomography (Samouëlian et al., 2005). The
140 work consisted of twelve (six per site) 2D electrical tomographies with 48 electrodes,

141 following a Wenner-Schlumberger array configuration by using a Syscal Pro Switch
142 resistivitymeter with 48 channels (Iris Instruments, Orleans, France). The 2D transects
143 were processed at 2.5D to obtain a system of 3D geo-resistive volumes by the software
144 of ERTLab Solver and RES2DINV (v.36.06).

145 In both sites no forest management has been carried out in the last decades. Juvenile
146 thinning with shrub clearing were executed by a contractor of the Forest Service. The
147 thinning removed the trees with smaller diameters and doubled-stemmed trees, trying to
148 achieve a relatively homogeneous forest cover distribution. CAL was treated between
149 January and October 2012, and HU in May 2012. Total basal area removed was 74% and
150 41% for CAL and HU, and the tree density reduction was of 94% and 73%, respectively
151 (Figure 1, Table 2). Coarse woody debris were placed outside the plots, whereas fine
152 woody debris were piled and grinded into mulch onto the plots. In each experimental site,
153 a representative control plot with no thinning treatment and a contiguous thinned plot
154 were established; each experimental plot having an area of 1500 m².

155 Forest structure was characterized in the experimental plots as detailed in del Campo et
156 al. (2018). The variables considered were basal area (BA, m² ha⁻¹), tree density (De, trees
157 ha⁻¹), forest cover (FC, %), leaf area index (LAI, m² m⁻²) and diameters at basal and breast
158 heights (D_B and D_{BH}, respectively, cm) (Table 2).

159 FIGURE 1

160 2.2. *Rainfall and meteorological variables*

161 Measurements were carried out from October 2013 to September 2016 in CAL and from
162 October 2012 to September 2016 in HU. In each experimental site, a data-logging system
163 was installed (CR1000, Campbell Sci., UT, USA) to register and store data, supplemented
164 with two AM16/32B multiplexers, two SDM-IO16 expansion modules, a solar panel and

165 a 12 V battery. The system was programmed to read the sensors output every 5 s or 1 min
166 and to record the values every 5 s or 10 min.

167 Gross rainfall (P_g) was continuously measured in each site at a height of 6 m above the
168 ground by means of a tipping-bucket rain gauge with 0.2-mm resolution (7852, Davis
169 Instruments Corp., Hayward, CA, USA). Rainfall events were classified according to a
170 minimum time of 1 hour between two successive rains (Llasat, 2001) with at least 1 mm
171 of rainfall depth (Muzylo et al., 2012). The following event variables were calculated for
172 10-minutes intervals: total rainfall depth (P_g , mm), duration (P_D , min), maximum and
173 mean rainfall intensity (P_1 and P_{Imx} , mm/h) and intra-event gaps as the proportion of time
174 within the event with no rainfall (P_{Gap} , %). Also, in order to characterize the convective
175 nature of the rainfall, we computed the β parameter (ratio of high intensity rainfall to total
176 rainfall) for each rainfall event, adopting a threshold of 50 mm/h during 1 min (Llasat,
177 2001). For a time period of 10 min, this value dropped to 23.8 and 21.5 mm/h for CAL
178 and HU, respectively. In this way, we computed the $\beta_{23.8,10}$ and $\beta_{21.5,10}$ parameters for
179 each event and site ($\beta=0$ non-convective; $0<\beta\leq 0.3$ slightly convective; $0.3<\beta\leq 0.8$
180 moderately convective and $0.8<\beta\leq 1.0$ strongly convective).

181 The meteorological conditions during rainfall were characterized by measuring mean and
182 maximum wind speed (U_{av} and U_{mx} , 7911 anemometer, Davis Instruments Corp.), air
183 temperature and relative humidity (T, RH sensor, Decagon Devices, Pullman, USA) and
184 vapour pressure deficit (VPD). Sensors were placed on the same mast as the rainfall
185 gauge, and data were recorded in the same time intervals as those for rainfall.

186 Throughfall and stemflow were simultaneously measured as described in our previous
187 work (del Campo et al. 2018). Stemflow was measured by sealing stem collars to 9-12
188 trees per treatment. After the plastic collars were attached, plastic tubes were inserted into
189 small holes located in the lowest part of the collars to collect the water and divert it to

190 tipping-buckets (Pronamic 100.05) and then to 25-L deposits. Throughfall, was measured
191 by setting out galvanized steel gutters per plot (9-15 per treatment), arranged following
192 contour lines and maintained in the same positions throughout the study. The devices
193 were set at 50 cm above the soil and sloping towards water counters (Altair V4, Diehl
194 Metering) equipped with pulse counters. Net rainfall (Pn, mm) was computed for each
195 event by the sum of throughfall plus stemflow, and rainfall interception was computed by
196 subtracting net rainfall to Pg.

197 *2.3. Soil water content and drainage*

198 The electrical tomography revealed that soil depth was very low in most of the cases with
199 the exception of the lower part in the plots of the CAL site, where a soil accumulation
200 was present and therefore more water holding capacity is expected (Figure 2). This is in
201 agreement with auger samplings and direct evidence from the observation of outcrops
202 and stoniness in the plots. Therefore, the below-described layout of soil moisture sensors
203 and piezometers was aimed to keep a comparable, balanced and equally depth-weighted
204 deployment of readings between treatments and sites.

205 FIGURE 2

206 Soil water content (θ , $\text{m}^3 \text{m}^{-3}$) was continuously measured every 10 min, or every 5 s when
207 raining, by means of FDR (frequency domain reflectometry or capacitive) probes (EC-5,
208 Decagon Devices Inc., Pullman, WA) connected to a CR1000 data-logger. Sensors were
209 installed by digging nine pits per plot (18 per experimental site) and grouped into three
210 groups following contour lines. In each group, one of the pits contained two sensors poked
211 horizontally at depths of 15 and 30 cm into the unaltered upslope pit face, whereas in the
212 other two pits one sensor was inserted at 15 cm depth (12 sensors per plot, 24 sensors per
213 experimental site). The pits were regularly placed following a mesh grid of 10 x 10 m in
214 order to obtain a good estimate of mean soil water content (Molina et al., 2014). After

215 installation, the pits were backfilled with the excavated soils and slightly compacted up
216 to similar bulk density to that of unaltered soil. Calibrations were discarded as the
217 observed increase in θ after calibrating some sensors was offset when taking into account
218 the stoniness effect in field conditions, and because of the temporal drifts observed in
219 several cases, preventing the use of a static calibration curve; thus, we used default
220 calibration (for mineral soils) in all cases (Detty and McGuire, 2010).

221 For each rainfall event and probe, we considered the first (θ_{st}) and the maximum (θ_{mx})
222 values of soil moisture. Field capacity for each probe (θ_{fc}) was calculated for gentle
223 rainfall events (>20 mm, usually the daily rainfall was much higher because of the
224 threshold considered for separating events), collected at night-time, by averaging the
225 (steady) values of θ during at least four hours once the exceeding water was lost. The
226 steady values were assessed to be preceded by a steep decrease (exceeding water) and
227 then reaching a constant value during at least four hours without rainfall at night time (the
228 decrease rate in 240 min was found to be less than 0.0034 and 0.0027 m³/m³ for
229 Calderona and La Hunde sites respectively). However, we observed that these θ_{fc} values
230 were not persistent as some probes showed temporal drifts (means \pm standard deviations
231 during the study period were 0.167 ± 0.043 m³/m³ and 0.192 ± 0.037 m³/m³ for CAL and
232 HU, respectively). Thereafter, we recomputed θ_{fc} for each sensor whenever P_g surpassed
233 the established 20 mm threshold (9 and 13 times for CAL and HU, respectively).

234 Soil water replenishment ($\Delta\theta$, m³ m⁻³) around a probe was computed for every rainfall
235 event as the difference between the θ maximum and first values, as $\Delta\theta = \theta_{mx} - \theta_{st}$ when
236 field capacity was not reached or as $\Delta\theta = \theta_{fc} - \theta_{st}$ when it was. $\Delta\theta$ data were converted to
237 $\Delta\theta$ depth (mm) by considering the soil column depth. When looking at the wetting process
238 of the soil profile, we observed in our dataset that soil depth differences were irrelevant,
239 showing a non-uniform behaviour probably led by preferential vertical channels through

240 macropores (Calvo-Cases et al., 2003; Lin and Zhou, 2008). Thus, we decided to use the
241 weighted depth of the moisture sensors (187.5 mm) to calculate soil water replenishment
242 for each sensor and subsequently average across the 12 probes to obtain the $\Delta\theta$ depth for
243 each experimental plot. Values of $\Delta\theta$ calculated in this way are likely to be
244 underestimated, as hydrological processes at the upper soil layers are not fully addressed
245 (e.g. light-rainfall events, litter interception, higher soil moisture holding capacity in the
246 upper organic horizons, etc.).

