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# Assessment of atmospheric and soil water stress impact on a tropical crop: the case of *Theobroma cacao* under Harmattan conditions in eastern Ghana.

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#### Abstract

In West Africa, Harmattan-induced atmospheric and soil droughts represent seasonally recurring hazards for *Theobroma cacao* L. agro-ecosystems. Under the influence of the Harmattan winds, precipitation is impaired and air humidity and temperature reach stressful levels. Climate change is causing an increase in temperature that will drive up the evaporative power of the atmosphere, risking to harshen both the soil and atmospheric stress. This would further threaten the viability of cacao cultivation in this region. To characterize the response of cacao trees to atmospheric and soil drought, we monitored two sub-plots, with and without irrigation, throughout one Harmattan season (November 2019 - March 2020) in the Eastern region in Ghana. For both treatments we recorded: sap flow velocity, photosynthetic active radiation (PAR) above and below the canopy, soil moisture, temperature, air humidity and daily precipitation. Leaf area index (LAI) was estimated from PAR measurements. To characterize drought responses of mature cocoa trees during the day and at the seasonal scale, we developed two boosted regression trees models (BRT)

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with the environmental variables measured. The atmospheric component of Harmattan-induced drought was found to affect the canopy to a similar extent as soil water stress, both causing a decline in LAI of 33%. This study confirmed the importance of soil drought but highlighted as well the crucial role of atmospheric drought for this species' transpiration control. Soil and atmospheric water stresses did not have a synergistic effect on transpiration under the studied conditions. The BRT models identified LAI as one of the most influential drivers for sap velocity, which, in turn was sensitive to the interactive effect of both atmospheric and soil drought. Our results highlight that not only reduced precipitation but also increasing atmospheric drought is likely to negatively impact on cacao production in West Africa under increasingly dry conditions imposed by the influence of the Harmattan winds.

*Keywords:* cacao, physiology, canopy, drought, sap velocity, boosted regression trees

#### 1 1. Introduction

The global climate has changed over the past century and is projected to 2 continue to change in the next decades at a higher pace than in the past [1]. 3 Global general circulation models (GCMs) agree that, except for an unlikely low emissions scenario, by the end of this century, global mean temperatures will rise by at least another 1.5°C and precipitation regimes over large areas worldwide will experience profound changes. In the tropical band (23°3'N-23°3'S), climate 7 change is expected to have a negative impact on agriculture, ultimately threat-8 ening the economic stability of countries that rely heavily on this sector. This is particularly true for West Africa where agricultural systems are among the most 10 vulnerable worldwide due to the economic constraints limiting access to agricul-11 tural technological advances, among other reasons [2, 3, 4]. West Africa's most 12 famous exported crop, cacao (Theobroma cacao L.), will experience a severe 13 reduction of its agro-pedo-climatic zone of cultivation [5, 6, 7]. Recent model 14 predictions based on the SRES-A2 greenhouse gas emission scenario [8] project continued rise in mean temperatures over the West African cocoa belt in the а 16 future and, while cumulative annual precipitations are not expected to change 17 significantly, their distribution over the course of the year is. The period with 18 no precipitations whatsoever may slightly shorten or remain unchanged in the 19 coming decades across West Africa but, due to the aforementioned increase in 20 temperatures, this region is expected to experience longer periods under greater 21 evaporative demand. Ultimately, this will result into increased frequency, sever-22 ity and duration of episodes of both soil and atmospheric water stress at the 23 plant level. Consequently, a large share of cocoa-producing regions in West 24 Africa will become unsuitable for production in the future [5, 6, 7], leaving mil-25 lions of smallholder farmers without a reliable source of income [9]. 26 The West African long dry season is characterized by the co-occurrence of low 27

precipitation and a dry wind, the Harmattan. The Harmattan is a north-easterly
trade wind blowing over North Africa that results from the continental-scale
pressure gradient between the subtropical subsidence zone and the Intertropical

Convergence Zone (ITCZ) [10]. During the West African dry season, corre-31 sponding to the boreal winter, the Harmattan advances to the southern part 32 of West Africa conveying a dry air mass from the Sahara to the south which 33 lingers around the northern edge of the cacao belt, along the Gulf of Guinea 34 [10]. The presence of the Harmattan hinders moist convection and suppresses 35 any chance of precipitations, only allowing for sporadic weak rains for hundreds 36 of kilometers south of the Intertropical Front, that marks the southern Harmat-37 tan extent at ground level [11]. The Harmattan further enhances soil and air 38 water stress as the evaporative demand increases due to higher wind speed and 39 reduced air humidity, ultimately leading to wide temperature differentials from 40 day to night. Hence, under the influence of the Harmattan, soil water content 41 decreases due to the lack of precipitation together with increased evaporative 42 demand at the leaf level. 43

Cacao is original from the Amazon basin, where water limitation is virtually in-44 existent [12, 13, 14]. The morphological traits of cacao are not adapted to deal 4 6 with water limitation: for example, cacao has large, broad leaves with minimal 46 waxing and high stomatal density that cause strong transpiration and evapo-47 ration rates under high irradiance or high vapor pressure deficit (VPD) [15]. 48 The hydraulic system of cacao is also poorly adapted to low water availability: 49 the main stem has wide xylem vessels [16] to pump water more efficiently from 50 the soil to the leaves, but this implies a greater risk of functionality loss due to 51 cavitation under water stress [17]. Moreover, the cacao root system is relatively 52 shallow, with high density of fine roots in the top 0.2-0.6 m of the soil, and hence 53 it cannot access deep water [18, 19]. The physiological performance of cacao is 54 also adapted to its native climatic conditions. Cacao optimum growth temper-55 ature is 24°C at night and 30°C during the day [20]. Cacao photosynthetic 56 efficiency starts declining at temperatures above 33°C, while night tempera-57 tures below 15.8°C suffice to observe a decline in photosynthesis and stomatal conductance [17]. Additionally, for cacao trees of the Amelonado family, genet-59 ically the most representative in West Africa, the reported base temperature 60 for vegetative growth is 19.7°C [21]. In West Africa, during the dry season and 61

under the influence of the Harmattan, air humidity is not sufficient to buffer 62 large daily thermal oscillations and air temperature can reach values as high as 63 44°C [22, 23] and as low as 12-14°C at night [24, 25]. These large daily temper-64 ature oscillations strongly inhibit the net growth and physiological performance 65 of cacao [20]. In addition, reduced air humidity due to the influence of the Har-66 mattan has a direct, negative effect on growth and physiological performance. 67 Indeed, in tropical environments it has been shown that 60% is the air humidity 68 threshold below which tree physiological performance starts to decrease and be-69 low 40%, in combination with high temperatures, it is considered that trees are 70 exposed to high atmospheric water stress [26, 27]. It can be assumed that such 71 conditions would be stressful for cacao as well, in line with [28]. Overall, it is 72 clear that cacao lacks high tolerance to drought or extreme temperatures. Thus, 73 the viability of cultivation of this crop outside its native range, in West Africa, 74 is severely threatened by future climate change. The future threats to cacao 75 cultivation are further exacerbated in full sun or lightly shaded monocoltural 76 systems, preferred by farmers across West Africa for the higher yields in the 77 short term but more exposed to atmospheric stress [23]. 78

Both soil and atmospheric drought impact negatively on plant growth and pro-79 ductivity [29, 30]. Plants first respond to increasing vapour pressure deficit 80 (VPD) by closing their pores on their leaf surfaces, the stomata, to reduce tran-81 spiration water loss, but this also entails a reduction in  $CO_2$  uptake to supply 82 photosynthesis and, eventually, reduced growth and production of reproductive structures [31, 32]. On the other hand, soil drought reduces the conductivity 84 of the soil-plant-atmosphere continuum, ultimately inducing stomatal closure 85 to protect the hydraulic system of the plant from embolism [33, 34]. Beyond 86 stomatal closure, high VPD also increases non-stomatal water losses, for exam-87 ple through the cuticle, further increasing risk of hydraulic failure [14, 16, 35]. 88 Besides reducing photosynthesis, soil drought also reduces xylem and phloem 89 transport and, hence, export of carbohydrates from the leaves to reproductive 90 organs for flower and fruit production [36, 30]. 91

<sup>92</sup> Several authors have already highlighted the need to better investigate the water

relations of cacao under field conditions [37], but we still lack a detailed char-93 acterization of how drought stress influences cacao physiology and reproduction 94 [17]. The effects of relative humidity [28] and temperature have been addressed 95 [20, 38, 21] and a few field trials have addressed soil drought stress but, to our 96 knowledge, no previous study has assessed the simultaneous effect of soil and 97 atmospheric drought, and their interaction [18, 19, 23]. The reduction of tran-98 spiration in response to soil water stress has been characterized in Brazil [13] 99 and Indonesia [18, 19], but in these locations VPD and air temperature rarely 100 reach stressful conditions for the trees. Such effects have not been addressed in 101 the West African cacao belt, where radically different atmospheric conditions 102 due to the influence of the Harmattan will likely impose drought stress levels 103 beyond those previously studied. The objective of this study is to clarify the 104 effects of atmospheric and soil water stress on cacao tree transpiration during 105 the dry Harmattan season and shed light on the drivers of transpiration under 106 such circumstances. Our hypothesis is that cacao trees will respond strongly to 107 both types of stresses but we expect VPD to be more influential, due to the ex-108 tremely low relative humidities and large temperature oscillations experienced 109 under the influence of the Harmattan. 110

#### 111 2. Materials and methods

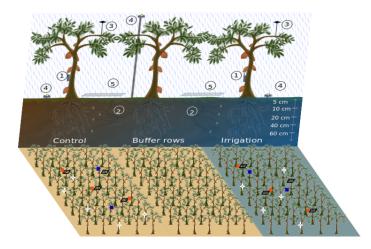


Figure 1: Schematic representation of the field experimental plan with an overview of the measurements taken. The experiment compared an irrigated plot (5 rows of 8 trees) to a control rain-fed one. The experimental plots were separated by a buffer of ten rows of trees. In each plot we measured 1) sap velocity (crosses), 2) soil moisture (squares), 3) temperature and air relative humidity (dots), 4) photosynthetically active radiation (PAR) above and below the canopy (dots) and 5) litterfall production (hatched parallelograms). The position of the sensors reflects their real position in the field experiment. Graphics by Pietro Della Sala, drawing of the cacao tree by Estelle Ribeyre.

#### 112 2.1. Study site and experimental design

Two plots of cacao trees, with and without irrigation, were monitored throughout the duration of the experiment  $(3^{rd}$  December 2019 to  $16^{th}$  March 2020) and their response to climate tracked closely by means of various sensors. The measured environmental variables were: soil volumetric water content at four depths from 10 to 60 cm, air temperature, air relative humidity and photosynthetically active radiation (Fig. 1).

The study was conducted at the experimental station of the Cacao Research Institute of Ghana (CRIG) located in New Tafo Akyem, Eastern Region, Ghana (6°13'53.7"N; 0°21'01.6"W; 203 m a.s.l.). At this location, the climate is warm and humid all year round except for two dry seasons. The main dry season has its core between December and February, but the onset can be as early as mid
November and lasts until sometime in March. The second dry season occurs
between the second half of July and the beginning of September and is typically
much less severe than the main dry season.

Throughout the year, temperature in Tafo oscillates between a monthly average minimum of 20 to 22°C and a monthly average maximum of 29 to 33°C [39]. Annual rainfall ranges between 1150 and 1800 mm with a mean value of 1565 mm per year [39, 40].