247 Drainage (D , mm) was estimated based on θ and P_g observations. For each probe, all P_g
248 collected when the probe was above θ_{fc} was considered drainage. Rainfall interception
249 was assumed to have already taken place by the time the sensors reached θ_{fc} . In case the
250 resulting drainage amount was higher than the mean net precipitation estimated for the
251 experimental plot (del Campo et al., 2018), then D was reset to that value. In case there
252 was run-off (see below), it was deducted from D . We established a threshold to consider
253 a spatially generalized drainage (D_{gen} , mm) by considering a minimum of 2/3 out of the
254 12 FDR sensors showing values above θ_{fc} in each experimental plot. As different sensors
255 entered into field capacity at different moments, we considered the mean P_g from these
256 *active* sensors for calculations. Drainage calculated in this way is independent from the
257 above-mentioned uncertainty in calculating $\Delta\theta$ depth and only depends on field capacity
258 and P_g , both with a more robust estimation. Also, the difference $D-D_{gen}$ can be
259 considered as a surrogate of soil water storage below 18.75 cm.

260 In addition, four 10 cm-diameter piezometers (from 1 to 4 m depth) were drilled at the
261 downslope side of each experimental plot (eight per site) according to the information
262 from tomography. These wells were aimed to detect deep infiltration and validate the
263 estimations of drainage based on θ and P_g observations. A slotted screen PVC pipe was
264 inserted into each well that was blind on its top 60 cm to avoid shallow flows getting in.

265 Also, a 10 cm layer of coarse bentonite was inserted at the top of the well around the pipe.
266 Every 1-3 weeks (depending on rainfall), we made qualitative checks of groundwater
267 movement by visual identification of the water depth in a transparent bailer placed inside
268 the wells (Model 428 BioBailer, ESP ltd.). Malfunctioning of the bailer in HU during
269 most of time led to implement automatic readings of water level (PX40-15G5V, Omega
270 Eng. ltd.) in both sites at the end of the study period, thus allowing for a more reliable
271 validation of the estimations.

272 *2.4. Lateral soil water flow*

273 Lateral soil water flow was collected by 30-40 cm depth trenches dug downslope in each
274 experimental plot. The upslope side of the trenches was open to subsurface flows by
275 attaching a fine mesh filter directly to the ground exposed to air (downslope and base
276 sides of the trench were water-proofed with concrete) (Figure 1). The surface and
277 subsurface flows (hereinafter considered as runoff, RO, to give more fluency to text)
278 collected were diverted to deposits equipped with automatic pulse counters (Altair V4,
279 Diehl Metering, minimum flow rate 5 l h⁻¹). Two 6 m long ditches per experimental plot
280 were dug in HU, whereas in CAL a single 35 m long ditch per experimental plot was dug
281 edging the lower boundary of the plots. The measurement time lapse was set to 5 s
282 (raining) or 10 min (no raining). Similar trenches were dug upslope on the upper boundary
283 of the plots in order to divert off the run-on onto the plots.

284 *2.5. Data treatment and analysis*

285 Data were quality-controlled for spikes and gaps (del Campo et al., 2018). To study the
286 relative contribution of rainfall, meteorological conditions during rainfall, antecedent soil
287 moisture and forest structure to the different BHP ($\Delta\theta$, D, Dgen, and RO), boosted
288 regression trees models (BRT) were performed in R software (RStudio Team, 2015),
289 using the “gbm” package (Ridgeway, 2017; Elith and Leathwick, 2017). In the BRT

290 analysis, a Gaussian distribution family, learning rate of 0.005, tree complexity of 4-5,
291 and bag fractions of 0.6-0.75 were considered. The minimum number of trees was in all
292 cases above 1500. The results of this analysis provided the relative influence (RI) of the
293 predictors on the response variable. This measures the number of times a predictor
294 variable is selected for splitting, weighted by the squared improvement in the model as a
295 result of each split, averaged over all trees, and scaled so that the sum adds to 100 (Elith
296 et al., 2008). The higher the RI, the stronger the influence of the predictor in the response
297 variable.

298 The effect of the forest treatment on depth and rates of the response variables ($\Delta\theta$, D,
299 Dgen, and RO), was analysed in each site through an ANCOVA analysis considering
300 treatment as factor and Pg as covariate. θ_{st} and θ_{mx} were also included in this analysis
301 after normalizing them to their relative values (θ_{st_rel} and θ_{mx_rel} , %, dividing their current
302 values by the overall range of each probe during the monitoring period). Because of the
303 inevitable heterogeneity of soil properties among different probes (texture, stoniness and
304 bulk density) makes that the water holding capacity of the soil may change, this
305 normalization ensures that the comparison between thinning and control treatments is
306 unbiased. Data were examined for normality, homoscedasticity and homogeneity in the
307 regression slopes and transformed with power functions if necessary or, alternatively, a
308 non-parametric Kruskal–Wallis test based on the chi-squared statistic was used.

309 Cumulative treatment impacts on the response variables were evaluated in terms of a shift
310 of the daily ratio of thinned/control following the intervention: L (T/C) (Perry and Jones,
311 2017). The water balance of net rainfall (Pn, mm) for each water year and experimental
312 plot was calculated as: $Pn = RO + \Delta\theta + D + U$, where Pn is the annual cumulated value
313 of the sums of throughfall and stemflow for each rainfall event calculated in del Campo
314 et al. (2018); RO is the annual cumulated runoff value; D is the annual cumulated value

315 of drainage; $\Delta\theta$ is the annual cumulated value of soil water replenishment; and U is an
316 unexplained term which cannot be explained and is mostly influenced by hydrological
317 processes taking place in the uppermost soil layers (soil water replenishment in the upper
318 organic layers and interception from litter, grass and dwarf shrubs).

319 Finally, in a context of rainfall scarcity and irregularity, the effectiveness of the treatment
320 can be assessed through observing the proportionality of the change in a target
321 hydrological process with regards to the change in Pg. To that end we performed an
322 elasticity analysis for soil moisture and drainage in order to address how the annual
323 variability in Pg is translated into relative changes in these key hydrological variables,
324 and how the treatment affected it. The mean annual θ_{mx_rel} , the cumulated annual Dgen
325 and the ratio $(Dgen+RO)/(Pg-Dgen-RO)$ as a proxy to the blue/green water ratio
326 (González-Sanchis et al., 2015) were used in this analysis. To that end, we used a similar
327 approach to that of Jiang et al. (2014) but comparing the different water years with the
328 wettest one instead of using the average of the series. We mostly registered dry years in
329 both sites, so by doing this selection we remain close to the expected average (Table 1),
330 and at the same time we tested for changes towards drier conditions. Elasticity (e) is
331 defined then as the proportional change in a given hydrologic variable (X) in response to
332 the proportional change in annual precipitation, and it is expressed as:

$$333 e_x = \text{Median} [(X_t - X_w) / (P_t - P_w) * P_w / X_w]$$

334 where X_w and P_w are the totals of the hydrologic variable and the precipitation in the
335 wettest water year of the series, and X_t and P_t are totals of the variable and precipitation
336 at any other given year t, respectively. High degree of elasticity ($e > 1$) means that the
337 variable X is very sensitive or vulnerable to a proportional change in P; unit elasticity
338 ($e \approx 1$) means that a relative change in P is translated into a proportional change in a given
339 variable (X); low elasticity ($e < 1$) or inelastic ($e \approx 0$), occurs when little or no change is

340 detected in the hydrological variable for a given change in P. This simple approach helps
341 to uncover, directly from observational data, the direction and strength with which
342 different hydrological variables respond to declining P either under managed or
343 unmanaged scenarios (Jiang et al., 2014).

344 **3. Results**

345 *3.1. Characteristics of gross rainfall and meteorological conditions*

346 Gross rainfall results have been also reported in del Campo et al. (2018). Briefly, total Pg
347 in CAL was 162, 370 and 246 mm for the water years 2013-2014 to 2015-2016
348 respectively, whereas 534, 271, 426 and 297 mm for the years 2012-2013 to 2015-2016,
349 respectively in HU. Non-convective rainfall events presented higher depth and duration
350 in HU (Table 3); however, Pg presented a higher convective pattern in CAL, with 38%
351 of total Pg fallen in this class (11% in HU), and higher Pg, P_D, P_I and P_{Imx}. Furthermore,
352 the convective rainfall events were accompanied with more evaporative conditions (Table
353 3).

354 The number of events that provoked either drainage or runoff regardless of the treatment
355 was higher in HU (118-149) when comparing with the coastal site of CAL (25-41) (Table
356 3). In HU, it is remarkable that the features of these events were only slightly higher than
357 the average rainfall event (Table 3): the ratio to the average event ranged between 1.2 and
358 1.5 for depth, duration and intensity. In contrast, there was a clear departure of these
359 events from the depth, duration and intensity of the average event in CAL (multiplicative
360 factor between 1.5 to more than 3.0). These differences in the threshold of Pg that trigger
361 the different hydrological processes between both sites were also associated to different
362 antecedent soil water conditions (see below).