At the study site, a 2 ha cacao plantation was established in June 2013. Cacao 131 trees (Theobroma cacao L.), of homogeneous genetic origin, were planted with 132 a 2.5 x 2.5 m spacing (1600 trees  $ha^{-1}$ ), underneath *Gliricidia sepium* Jacq. 133 previously planted at a density of 10 trees  $ha^{-1}$ . The plantation consisted of 134 four blocks with 10 plots each. Each plot contained 40 trees planted in five rows 13 with eight trees per row. In November 2019, two plots of 40 hybrid Amelonado 136 trees with mean canopy height of 3 meters were selected for the study. The se-137 lected plots were located at least 20 m away from the nearest shading tree and, 138 therefore, were considered as a "typical West African full sun system". The two 139 study plots were separated by ten rows of cacao trees to avoid any edge effect or 140 interaction between the two (Fig. 1). From the  $26^{th}$  of November 2019 to the 141  $16^{th}$  of March 2020, trees in one plot (irrigation treatment) were irrigated with 142 60 L per tree (equivalent to approximately 9.6 mm per tree) on alternating dates 143 using a hose, whereas trees in the second plot (control) only received ambient 144 precipitation. The irrigation was evenly applied within 50 cm from the trunk, 145 where most of the roots were believed to be distributed [37]. Irrigation close to 146 the trunk was not reduced by losses due to canopy interception, thus its efficacy 147 was higher than a rainfall of 9.6 mm. 148

#### 149 2.2. Soil properties

According to a soil analysis of the experimental site conducted in March 2020, the upper soil (0-15 cm) was a eutric fluvisol with sandy-loam texture and below 15 cm of depth the soil texture was sandy clay loam. The pH was 6.5

across the entire profile. The upper soil was poor: organic matter content was 153 1.36%, magnesium was 2.06  $me * 100g^{-1}$ , total nitrogen was 0.15%, ammonium 154 was 14.2 ppm, phosphorus was 12.96 ppm, potassium was  $0.049 \ me * 100g^{-1}$ 155 and exchangeable calcium was 3.72  $me * 100g^{-1}$ . Based on the soil texture at 156 a depth of 30 cm, the bulk density (BD), field capacity (FC) and permanent 157 wilting point (PWP) were estimated in 1.45  $q * cm^{-3}$ , 34% and 13%, respectively. 158 Below 30 cm, the values of BD, FC and PWP were 1.55  $g * cm^{-3}$ , 35 and 16%, 159 respectively. 160

#### 161 2.3. Transpiration measurements

For the entire duration of the experiment, tree transpiration was estimated 162 from measurements of sap velocity using the Heat Ratio Method (HRM: [41]). 163 Five trees in each plot were selected, with mean±sd diameters (measured 20 164 cm below the main branch) of  $11.31\pm1.84$  and  $12.23\pm1.34$  cm in the control 165 and irrigated plots, respectively (Fig. 1, (1)). Within each plot, the trees were 166 selected based on a visual scoring system of overall condition (canopy density, 167 leaf greenness, number and diameter of jorquette branches etc.) and avoiding 168 spatial clustering of monitored trees. In November 2019, we installed one heat 169 probe sensor (SF-3, East30sensors, USA) on each selected tree. Each sensor 170 consisted of three probes, 35 mm in length, 1.3 mm in diameter, and 6 mm 171 axial distance apart. The central probe contained an evanohm heater, and the 172 lateral two probes, one upstream and one downstream with respect to the heater, 173 contained 3 thermistors placed at 5 mm, 17.5 mm, and 30 mm from the sensor 174 tip to monitor sap flow across the entire depth of the sapwood. The thermistor 175 temperature sensor consisted of a 10K precision resistor and a 10K thermistor 176 wired through a three wire half bridge connected to a datalogger per irrigation 177 treatment (CR800, Campbell Scientific, Logan, UT, USA). The accuracy of the 178 thermistors was  $\pm 0.2$  °C, and the resolution was 0.001 °C. The central needle 179 was heated by a 12V pulse of 3 seconds powered by the datalogger and reduced 180 to 5V through a heat control board (East30sensors, USA). 181

182 Sensors were installed on the trunk following the xylem direction, at a minimum

distance of 20 cm from any node or branching and at a minimum height from 183 the soil of 50 cm. Bark thickness was  $0.3\pm0.15$  cm (n = 10 trees) and was kept 184 in place to protect the wound from dryness and fungal attacks. The probes were 185 programmed in accordance with the Dual Method Approach (DMA: [42]). The 186 DMA combines the traditional approach to calculate sap velocity from heat 187 pulse velocity [41] with the Tmax Method [43]. The latter allows to capture 188 high and low flow rates both upward and downward along the stem [42]. Three 189 values of heat pulse velocity were recorded every 30 minutes per sensor. Each 190 value was calculated from the temperature difference between each pair of up-191 and down-flow thermistors located at three depths within the sapwood (5, 17.5)192 and 30 mm from the bark). Sap velocity was calculated for each of the three 193 positions and then upscaled to an integrated value of sap flux for each tree 194 through a weighted sum based on the sapwood area, estimated through wood 195 coring, associated to the specific radial position [42]. Before upscaling to a single 196 value per tree, each couple of thermocouples was calibrated to have the zero for 197 sap velocity when night potential evapotranspiration was zero (S. 2.6) [44, 45]. 198 In this study it was assumed that during the three months of dry season the 199 sapwood area increment is negligible, therefore, changes in transpiration are 200 approximated to variations in sap flux. 201

## 202 2.4. Soil VWC

Soil volumetric water content (VWC) was monitored in two flat locations per 203 plot, equidistant (170 cm) to all surrounding trees, at four depths (10, 20, 40, 204 and 60 cm), with TEROS 10 capacitance probes (METER group, Pullman, WA, 205 USA) (Fig. 1, (2)). The chosen distance allowed to capture the average VWC 206 of both the irrigated and the non-irrigated plots, avoiding potential biases due 207 to uneven irrigation in the former. Despite their distance from the trees, the 208 locations for VWC monitoring were shaded by the plots' closed canopy, effec-209 tively limiting quick evaporation of water after a watering event, be it rainfall 210 or irrigation. The TEROS 10 sensors estimate soil VWC from measurements of 211 the apparent dielectric permittivity in the 430 mL of surrounding medium. For 212

this study, the manufacturer's calibration for a generic mineral soil was used. 213 A value of soil VWC per probe was recorded on a CR800 datalogger every 30 214 minutes from November  $26^{th}$  2019 to March  $16^{th}$  2020. In one instance, the soil 21 VWC was transformed into water potential to compare our results with other 216 reported results. To obtain the pedotransfer function it was adopted the model 217 by Van Genuchten [46]. The parametrization of the model was done with the 218 R package soilphysics 4.0 [47] and the soil properties measured with the soil 219 analysis. 220

#### 221 2.5. Leaf area index

Leaf Area Index (LAI) was estimated from measurements of photosyntheti-222 cally active radiation (PAR) above (I) and below  $(I_0)$  the canopy (Fig. 1, (4)) 223 in an inverted form of Beer's law (Eq. (1)). Measurements of PAR were col-224 lected from November  $26^{th}$  2019 to March  $16^{th}$  2020, with a brief interruption 225 from December  $27^{th}$  2019 to January  $8^{th}$  2020. Incoming photosynthetic active 226 radiation (PAR;  $\mu mol * s^{-1} * m^{-2}$ ) above the canopy was measured every 30 227 minutes at one position in the irrigated plot. The PAR sensor (SQ110, Apogee 228 instruments, Santa Monica, CA, USA) was mounted on top of a 5 m levelled 229 iron pipe planted in the soil, i.e., 2 m above the canopy. Additionally, PAR was 230 monitored under the canopy at three locations in each plot with three sensors 231 mounted on levelled poles at 20 cm height. The extinction coefficient (K), nec-232 essary to calculate LAI from PAR data (Eq. (1)), depends on the solar angle ( $\phi$ ) 233 and a leaf angle distribution coefficient (x) and was calculated applying Eq. (2) 234 [48]. The parameter x was calculated as the ratio of vertical to horizontal pro-235 jections of the canopy [48]. Based on field measurements, x was evaluated at 1.2, 236 corresponding to an ellipsoid leaf angle distribution. A reliable estimation of K 237 is possible only when the solar angle is close to the zenith [49, 50], therefore, 238 we estimated LAI from PAR measurements collected between 10:30 and 14:30 239 (solar time). Furthermore, to avoid overestimation of LAI due to excess direct 240 radiation reaching the sensors under the canopy, data from each sensor were 241 smoothed and interpolated with the *Daniell* modified Fourier method [51, 52] 24 2

and the three resulting curves averaged to get a final LAI value for the entire plot.

$$\frac{I}{I_0} = e^{-K * LAI} \tag{1}$$

$$K = \frac{\sqrt{x^2 + 1/tan^2\phi}}{1.47 + 0.45x + 0.1223x^2 - 0.013x^3 + 0.000509x^4}$$
(2)

LAI change due to defoliation was also monitored through the same monitor-245 ing period with a set of four litter traps per treatment (Fig. 1, (5)). Each trap 246 consisted on a suspended fine net of 1.2  $m^2$  that was positioned in randomized 247 locations within each plot. The litter production was collected every 15 days 248 and oven-dried at 100 °C for 36 hours to obtain the dry weight. To convert the 24 9 leaf dry weight in LAI, in the beginning of the experiment the mean specific leaf 250 area (SLA  $cm^2 * g[dryleaf]^{-1}$ ) was calculated. A sample of 30 leaves (10 from 251 the lower, middle and top canopy) for each treatment plot was scanned on a 252 reference surface (an A4 sheet). The total area was estimated as the percentage 253 of the images that was not white with ImageJ 1.53a [53]. To obtain the SLA, 254 the average leaf area was divided by its oven-dry weight. 255

#### 256 2.6. Atmospheric conditions

Air temperature and relative humidity were logged hourly in each plot, with 257 with two iButtons (DS1923-F5: Hygrochron, iButtonLink LLC, Whitewater, 258 WI,USA) above the canopy at c.a. 5 m height. To protect the sensors from 259 direct radiation and precipitation, these were installed facing the ground, glued 260 to the internal part of bottle caps (Fig. 1, (3)). As temperature and relative 261 humidity were logged hourly and as they represent continuous variables, a linear 262 interpolation was applied in order to obtain half-hourly time series that matched 263 those of the other measured variables in the data set. The vapour pressure 264 deficit (VPD) and night-time potential evapotranspiration (required to calibrate 265 the zero point for sap velocity) were calculated following the FAO-56 Penman-266 Monteith method [54] by means of the Python package 'opencroplib = =0.1.5' 267 [55]. The calculation of the night-time potential evapotranspiration used a wind 268

velocity above the canopy of 0.77  $m * s^{-1}$  at 5 m height, in compliance with the average night-time value from a weather station at less than 1 km from the site. Daily precipitation data between October 1<sup>st</sup>, 2019 and March 16<sup>th</sup>, 2020 were retrieved from the Unified Gauge-Based Analysis of Daily Precipitation of the NOAA Climate Prediction Center (CPC; [40]).