363 *3.2. Influence of rainfall, meteorology, forest structure and antecedent soil moisture on* 364 *soil water replenishment, drainage and runoff*

365 The four sets of independent variables considered (rainfall characteristics: P_g , P_D , P_I , $P_{I_{mx}}$,
366 P_{Gap} , β ; meteorology during the event: T , VPD , RH , U_{mx} , U_{av} ; forest structure: FC , BA ,
367 LAI , De ; and antecedent soil moisture θ_{st_rel}), presented significant correlations with the
368 studied BHP ($\Delta\theta$, D , D_{gen} and RO) in both sites (Table 1S in supplementary material).
369 The analyses performed through BRT models yielded better cross-validation correlations
370 for the depth-expressed (mm) variables than when they were expressed by rates (% P_g),
371 and those tested for CAL showed better fits than for HU in both cases (Table 4). The
372 relative importance (RI) of the four sets of independent variables on explaining the
373 hydrological processes is shown in Figure 3, weighted by the corresponding correlation
374 coefficients from Table 4 (Figure 3S shows the RI individually for the most influential
375 independent variables). It is remarkable in Figure 3 the overwhelming importance of the
376 rainfall variables (mostly P_g , followed by P_D and $P_{I_{mx}}$) on the magnitude (mm) of the
377 different BHP, with a cumulative RI between 30-76%, depending on the site/dependent
378 variable. On the contrary, the contribution of these variables to the rates was modest
379 regardless of the site (cumulative RI between 5-29%). The meteorological conditions
380 during rainfall and the antecedent soil moisture showed modest RI's for depths (2-21%
381 and 5-25% in CAL and HU respectively) but increased their RI for rates in both sites
382 (between 11-43%). Forest structure remained with low RI whatever the case, although it
383 improved when combining both sites in the model (RI for $\Delta\theta$ as % P_g was 8.3%)

384 FIGURE 3

385
386 *3.3. The effects of thinning on soil water, drainage and runoff*

387 *3.3.1. Effects at the event scale*

388 The difference between thinned and control plots for the different variables describing
389 the net rainfall redistribution processes was tested through ANCOVA test (see Table 2S
390 for a summary of the analyses). CAL presented only significant differences in relative

391 soil water content both at the first (θ_{st_rel}) and the maximum (θ_{mx_rel}) values, showing
392 higher values in the thinned plot (Figure 4); thus, no further effect of thinning was
393 significantly detected in CAL, although the averages of $\Delta\theta$, D, Dgen and RO were higher
394 in the thinned plot (Figure 5). In contrast, the thinned plot in HU presented significant
395 differences with the control in all the BHP considered except for RO (Table 2S, Figures
396 4 and 5).

397 FIGURE 4

398 FIGURE 5

399 3.3.2. Temporal dynamics and annual balances

400 The temporal dynamics of $\Delta\theta$, D, Dgen and RO for both plots and sites are shown in
401 Figures 6 to 8. Regarding RO, although it represented less than 1% of total Pg, its
402 dynamics were markedly different between the sites. CAL presented very few RO events,
403 basically responding to high-intensity storms (Table 3) and showing in some cases
404 overland hortonian flow (e.g. the event in Aug 10th, 2016 with Pg=19.0 mm, $\beta=0.41$,
405 $P_D=167$ min, $P_{Imx}=32.1$ mm/h, $P_I=6.8$ mm/h). In contrast, RO in HU was observed within
406 a higher range of rainfall characteristics, indicating a higher dependency on antecedent
407 soil moisture content (e.g., an event observed in Jan 29th, 2014 with only 2.8 mm of Pg
408 generated RO in both the control and thinned plots).

409 Regarding the dynamics on drainage flows, Figure 8 shows that CAL presented a more
410 static pattern with little or no differences in either D or Dgen between both plots
411 ($\ln(D_{genT}/D_{genC})\approx 0$), but a clear differentiation in HU for this flow, in agreement with
412 the results of the ANCOVA. With regards to $\Delta\theta$, the lower θ_{st} in the control plot in HU
413 made more infiltrated water to be allocated into soil water replenishment (so that
414 $\ln(\Delta\theta_T/\Delta\theta_C)<0$, Figure 8), whilst higher θ_{st} in the thinning treatment (closer to θ_{fc}) reduced
415 the availability for soil water storage. In contrast, θ_{st} in CAL was lower in both plots and

416 the increased Pn led by thinning was almost fully allocated to increase this state variable,
417 thus making the series $\ln(\Delta\theta_T/\Delta\theta_C) > 0$ (Figure 8). Also, whereas in HU the temporal
418 pattern showed a diminishing trend of the series $\ln(\Delta\theta_T/\Delta\theta_C)$ with time, CAL presented a
419 static pattern mainly due to the lack of rainfall (Figures 6 and 8).

420 The comparisons between the underground water movement from observations in the 1-
421 4 m depth piezometers and the estimations of drainage flows based on θ and Pg agreed
422 well in both sites and plots (Figures 6 and 7), thus validating the occurrence of D and
423 Dgen. In CAL, the rise of the water level in the piezometers during the period they were
424 monitored matched 100% of the events presenting either D or Dgen in the thinned plot,
425 and most of the estimated D (95%) and Dgen (95%) was paired with a response in the
426 wells for the control plot. In HU, with less validation period, the wells responded for 92%
427 of the calculated D flow in either plot, and 96% of Dgen in the thinned plot; however, the
428 lower matching between the response of the piezometers and Dgen in the control plot,
429 means that in this site with shallow soil, higher Ks and organic matter (Table 1), D is a
430 good estimator of total drainage.

431 FIGURE 6

432 FIGURE 7

433 FIGURE 8

434 The annual balances in Figure 9 represent the cumulative values of the BHP (RO was
435 negligible in all cases and therefore not represented) sorted by water years, sites and plots.
436 It is remarkable how the cumulative Pn increase led by thinning was differently translated
437 into $\Delta\theta$ and drainage, thus agreeing the previous event-based results. In CAL, D was very
438 similar between the experimental plots (mean values of 89 vs. 97 $\text{mm}\cdot\text{year}^{-1}$ for C and T
439 respectively), whereas in HU percolation was always higher in the thinned plot (mean
440 values were 118 vs. 169 $\text{mm}\cdot\text{year}^{-1}$ for C and T, respectively). In contrast, the differences

441 in $\Delta\theta$ between T and C ranged from 10 to 52 mm in CAL and between -9 and -75 mm in
442 HU. In addition, the unexplained term ranged between 23-51 mm (8-31% of Pn) in CAL
443 and 2-114 mm (2-29% of Pn) in HU.

444 FIGURE 9

445 3.3.3. Elasticity analysis

446 The elasticity analysis indicated that the implementation of silvicultural treatments
447 improved the elasticity for Dgen and the ratio blue/green water (B/G) in dry years with
448 regards the reference year (near the average P_g) (Figure 10). By sites, in HU higher
449 vulnerability in Dgen and B/G in the control ($e > 1.65$) in dry years turned into a more
450 proportionate change to P_g when management took place ($e \approx 1$). In CAL, with very low
451 values observed in both plots either for Dgen or B/G, the thinning treatment changed the
452 elasticity from vulnerable ($e = 1.25$ in Dgen) or insensitive ($e = 0.68$ in B/G) to elastic for
453 both variables ($e \approx 1.1$). Regarding θ_{mx_rel} (i.e., the maximum soil hydration achieved in an
454 event, relativized, so that soil properties do not affect this value), forest management
455 turned it more insensitive ($e < 1$) to P_g decreasing. In this sense, as the thinning improved
456 significantly θ_{mx_rel} in both sites, this change in elasticity is certainly desirable (soil
457 moisture is still higher despite years become drier).

458 FIGURE 10

459 4. Discussion

460 A previous work regarding the effects of thinning on rainfall partitioning (del Campo et
461 al., 2018) showed that the reduction of forest biomass led to significant effects on net
462 rainfall (P_n) in both study sites, increasing from 554 to 634 mm in CAL and from 1088
463 to 1294 mm in HU when looking at the cumulated values in C and T respectively within
464 the same study period (cumulated P_g was 779 and 1528 mm for CAL and HU,
465 respectively). The present work is thus focused on further studying the distribution of P_n

466 into soil, finding that thinning affected most of the belowground hydrological processes
467 (BHP) in a contrasted way between both sites, which confirms partially our hypotheses.

468 *4.1. Driving variables of the belowground hydrological processes*

469 Forest cover reduction through thinning enhances P_n mainly due to an increase in
470 throughfall (Chirino et al., 2006; Molina and del Campo, 2012; del Campo et al., 2018),
471 so that the question that arises is how this increased throughfall is partitioned into the
472 different BHP (Coenders-Gerrits et al., 2012). In our case, the different types of
473 independent variables considered in the BRT were shown to affect these BHP to
474 contrasted extents when expressing the variables either as depth or rate. However, forest
475 structure showed very low RI in either case, which contrasts with the significant effects
476 observed in the ANCOVA's. In order to explain this conflicting result, we must pay
477 attention to the antecedent soil moisture (θ_{st_rel}), which was significantly improved by
478 10% (about 5% in absolute value) with the thinning treatment (Figure 4), and it was the
479 most important variable in explaining the rate of P_g which is diverted into the different
480 BHP (RI of 29% on average, Figure 3). This means that the eco-hydrological impact of
481 forest management should be better evaluated in terms of changes in the soil water
482 content, which is in fact one of the most important changes following thinning treatments
483 (Breda et al., 1995; del Campo et al., 2014; Cabon et al., 2018; del Campo et al., 2019).
484 Other independent variables were also brought up in the BRT as key drivers of the
485 different BHP. The control that rainfall exercises on the magnitude of water components
486 in soils is recognized elsewhere (Calvo-Cases et al., 2003; Chirino et al., 2006; Owor et
487 al., 2009; Bachmair et al., 2012), and in our case, it played the most important role in
488 explaining the depth-expressed variables (RI>75% in some cases, Figure 3), with the
489 magnitude of P_g being by far the most explicative variable, followed by P_D , $P_{l_{mx}}$ and β
490 (Figure 3S). From these results, it can be deduced that certain thresholds and patterns in

491 the rain features and meteorology that characterize a particular rainfall regime are
492 important on evaluating the effectiveness of water-oriented forest treatments. For
493 instance, the partial dependence of $\Delta\theta$ on rain convection presents a threshold at about
494 $\beta > 0.3$ (Figure 3S).