#### 274 2.7. Statistical analysis

Prior to analysis, we checked for sensor glitches, numerical artefacts of the 275 sensor raw signal and measurement errors caused by faulty sensors. All mea-276 sured variables but sap velocity were treated as continuous, with the hypothesis 277 that they cannot abruptly change over half an hour. For this reason, it was 278 decided to study the evolution of their first derivative in time and consider as 279 outliers the points whose absolute value lied outside the two standard deviations 280 confidence interval. A graphical evaluation of the data points flagged as outliers 281 was carried out before they were discarded. 282

#### 283 2.7.1. GAMM analysis

A generalized additive mixed model (GAMM) was used to model and as-284 sess the differences between the irrigated and non-irrigated plot dynamics of 28 LAI over time. The GAM family of models was chosen primarily because LAI 286 was expected to exhibit a complex non-linear relationship with the environ-287 ment. Secondly, it was necessary to use a GAMM because LAI measurements 288 in time were not completely independent as they were taken by the same sen-289 sors. Lastly, the choice to use a GAMM was dictated by the fact that for each 290 date we had only few points for LAI, thus large variance. It would have been 291 difficult to appreciate the difference without the GAMM model capturing the 292 time evolution of LAI. The built GAMM model fitted a gaussian distribution 293 for LAI (continuous variable) using the treatment (irrigated or control) as fixed 294 effects and taking into account the random sensor-to-sensor variability. The 295 effect of time was fitted by a smooth term using Duchon splines, allowing the 296 the predictions to take into account the differential in water availability due to 297

irrigation. The GAMM-modeled LAI for both watering treatments was plotted
and we interpreted non-overlapping 95% confidence intervals as a significant
difference between treatment levels for a given period. All these analyses were
performed in the R environment v3.6.1 [56] using packages plyr [57], tidyverse
v1.3.0 [58], mgcv [59] and itsadug v2.3 [60].

#### 303 2.7.2. Boosted Regression Tree analysis

We used Boosted Regression Trees analysis (BRT) modelling to predict tran-304 spiration from climatic variables [61]. BRT uses two algorithms: regression tree 305 and boosting. Tree-based regression models, described for use in ecology by 306 [62], partition the solutions space with a set of rules, identifying the most ho-307 mogeneous regions in terms of response to predictors. They then fit a constant 308 to each region, fitting the average response in that region with the assumption 309 of normally distributed errors. With each iteration the tree grows by repeat-310 edly applying the analysis of the predictors space to its own output until a 311 user-defined stopping criterion is reached. Tree-based models are intuitive, easy 31 2 to visualize and are fairly insensitive to outliers, missing data and data types 31 3 (numeric, binary, categorical etc.) but they lack the accuracy of other methods, 314 such as GLM and GAMM. To compensate for this downside it is convenient 31 5 to combine tree-based models with boosting. The idea behind the boosting 316 methods is that it is more probable to find many rules of thumb, than to find 317 a single, highly accurate prediction rule [63]. It is, therefore, more convenient 31 8 to approximate the solution by averaging the results of a large number of rules 31 9 of thumb rather than aiming for a unique highly accurate one. The BRT uses 320 boosting as a way to evaluate the gradient of the predictors space by focusing 321 on the variation in the response not explained yet by the model at a given step 322 and fitting a new tree to its residuals [64]. Through boosting, decision trees are 323 fitted iteratively to the training data, increasingly emphasising the still poorly 324 modelled observations. As the boosting process is stage-wise, existing trees are 325 left unchanged as the model is enlarged but the fitted value is estimated at each 326 step to reflect the contribution of the newly added tree. The final BRT model 327

328 is a linear regression model where each term is a tree.

In order to ensure the stability of the models' results and avert over-fitting, the evaluation looked at the difference between the training data coefficient of correlation and the coefficient of correlation for the 100-fold cross-validation. The skill of the models, instead, was assessed by plotting the predicted values against the measured ones [61, 64].

The BRT analyses were carried out using R v3.6.1 [56] and the gbm [65] and dismo [66] packages. The parameterization of the two models can be found in Tab. C.3.

Two BRT models were built using 75% of the dataset to explain the relative 337 importance of the potential drivers of transpiration during the dry Harmattan 338 season. The remaining 25% was used to fit the models and evaluate them 339 against the measured values. The two BRT models considered soil volumetric 34 0 water content (VWC), photosynthetic active radiation above the canopy (PAR) 341 and the vapour pressure deficit of the atmosphere (VPD) as environmental 342 predictors, the leaf area index (LAI) of the two plots as indicator of the general 343 state of the canopy and the diameter of individual trees as a proxy for their 344 dimension. The first model (model 1) used the half-hourly daytime data (PAR 34 5 > 15  $\mu mol * s^{-1} * m^{-2}$ ) to investigate the importance of each aforementioned 34.6 predictors in determining the daily daytime cycle of sap velocity. The second 347 model (model 2) investigated the role of the same predictors at the time scale 348 of one day; for this, the input variables as well as sap velocity were averaged 34 over the period of the day with a PAR above 15  $\mu mol * s^{-1} * m^{-2}$ . 350

To avoid possible co-variations due to a common daily cycle, for the first model it was decided to remove the daily pattern from the vapor pressure deficit and radiation, maintaining only the effects due to the variation from the average daily cycle. The global daily pattern was maintained as a separate variable, i.e., the hour of the day (Hour), and included among the predictors.

The two models were based on the assumption of a normal distribution of the data (family = "Gaussian") and parameterized to avoid over-fitting (Tab. C.3) [61, 64].

#### 359 3. Results

#### 360 3.1. Climatic conditions

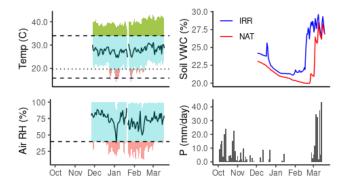


Figure 2: Climatic conditions during the study period. Left top: mean daily air temperature (solid black line), the thresholds for photosynthetic decline (34 and  $15.8^{\circ}$ C) and observed base temperature for vegetative growth (19.7°C) [17] (dashed lines) and the range of values (shaded area) are depicted. Left bottom: mean daily relative air humidity (solid black line) with the threshold of 40% (dashed line) and its range of observed values (shaded area). Right top: average soil VWC for the entire profile (10 to 60 cm) for the irrigated (blue line) and non irrigated (red line) plots. Right bottom: daily precipitation.

The average daily temperature during the experiment  $(26^{th}$  of November, 361 2019 -18<sup>th</sup> of March, 2020) was  $27.8 \pm 1.5$  °C, and the maximum and minimum 362 recorded temperatures were 43.1 °C and 13.6 °C respectively (Fig. 2). Between 363 January and February, the site experienced the hottest temperatures and the 364 largest daily thermal oscillations, surpassing both the upper and lower thresh-365 olds for maintaining photosynthesis for several hours. Temperatures above the 366 threshold at which photosynthesis declines (34 °C), were recorded throughout 367 the entire period under analysis on average for  $6.5 \pm 1.5$  hours a day (Fig. A.1). 368 Temperatures below the lower threshold for photosynthetic efficiency (19.7 °C), 369 instead, occurred only in January and February when the Harmattan winds 370 reached the site. Relative humidity fell below 40% for the first time in early De-371 cember and, from late December until March, the site experience several hours 372 with RH below 40% almost daily (Fig. A.1), reaching up to fifteen hours per 373

day below 40% in January. During our study period, the total precipitation 374 was 116 mm, of which 92 mm fell in March while the remaining 24 mm were 375 distributed in sporadic events from December to February. Prior to the onset of 376 our study, in October and November 2019, the site received 222 mm of precip-377 itation. The average soil VWC across the entire 10 - 60 cm profile was always 378 higher in the irrigated treatment than in the control throughout the experiment. 379 Soil VWC was above 21% in January and February in the irrigated plot, while 380 in the non-irrigated (control) plot, VWC continued to decline below 20% over 381 the same period. VWC quickly recovered in both treatments in March, when 382 rains resumed. 383

384

385 3.2. Leaf area index (LAI)

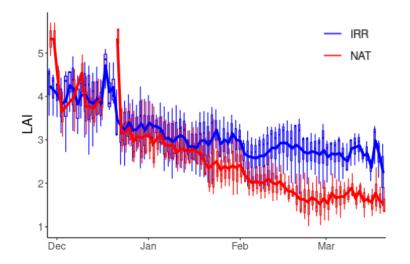


Figure 3: Leaf area index (LAI:  $m^2[leaf]*m^{-2}[soil]$ ) in the two watering treatments: irrigated (blue) and control (red), throughout the observed dry season. The lines represent the three day-moving averages of the LAI times series. Box-plots represent the variability between the three sensors in each plot for every date.

Overall, LAI declined steadily throughout the experiment both in the irrigated and in the control plot. Throughout December, the LAI remained constant and started to decrease in January, in both treatment plots, at the same

time as the number of days with RH < 40% started to increase (Fig. 3, Fig. A.1). 389 In February, LAI continued to decrease, faster in the non-irrigated (control) plot 390 than in the irrigated one. At the end of the study, the estimated LAI was 2.7 39:  $m^{2}[leaf] * m^{-2}[soil]$  at the irrigated plot and 1.5  $m^{2}[leaf] * m^{-2}[soil]$  at the 392 non-irrigated plot compared to an estimated 4  $m^2[leaf] * m^{-2}[soil]$  in the be-393 ginning of the study for both treatments (Fig. 3). The LAI, thus, dropped by 394 approximately 32.5% in the irrigated plot and by approximately 62.5% in the 395 control. According to the GAMM, from mid-February onwards, LAI in the ir-396 rigated plot was significantly higher than in the control plot (Fig. B.1). 397

The total litter collected throughout the entire experiment was greater in the 398 control (309  $q[dry] * m^{-2}$ ) than in the irrigated (247  $q[dry] * m^{-2}$ ) treatment. 399 Most of the shedding appears to have occurred in late December for both treat-400 ments, while in March, with the resumption of rain, the litterfall went to zero 401 (Fig. B.2). The average litterfall was lower in the irrigated plot, notably in the 402 first week of January (Fig. B.2), when air humidity dropped below 40% for the 403 first time. (Fig. 2). This is in agreement with the results of LAI dynamic that 404 evidenced a difference in LAI between plots at the end of the season resulting 405 from a steadily larger foliage loss in the non-irrigated plot (Fig. 3, Fig. B.2). 406 407

#### 408 3.3. Sub-daily patterns of sap velocity in response to climatic drivers

Figure 4 shows the response of sap velocity to VPD under different levels of 409 PAR above the canopy in the two treatments. At low VPD values ( $<1 \ kPa$ ) 410 trees in the control plot seem to have transpired more than those in the irrigated 411 plot. Under high VPD (> 4 kPa), measurements of sap flow velocity from trees 412 from both treatments presented a high dispersion regardless of the PAR level. 413 Under intermediate VPD (2-4 kPa), sap velocity appeared to respond more to 414 PAR in trees from the irrigated plot. Figure 5 shows the average daily cycle 41 ! of sap velocity during the 25% most and least stressing days over the study 416 period. Sap velocity at low VPD followed the same cycle in the two plots 417 without significant differences except in the early morning when the control 418

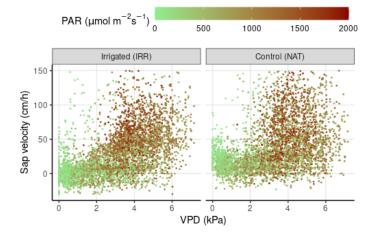


Figure 4: Sub-daily daytime measurements of cacao trees sap velocity in response to varying VPD: in the irrigated plot (left) and in the control plot (right) during the dry Harmattan season. Symbol colours depict PAR levels measured above the canopy (in  $\mu mol * s^{-1} * m^{-2}$ ).