495 *4.2. Lateral soil water flow*

496 Runoff in mature forest stands or wooded vegetation ecosystems with high biomass
497 quantity is generally low given both the high soil organic carbon content and the high
498 infiltration capacity led by macropores (Beven and Germann, 2013). Chirino et al. (2006)
499 obtained runoff coefficients less than 1% of P_g in a 30-year-old semiarid Aleppo pine
500 plantation, while Merino-Martin et al. (2012) and Moreno-de las Heras et al. (2010)
501 observed runoff coefficients less than 4.5 % in a reclaimed slope with natural regenerated
502 grassland covering more than 50% of soil. In our case, the mean annual coefficients
503 remained below 1% of gross rainfall for C and T plots in both sites, and they accounted
504 not only for runoff, but also for subsurface flows. Cerdá et al. (2017) reported values of
505 this water component averaging 8.3% of P_g on Aleppo pine and 1.4% on Holm oak plots
506 (1m^2) in a site similar to that of CAL regarding soil, vegetation and meteorological
507 conditions (slope gradient was not reported). The required depth of an event in CAL (pine
508 stand) to originate runoff (mean of 16.3 mm, Table 3) was lower than in the referred work
509 (minimum of 18.7 mm) despite the fact that runoff was much higher in that study. In this
510 respect, scale effects (1500 vs. 1 m^2) seem to be the most likely reason of this discrepancy.
511 Moreno-de las Heras et al. (2010) reported in a dry Mediterranean site with similar soil
512 organic matter to ours, a generalized decrease of 79% in runoff per unit area with
513 increasing slope length from 1 to 15 meters. As these authors pointed out, the locally
514 generated runoff re-infiltrates farther downslope depending on soil degradation, micro-
515 topography and rainfall conditions, and the same factors together with the macroporosity

516 structure are expected to control subsurface flows (Lin and Zhou, 2008). Our data on soil
517 water content suggest an important macropore flow, with similar values between the FDR
518 probes at different depths regardless the time during several rainfall events. By the same,
519 the available automatic readings from the piezometers showed that deep infiltration began
520 almost simultaneously with rainfall (Figure 4S).

521 In this regard, the results indicate that thinning did not significantly affect the soil
522 characteristics in the uppermost layers controlling surface and subsurface water flows,
523 given our non-significant differences on RO (Figure 5 and Table 2S). In this sense,
524 measurements of ponding and tension infiltration in the holm oak site (HU) confirmed
525 that both saturated and unsaturated hydraulic conductivity were not significantly different
526 between the plots (di Prima et al., 2017).

527 *4.3. Soil moisture replenishment and drainage*

528 The most relevant finding observed in the present study is the differential effect that
529 thinning produced in $\Delta\theta$ and drainage between both sites (Figures 6 to 8). Increased Pn
530 in the thinned plot of CAL led to higher $\Delta\theta$ in a relatively dry soil ($\theta_{st_rel}=45\%$), but
531 negligible or little further movement of this water below the upper soil layers. The reason
532 is that less and sparser rainfall events together with higher evaporative conditions in this
533 site provoked low antecedent soil water content in both plots, and hence significant
534 amounts of net precipitation were required to surpass θ_{fc} (full soil water replenishment).
535 Consequently, only those rainfall events with enough volume were able to gently
536 overcome that threshold (mean Pg for Dgen was 13 mm), but at the same time these
537 showers quickly saturate the canopy resulting in little differences in Pn between C and T,
538 so that θ_{fc} is exceeded in either case, and drainage was non-significantly different between
539 plots. In contrast, HU showed a marked rainy season with low evaporative conditions,
540 leading to higher θ_{st} ($\theta_{st_rel}=66\%$) in the thinning treatment, and the increased Pn here

541 moved downward as drainage, as θ_{fc} was surpassed in more rainfall events (mean P_g for
542 D_{gen} was 7 mm). Drainage is usually associated to consecutive rainy days out of the
543 growing season (García-Prats et al., 2016), a pattern that is almost absent in the drier site
544 of CAL. The important consequence is that, when managing forests for increasing
545 infiltration (Ungar et al., 2013; del Campo et al., 2014; Ilstedt et al., 2016), changes in the
546 soil moisture regime are the critical factor that triggers when the increased P_n after
547 thinning turns into deep water reservoir. It must be stated that the drainage rates measured
548 in our plots do not necessarily reflect current recharge rates to downslope streams or water
549 tables, but rather deep-water storage for ET, as plant-water uptake cannot be considered
550 negligible at the depths observed in our plots (del Campo et al., 2019).

551 *4.4 Temporal dynamics and elasticity*

552 In the drier site of CAL, both low rainfall and the subsequent slow growth of the
553 vegetation have made the temporal series $L(T/C)$ remaining with little change for the
554 BHP, as already observed for the rainfall partitioning (del Campo et al., 2018). On the
555 other hand, in HU $L(T/C)$ showed changing temporal trends in the BHP reflecting the
556 impact of wet/dry spells, together with the combined effect of tree growth especially in
557 the thinned plot (del Campo et al., 2019): $L(D_T/D_C)$ remained relatively steady along the
558 4-year period, whereas in the case of D_{gen} , it can be observed in Figure 8 that T and C
559 did not differ in the driest spell of 2013-2014 ($L(D_{gen_T}/D_{gen_C}) \approx 0$), although when wetter
560 conditions returned, the series became positive again. This absence of differences under
561 dry conditions shows similar pattern to that observed in CAL and has also been previously
562 modelled for Aleppo pine in HU (González-Sanchis et al., 2015). It is also worth to
563 mention that during this dry spell (Aug-2013 to Oct-2014), the series $L(\Delta\theta_T/\Delta\theta_C)$
564 becomes more negative, indicating that the drier soil in the control retains more net
565 rainfall as $\Delta\theta$, whereas the opposite pattern for $L(D_T/D_C)$ indicates more water allocated

566 to drainage in the thinning treatment. This implies that even in dry years, deep soil
567 moisture is improved with thinning, a fact that was brought up in the elasticity analysis.
568 From late 2014 on, the growth of the oaks speeded up in the cleared plot and further
569 affected soil moisture patterns (del Campo et al., 2019).

570 Vegetation management in semiarid shrublands is known to be of limited effectiveness
571 when aiming to increase water yield at the catchment scale (Bosch and Hewlett, 1982;
572 Hibbert et al., 1982), but it may be certainly opportune when dealing with global change-
573 induced impacts by enhancing ecosystem resilience (Grant et al., 2013; Lindner et al.,
574 2014; Seidl et al 2016). In this sense, our results proved that the irregular and dry
575 conditions that prevailed throughout the study impacted differently in control and thinned
576 plots as revealed on the elasticity of key eco-hydrological variables (soil moisture and
577 deep-water reservoir). Although the comparison between control and thinning treatments
578 is objective and robust, some uncertainty still persists in our study, as part of the net
579 rainfall was unexplained when totalizing the studied BHP. Since our sensors were
580 installed at 15 and 30 cm depth, soil evaporation from the surficial layers and litter
581 interception are likely the most contributing factors to this unexplained term, given their
582 importance in semiarid environments (Tsiko et al., 2012; del Campo et al., 2019).
583 Therefore, further measurements are required to fully account for soil water content
584 variation.

585 **5. Conclusions**

586 Proactive-adaptive silviculture needs to be necessarily eco-hydrologically oriented in the
587 case of semiarid regions. Our work evaluated the short-term effect of thinning on the
588 redistribution of net rainfall into the soil of two semiarid forests with contrasted rainfall
589 and vegetation characteristics. Lateral water flows were negligible and not significantly
590 affected by tree removal in either site, so the protective role of forest in controlling soil

591 erosion was not modified through thinning. In contrast, soil water replenishment and
592 drainage were affected to different extents depending on the site, highlighting that, under
593 the same semiarid climate and low-biomass vegetation, responses to water-oriented
594 treatments can be very different according to local/regional variations in climate and
595 forest-types: while in the most arid conditions thinning improved soil water content and
596 made it more independent from annual changes in Pg, drainage was only enhanced in the
597 wetter site (HU) given the higher antecedent soil water conditions. Thus, under the
598 expected global change-induced impacts on vegetation due to increased soil water
599 shortage in semiarid environments, biomass reduction seems to be an effective tool for
600 adapting forest, and in some cases, also for increasing deep soil water reservoirs. Further
601 questions to be addressed are the mid-term duration of the effects and the intensity of the
602 treatments, the later to test whether drainage can be positively affected in the most arid
603 conditions when higher biomass reduction is considered.

604

605 **Acknowledgements**

606 This study is a component of research projects: HYDROSIL (CGL2011-28776-C02-02),
607 SILWAMED (CGL2014-58127-C3-2), CEHYRFO-MED (CGL2017-86839-C3-2-R)
608 funded by the Spanish Ministry of Science and Innovation and FEDER funds and LIFE17
609 CCA/ES/000063 RESILIENTFORESTS. The authors are grateful to the Valencia
610 Regional Government (CMAAUV), VAERSA, ACCIONA, the “Sierra Calderona”
611 Natural Park and the communal authority of Serra, for their support in allowing the use
612 of the experimental forest and for their assistance in carrying out the fieldwork. A.J.
613 Molina is beneficiary of a “Juan de la Cierva” post-doctoral fellowship funded by the
614 Spanish Ministry of Economy and Competitiveness.

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803

804 TABLE CAPTIONS

805 **Table 1.** Physiographic, climatic and edaphic features in both experimental sites
806 (Calderona, CAL and Hunde, HU). ^aFractions in the following order: sand (2–0.02mm),
807 silt (0.02–0.002mm), clay (<0.002 mm). ^b Intervals are for soil depth (cm), with O
808 referring to the superficial organic layer. ^cWHC: water-holding capacity (water content
809 by weight). ^eSOC: soil organic carbon (Walkley–Black). Values are mean±standard
810 deviation.

811 **Table 2.** Means ± standard deviations of forest structure metrics in control (C) and
812 thinned (T) plots for both sites (CAL: Calderona; HU: La Hunde). BA: basal area; D_B:
813 basal diameter; D_{BH}: diameter at breast height; LAI: leaf area index.

814 **Table 3.** Characteristics of the studied rainfall events and meteorological conditions
815 during rainfall in both sites (Calderona, CAL and La Hunde, HU) first classified by
816 convectiveness ($\beta_{L,10}$; $\beta=0$ for non-convective and $\beta\neq 0$ for convective events), and second
817 by those causing either drainage, P(D); generalized drainage, P(Dgen); or runoff P(RO).

818 Pg: gross rainfall, P_D: duration of gross rainfall; P_I: mean rainfall intensity, P_{I_{mx}}:
819 maximum rainfall intensity (10 min); P_{Gap}: time frequency without rainfall in an event, T:
820 air temperature, VPD: water vapor deficit, U_{mx}: maximum wind speed, U_{av}: mean wind
821 speed. Values reports are means; medians (left) and standard deviations (right) in
822 parenthesis (omitted in some variables for simplicity).

823 **Table 4.** Cross-validation correlation coefficients obtained in the BRT models fitted for
824 soil water replenishment ($\Delta\theta$), runoff (RO) and drainage (D, Dgen for generalized
825 drainage) in both sites separately and together (overall). The analyses were performed
826 either on the depth- or the rate-expressed variables. The average and range of the standard
827 error of the coefficients presented are 0.032 and 0.01-0.08 respectively.

828

829 FIGURE CAPTIONS

830 **Figure 1.** Pictures showing the thinned plots in the studied sites (a, CAL; b, HU) and the
831 trenches for collecting the surface and sub subsurface flows (c, CAL; d, HU). A location
832 map of the studied sites is showed in del Campo et al. (2018).

833 **Figure 2.** Three-dimensional representation of the resistivity values (Ohm.m) obtained in
834 the electrical tomography survey in CAL (top) and HU (bottom) sites. Blueish colors
835 correspond to soil (B horizons), green to yellowish to soil with high rock fraction and C
836 horizon and red to purplish to the rock substrate. Units in axes are meters.

837 **Figure 3.** Relative importance (RI, %) obtained in the BRT models for the four groups of
838 independent variables (antecedent soil moisture content, θ_{st_rel} ; forest structure, FOR_Str;
839 meteorological conditions during rainfall, MET; and rainfall characteristics, RAIN)
840 affecting soil water replenishment ($\Delta\theta$), runoff (RO), drainage (D, drainage) and
841 generalized drainage (Dgen), in both sites separately and together (overall). The
842 dependent variables were expressed in mm (left) or as % of gross rainfall (right).

843 **Figure 4.** Mean and median values in CAL and HU for the antecedent and maximum soil
844 moisture ($\theta_{\text{mx_rel}}$ and $\theta_{\text{st_rel}}$, % of range), soil water replenishment ($\Delta\theta$, mm and % of P_g).
845 C: control, T: thinned plot. Error bars indicate standard deviation of the means. Different
846 letters indicate significant differences between the plots for a site (ANCOVA results are
847 presented in Table 2S as supplemental material).

848 **Figure 5.** Mean and median values in CAL and HU for drainage (D, mm and % of P_g),
849 generalized drainage (Dgen, mm and % of P_g) and runoff (RO, mm and % of P_g). C:
850 control, T: thinned plot. Error bars indicate standard deviation of the means. Different
851 letters indicate significant differences between the plots in a site (ANCOVA results are
852 presented in Table 2S as supplemental material).

853 **Figure 6.** Temporal dynamics (mm) of runoff (RO), drainage (D) and generalized
854 drainage (Dgen) in the CAL plots. P_g is also represented and those events producing
855 response in each variable are highlighted in red color. The background shaded area
856 indicates the period when the piezometers were monitored, whereas the dark grey bars
857 indicate positive response, with their breadth indicating the period corresponding to a
858 single observation and their height the proportion of piezometers in which the response
859 was detected (% , 4 piezometers for each experimental plot). T: thinned (left); C: control
860 (right).

861 **Figure 7.** Temporal dynamics (mm) of runoff (RO), drainage (D) and generalized
862 drainage (Dgen) in the HU plots. P_g is also represented and those events producing
863 response in each variable are highlighted in red color. The background shaded area
864 indicates the period when the piezometers were monitored, whereas the dark grey bars
865 indicate positive response. Height of these dark grey bars indicate the proportion of
866 piezometers in which water was detected (% , 4 piezometers for each experimental plot).
867 Bailer malfunctioning made that only automatic readings during the last year of

868 experiment were available for comparisons. Figure 4S in supplemental material shows
 869 examples of automatic readings for two rainfall events. T: thinned (left); C: control
 870 (right).

871 **Figure 8.** Cumulative treatment impacts according to site (CAL, HU) on response
 872 variables (soil moisture replenishment, $\Delta\theta$; drainage, D; and generalized drainage, Dgen)
 873 as a shift in the daily ratio of thinned/control, L (T/C), following the intervention (baseline
 874 before thinning treatment is assumed to be zero). Gross rainfall (Pg) is also represented

875 **Figure 9.** Annual water balance (mm) according to site (CAL, HU) and treatment (C, T).
 876 Net rainfall (Pn_Thr+Stf) was obtained from del Campo et al. (2018) as the sum of
 877 throughfall and stemflow per event. Pg represents gross rainfall, $\Delta\theta$ represents soil
 878 moisture replenishment and D represents drainage.

879 **Figure 10.** Elasticity analyses comparing the median values for the experimental plots
 880 in the two studied sites (CAL and HU). The ratio $[Dgen+RO]/[Pg-Dgen-RO]$ was
 881 considered as a proxy to the blue to green water ratio (b/g), Dgen represents generalized
 882 drainage and θ_{mx} is the maximum soil moisture.

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Table 1.

Site characteristics	Units	CAL	HU
Slope	%	27.8±8.5	32.0±6.0
Aspect	°	311±10 (NW)	319±21 (NW)
P	mm	342	466
T	°C	14.0	12.8
PET	mm	837	749
Aboveground Biomass (Carbon)	Mg ha ⁻¹	47.3 (22.2)	49.7 (23.1)
Texture ^a	%	42.1±7.4;31.4±6.3;26.4±4.0	49.7±6.6;29.3±5.5;20.7±2.1
Stoniness ^b	% vol.	O: 55±25 0-10: 34±21 10-30: 20±19 >30: 5±5	O: 53±8 0-10: 64±8 10-30: 66±17 30-40: 56±5
WHC ^c	%	73±2	122±15
Ks	mm h ⁻¹	461	785

SOC^e	g Kg ⁻¹	28.5±22.1	103.5±32.4
pH	(water)	8.3±0.2	7.9±0.2
Carbonates	(g g ⁻¹ dry soil)	0.34±0.15	0.24±0.10

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887

888 **Table 2.**

Plot	D_B (cm)	D_{BH} (cm)	BA (m ² ha ⁻¹)	Density (tree ha ⁻¹)	Cover (%)	LAI (m ² m ⁻²)
CAL-C	4.60±3.65	2.92±2.52	18.6±4.6	12280±12202	78.7±14.0	1.5±0.2
CAL-T	13.18±4.2	8.51±3.12	4.5±0.8	703±232	38.7±11.5	0.5±0.1
HU-C	11.87±0.25	8.5±0.5	8.2±3.8	1112±308	62.7±2.3	1.1±0.2
HU-T	16.7±2.2	12.9±1.8	5.0±1.2	306±34	39.3±4.2	0.6±0.1