trees transpired significantly more than the irrigated ones (Fig. 5 B). In days 419 when VPD was high the average sap velocity peaked around 70 cm/h for both 420 treatments, but the irrigated plot presented on average a sap velocity higher 421 than the control during the late mornings while the control tended to maintain 422 a higher sap velocity in the late afternoon (Fig. 5 A). In the morning the 423 control presented a peak in sap velocity around 6 a.m. (20 cm/h), regardless 424 of the level of VPD. An early morning peak in sap velocity was also found for 425 the irrigated plot in days when VPD was high. Both Figure 4 and 5 hinted to 426 some behaviours that were not always easy to grasp, thus the importance of the 427 study with the two BRT models. 428

The main drivers of half-hourly variations in sap velocity during the day were identified by means of a BRT model (model 1) (Fig. 6). The model averted overfitting and was considered to be stable (training data correlation = 0.885; crossvalidation correlation coefficient = 0.741). A regression analysis of predicted against measured values for the test data set showed that model 1, at low values (up to 30  $cm * h^{-1}$ ), slightly overestimated half-hourly sap velocity, otherwise it underestimated sap velocity (Fig. C.1). The "fitted function" in Figure 6

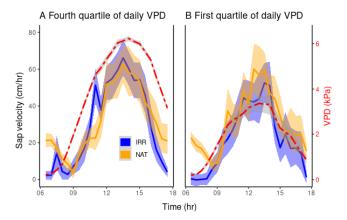


Figure 5: Sap velocity average daily cycle for cacao trees under high (A) and low (B) VPD conditions. The dashed line represents the average VPD cycle. The average cycles for sap velocity and VPD were obtained using the data from the 25% most and least stressing days of the dry Harmattan season 2019/2020. The shadings depict the the 95% confidence interval.

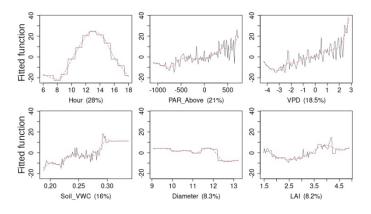


Figure 6: Overview of daytime sap velocity response to the main environmental drivers as identified with a BRT model (model 1). The responses are calculated on half-hourly data during the entire period of this study. Functional shapes of the response of sap velocity to the environmental variables: Hour, VPD, PAR, Soil VWC, LAI and tree diameter. PAR and VPD were expressed as variations from the respective daytime cycle. Each variable was presented with the relative weight of its variation in that of sap velocity within parentheses. The red dotted lines represent the LOESS smoothing for each response function.

represents the partial dependency of sap velocity to the different variables andit is obtained by averaging out the effects of all the variables in the model except

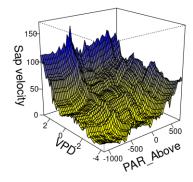
one, and then plot the average fitted value with respect to the one variable left. 438 Furthermore, it was chosen to plot a smoothing of the BRT functions using a 439 LOESS smoothing - the smoothed functions are shown in red (Fig. 6). Once 44 0 the effect of the daily cycle (Hour), which has a relative influence of 28.0% 441 in explaining sap velocity, was removed, the prevalent environmental variables 442 explaining the variability in sap velocity were PAR above the canopy (21.0%)443 and VPD (18.5%), followed by soil VWC (16.0%). The evolution of LAI over 444 the season in the two plots explained 8.2% of the variability, while tree diameter, 44 9 proxy for the different trees, accounted for 8.3% of the total variability. 44 <del>6</del>

Sap velocity responded linearly to the variations of VPD around its mean daily 447 cycle ( $\Delta$  VPD) up to 2 kPa. Beyond this point, the BRT model suggests 448 an increase in the slope of this relationship. The variability in PAR above 44 9 the canopy was a key driver of sap velocity. When PAR was above-average 450 compared to its mean daily cycle ( $\Delta PAR > 0$ ), sap velocity increased with PAR 451 in a linear fashion. For below-average PAR ( $\Delta$  PAR <0), sap velocity remained 452 relatively stable until  $\Delta$  PAR reached  $-500 \mu mol * s^{-1} * m^{-2}$ . Beyond this point, 453 the response of sap flow to PAR declined and then stabilized. Sap velocity 454 increased with soil VWC up to 0.24  $m^3 * m^{-3}$  and plateaued until it reached 455 a threshold value of 0.27  $m^3 * m^{-3}$ . Beyond this value, sap velocity increased 456 steeply until VWC reached a value of 0.29  $m^3 * m^{-3}$ . At a VWC above 0.29 457  $m^3 * m^{-3}$  the response of sap velocity flattened out until field capacity (0.34) 458  $m^3 * m^{-3}$ )(Figure 6). The LAI and tree diameter showed a rather flat relation 45 with sap velocity and were negligible drivers of sap velocity at the sub-daily 460 timescale. 461

In the sub-daily model, the interactions between VPD and PAR and VPD and LAI were more important than the other interactions (Tab. 1) which means that VPD modified the sap velocity response to both LAI and PAR significantly. When VPD and PAR were low compared to the average daily cycle, sap velocity was very low (Fig. 7). When VPD was high, the sap velocity was very high, for any value of PAR. Similarly, when PAR was high, sap velocity was high and stable whatever the value of VPD, except when VPD was very high, where

	Hour	Diameter	VPD	PAR_Above	LAI	Soil_VWC
Hour	0	15532.81	43578.24	40167	8701.57	13663.24
Diameter	0	0	23831.45	12599.22	10280.22	25915.23
VPD	0	0	0	82788.32	71754.54	59080.78
PAR_Above	0	0	0	0	35495.65	23574.91
LAI	0	0	0	0	0	27616.26
Soil_VWC	0	0	0	0	0	0

Table 1: Interactions between the variables considered in the BRT Model 1 in explaining cacao sap velocity at a sub-daily timescale. For each pair of variables, the table reports the mean value of the residuals, which represents the strength of the interaction. In bold: the interactions considered strong, thus significant, by the model.