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891 **Table 3.**

Gross Rainfall types	N	P_g (mm)	P_D (min)	P_I (mm h ⁻¹)	P_Imx (mm h ⁻¹ , Δt=10 min)	P_{Gap} (% time)	β_{L,10}	T (°C)	VPD (Pa)	U_{mx} (m s ⁻¹)	U_{av} (m s ⁻¹)
CALDERONA (CAL)											
P_g (non-convective)	106	4.0 (2.6;3.9)	128 (91;118)	2.9 (2.1;3.2)	5.9 (4.8;4.7)	45	0	9.6	138	6.6	2.3
P_g (convective)	10	26.2 (19.7;24.4)	171 (95;190)	12.4 (7.4;8.6)	40.4 (40.8;11.5)	20	0.50	13.3	247	10.9	3.7
P_D	41	11.6 (7.2;15.1)	182 (125;165)	5.1 (3.0;6.0)	16.6 (10.4;15.7)	39	0.121	9.7	112	7.8	2.5
P_{Dgen}	32	12.9 (7.7;16.7)	184 (134;160)	5.2 (3.7;5.9)	18.2 (12;16.1)	41	0.127	9.9	106	7.3	2.2
P_{RO}	25	16.3 (11.6;17.6)	236 (178;184)	5.7 (3.7;6.5)	20.8 (16.8;16.6)	38	0.149	9.2	74	8.4	2.6
LA HUNDE (HU)											
P_g (non-convective)	226	5.0 (2.8;6.2)	240 (150;303)	1.7 (1.2;1.7)	4.1 (2.9;3.7)	40	0	8.1	127	3.5	0.7
P_g (convective)	10	13.4 (12.4;8.0)	98 (83;73)	11.3 (9.8;7.0)	34.3 (31.5;11.1)	40	0.61	16.2	186	3.4	0.5
P_D	146	6.5 (3.6;7.2)	280 (180;347)	2.1 (1.3;2.9)	5.7 (3.2;7.9)	23	0.030	7.4	74	4.3	0.9
P_{Dgen}	118	7.4 (4.0;7.7)	321 (205;374)	2 (1.3;2.8)	5.7 (3.6;7.9)	27	0.022	7.2	66	4.3	0.9
P_{RO}	149	6.6 (4.0;7.0)	264 (170;343)	2.5 (1.4;3.4)	6.6 (3.6;8.8)	21	0.041	8.7	106	4.2	0.8

892

893 **Table 4.**

Hydrological Variable	CAL	HU	OVERALL
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Depth (mm)	$\Delta\theta$	0.873	0.823	0.877
	D	0.956	0.929	0.923
	Dgen	0.918	0.853	0.875
	RO	0.921	0.572	0.864
Rate (%Pg)	$\Delta\theta$	0.783	0.726	0.765
	D	0.902	0.805	0.833
	D gen	0.820	0.700	0.734
	RO	0.800	0.577	0.734

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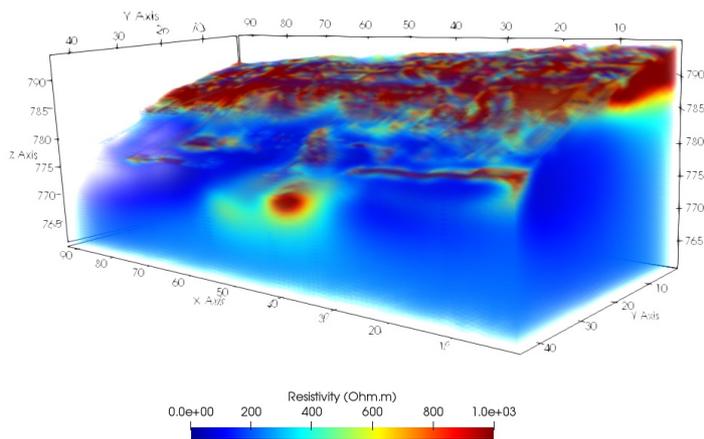
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895 **Figure 1**

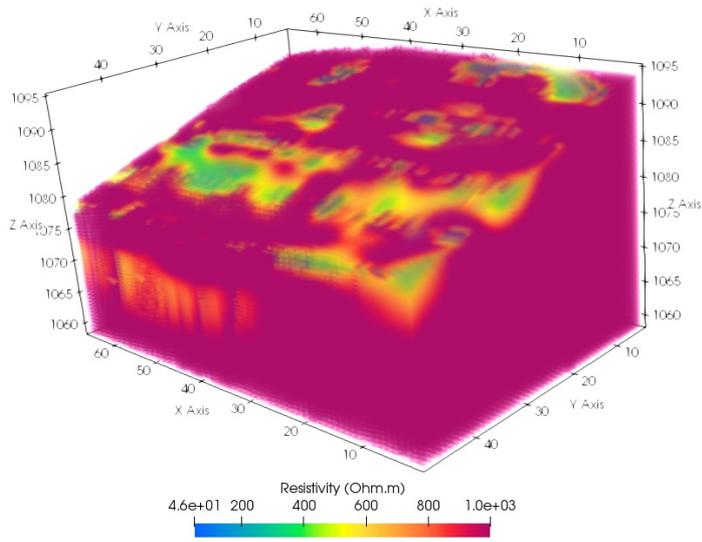


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897 **Figure 2**



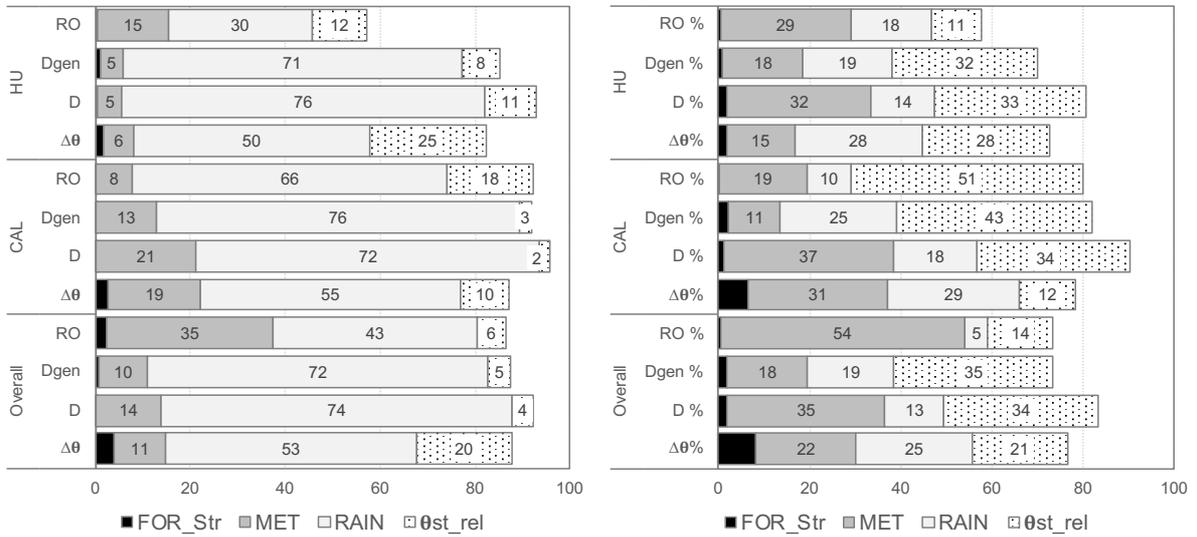
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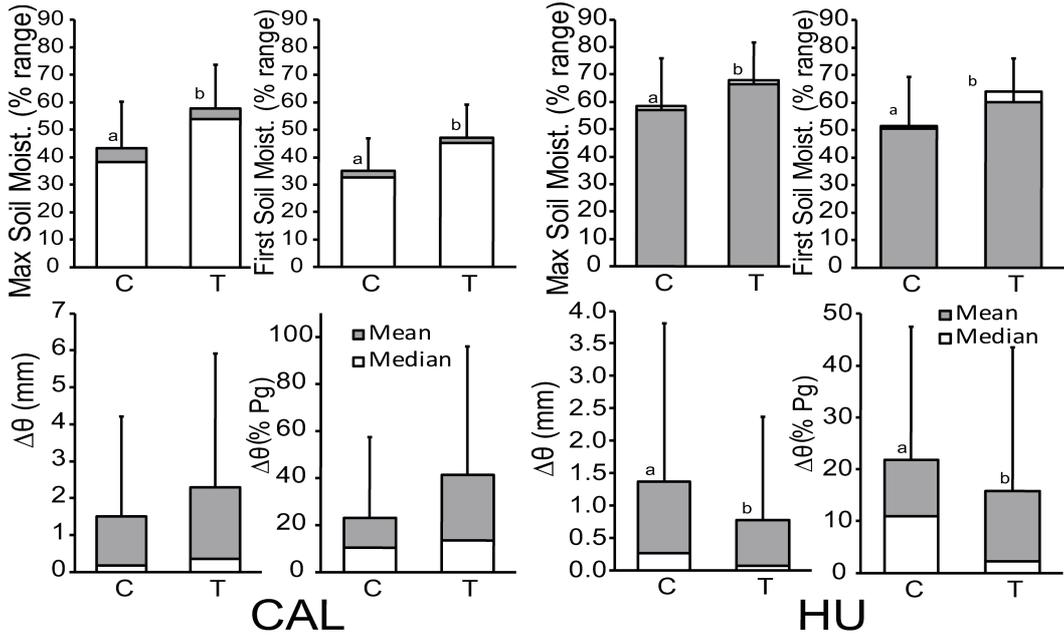
900 **Figure 3**

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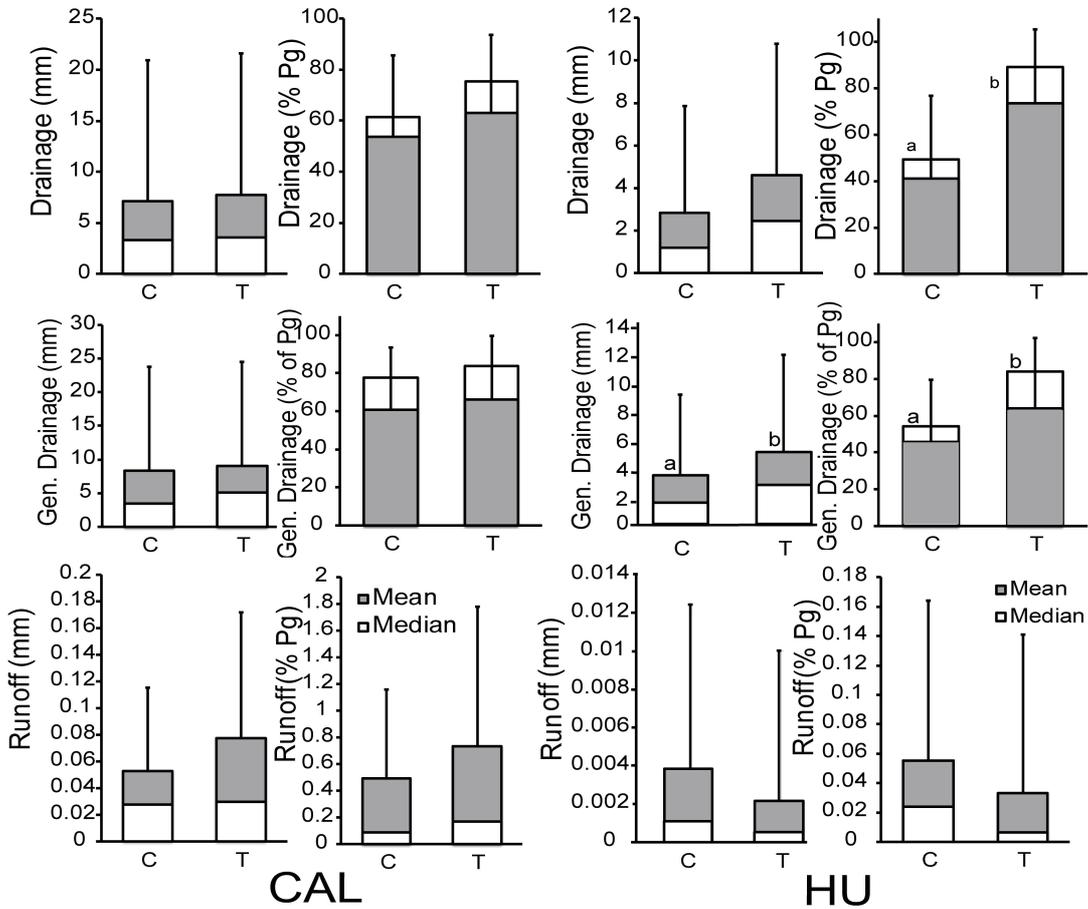
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903 **Figure 4**



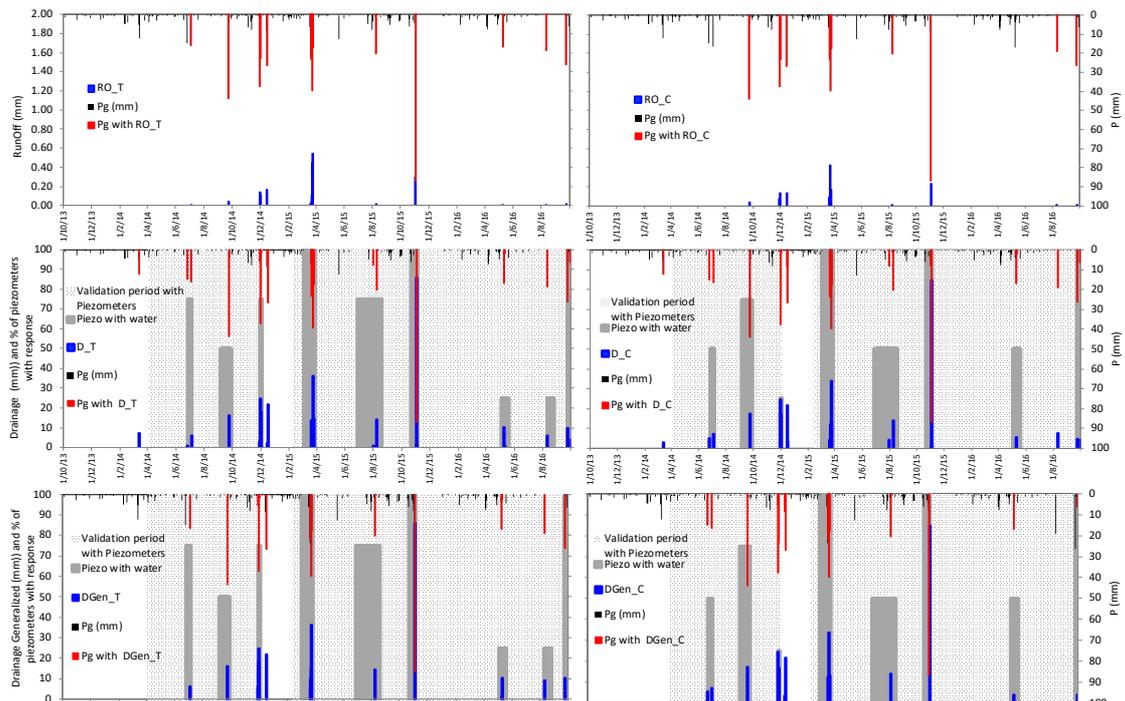
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905 **Figure 5**



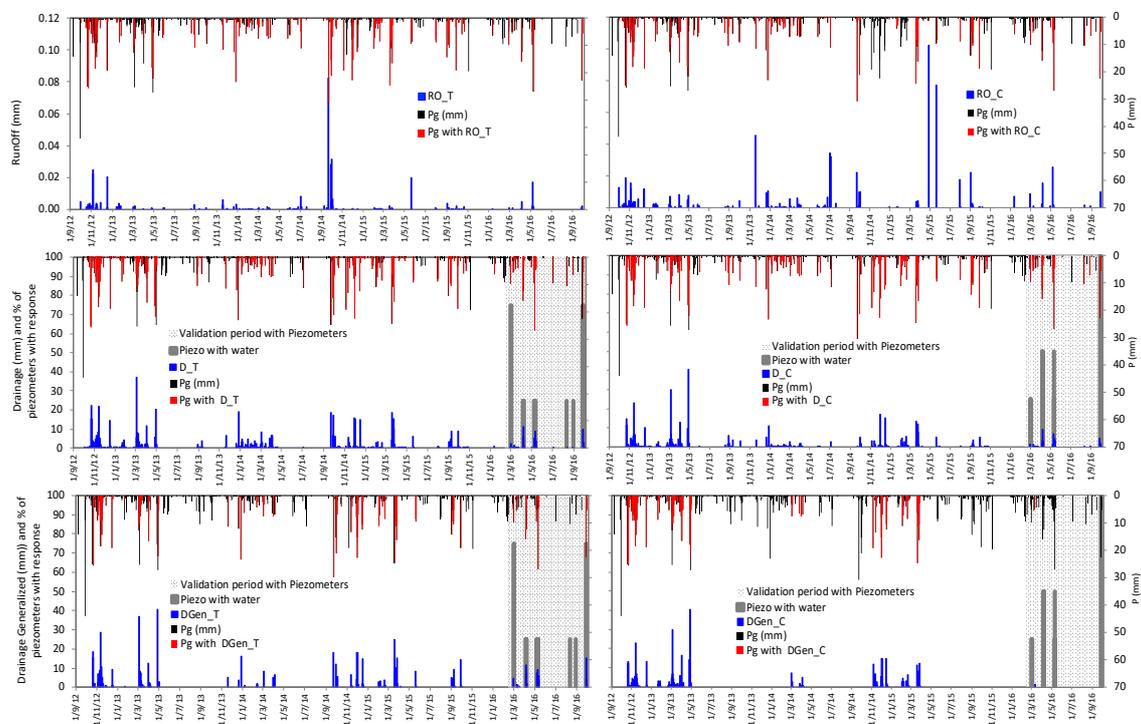
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907 **Figure 6**



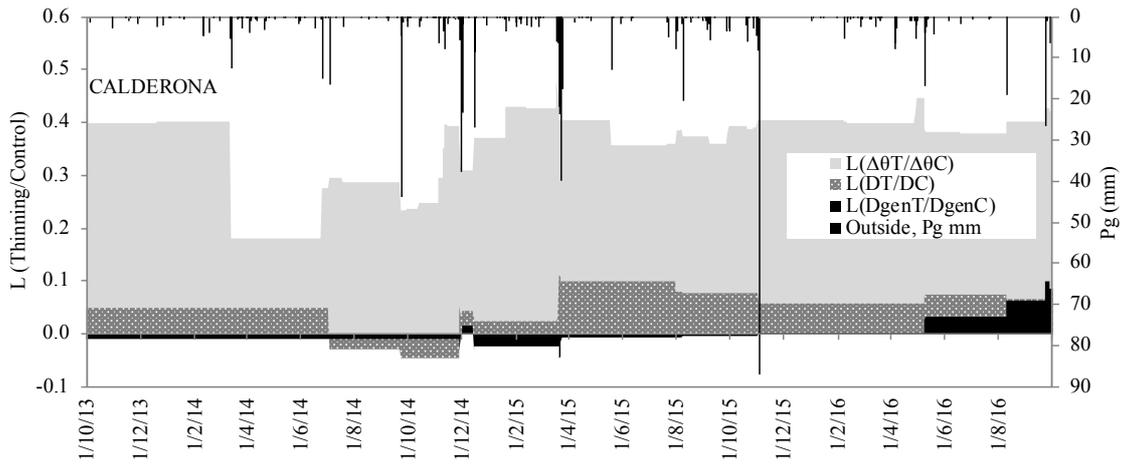
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909 **Figure 7**

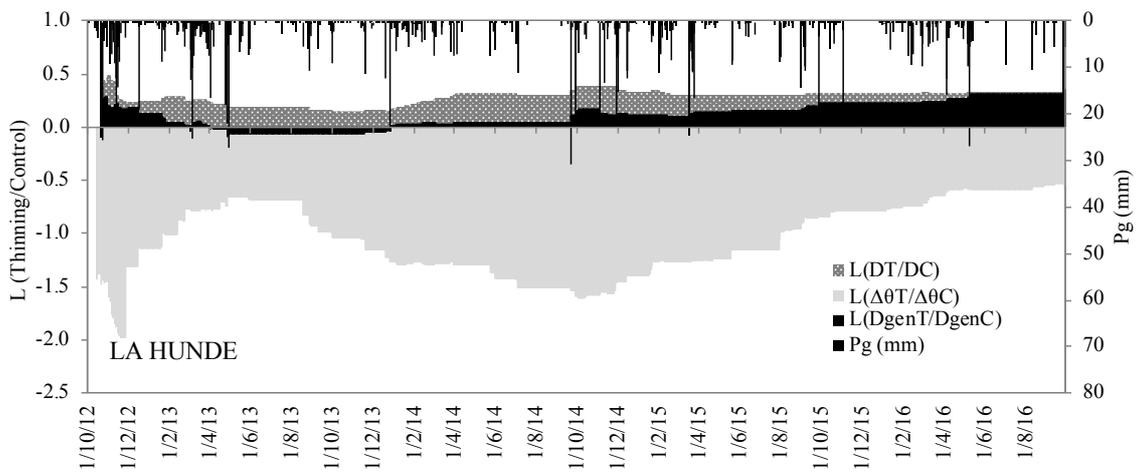


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911 **Figure 8**



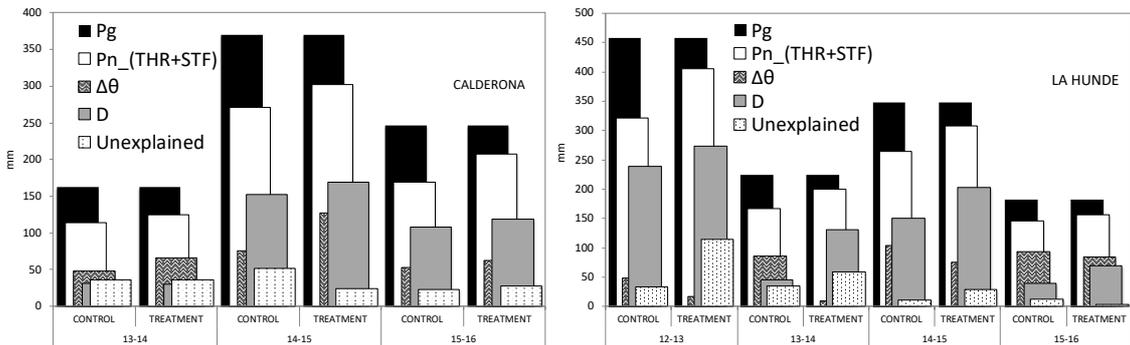
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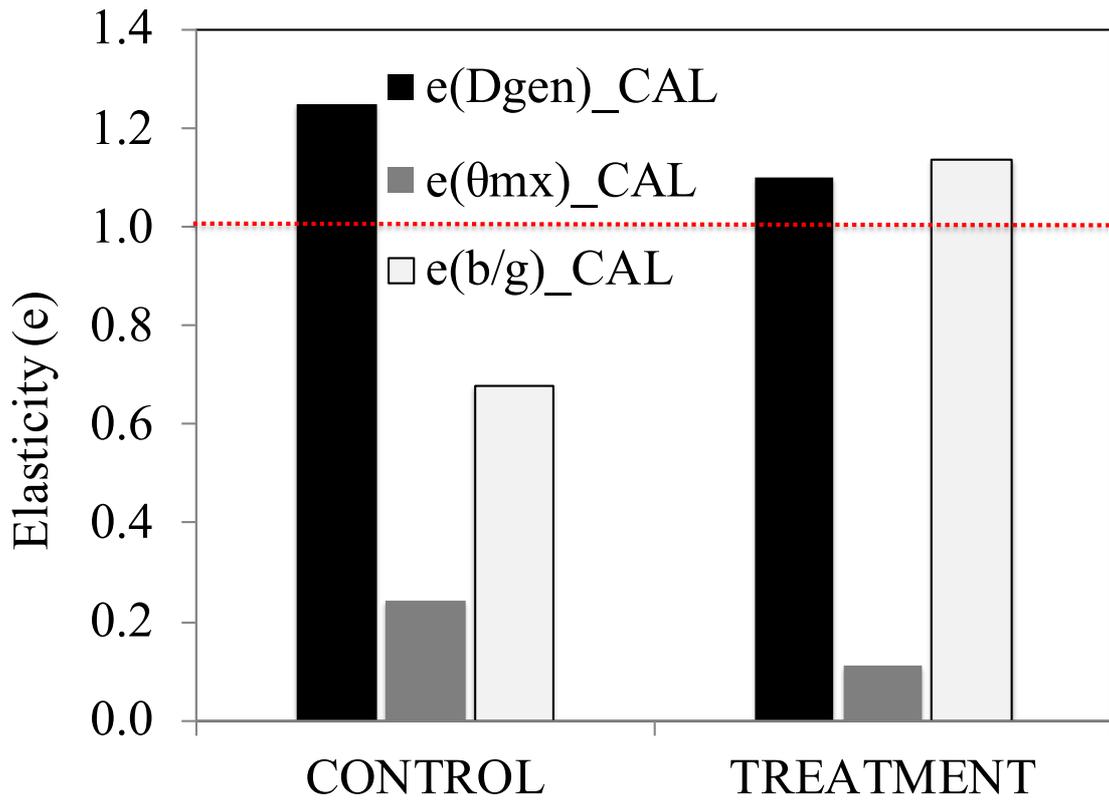
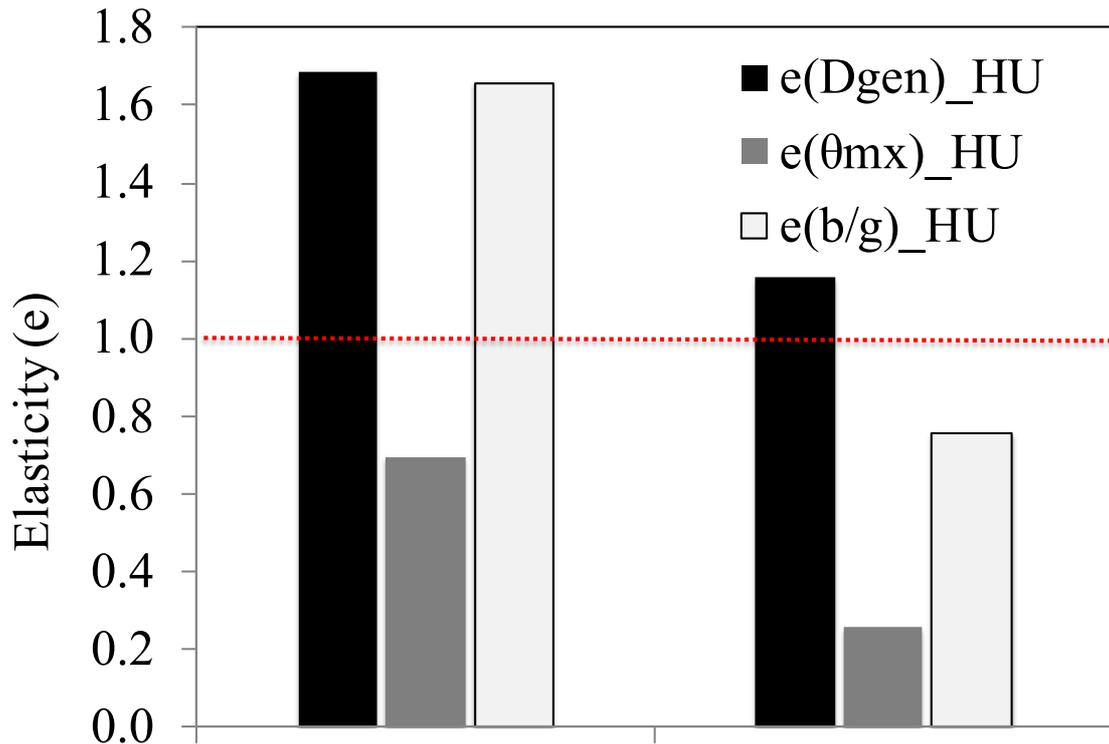
914 **Figure 9**

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917 **Figure 10**



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