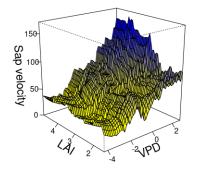


Figure 7: Three-dimensional partial dependence plots for the interaction between VPD  $(\Delta k Pa)$  and PAR  $(\Delta \mu mol * s^{-1} * m^{-2})$  variations from the mean daily cycle in the BRT model 1 for half-hourly sap velocity  $(cm * h^{-1})$  in cacao trees. All variables except those plotted are held constant at their mean values.

Figure 8: Three-dimensional partial dependence plots for the interaction between LAI  $(m^2[leaf] * m^{-2}[soil])$  and VPD ( $\Delta kPa$ ) variation from the mean daily cycle in the BRT model 1 for half-hourly sap velocity  $(cm * h^{-1})$  in cacao trees. All variables except those plotted are held constant at their mean values.

sap velocity was very high. When VPD and LAI were low, the sap flow was
also very low (Fig. 8). When VPD was low, the sap velocity remained low,
whatever the value of LAI. When LAI was low (<2.5), sap velocity increased</li>
with the VPD anomaly with the same evolution and reached similar high values
for positive VPD anomalies. But, when both LAI and VPD were high (LAI>2.5;
VPD anomaly >0), the response of sap velocity was characterized by a steeper

slope, indicating a possible positive synergistic effect on sap velocity between 475 VPD and LAI. This synergy meant that the positive response of sap velocity to 476 positive VPD anomalies (Fig. 6) was amplified by the canopy density when LAI 47 was above 2.5 (Fig. 8). If the LAI alone was not able to sensibly change the 478 sap velocity (Fig. 6), at high values (LAI>2.5) its importance was inflated by 479 positive VPD anomalies (Fig. 8). Ultimately, the sap velocity of trees with an 480 LAI above 2.5 under a stronger pulling force from the atmosphere (positive VPD 481 anomalies) resulted to be higher than the single responses to each of the two 482 predictors alone. Conversely, the potentially interesting interaction between 483 VPD and soil VWC was not found strong enough to be considered relevant 484 according to model 1 (Tab. 1). 485

#### 3.4. Daily variations of sap velocity in response to climatic drivers

The BRT model based on the daily averages of sap flow velocity rendered 487 a strong training data correlation (0.944), a good cross-validation correlation 488 (0.803). Thus, the model was found to be stable and reliable. An evaluation 489 of predicted against measured values on the test data showed that model 2 490 predicted well sap velocity despite a slight underestimation at high daily aver-491 age sap velocities (above 57  $cm * h^{-1}$ ) (Fig. C.2). Furthermore, when scaling 492 up from the half-hourly to the daily analysis, the weight of the predictors on 493 sap velocity changed (Fig. 9). The variability in daily sap velocity was mostly 494 explained by LAI (27.2%) and soil VWC (21.4%). The relative importance of 495 radiation (PAR) and VPD decreased to 18.1% and 17.0% respectively. 16.3% 496 of the variability in sap velocity was explained by variations in trunk diameter. 497 498

The shape of the response curves of daily sap velocity to the considered variables was different to that observed for half-hourly values (Fig. 6, 9). Sap velocity increased nearly linearly with LAI above  $3.5 m^2[leaf] * m^{-2}[soil]$ . Sap velocity was insensitive to LAI, for values between 2 and  $3.5 m^2[leaf] * m^{-2}[soil]$ . Below  $2m^2[leaf] * m^{-2}[soil]$  and down to  $1.5 m^2[leaf] * m^{-2}[soil]$  (minimum value measured in our study), sap flow velocity also responded to changes in

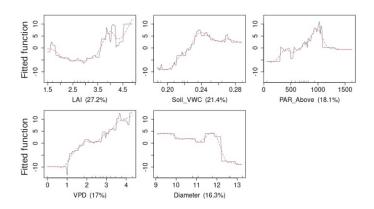


Figure 9: Overview of cacao's daily sap velocity response to the main environmental drivers according to a BRT model (model 2). The responses are calculated for daily averages during the entire period of this study. Functional shapes of the response of sap velocity to the environmental variables: VPD, PAR, Soil VWC, LAI and tree diameter. Each variable is presented with the relative importance of its variation on that of sap velocity within parentheses. The red dotted lines represent the LOESS smoothing for each response function.

LAI, increasing as LAI was lower. Sap flow velocity increased linearly with 505 soil VWC until a threshold level of 0.24  $m^3 * m^{-3}$  (approximately -0.16 MPa 506 for our soil), past which it plateaued. The shape of the response curve of sap 507 velocity to mean daily PAR was almost linear in the range from 0 to 1100 508  $\mu mol * s^{-1} * m^{-2}$ . Beyond this value, the response of sap velocity captured by 509 the BRT model leaned on few points, thus was not considered reliable. Finally, 51 C the mean daily sap velocity increased linearly with daily average VPD in a sim-511 ilar fashion to that observed in the half-hourly model 1 (Fig. 6, 9). 512

513

A significant interaction was found between LAI and soil VWC and, to a lesser extent, between VPD and PAR (Tab. 2). When LAI was low, sap velocity increased slightly with soil VWC (Fig. 10). But when LAI reached the value of  $3 m^2 * m^{-2}$ , the increase with soil VWC became much more important and sap velocity reached very high values. The interaction between average daily VPD and PAR, although less pronounced, showed that the increase in VPD and PAR may have a synergistic effect on sap velocity (Fig. 11). As PAR increased (up

	PAR_Above	LAI	$Soil_VWC$	VPD	Diameter
PAR_Above	0	2692.77	340.99	5507.51	4970.65
LAI	0	0	19350.06	5019.99	5275.63
Soil_VWC	0	0	0	510.93	1408.86
VPD	0	0	0	0	2304.21
Diameter	0	0	0	0	0

Table 2: Interactions between the variables considered in the BRT Model 2 in explaining cacao sap velocity at a sub-daily timescale. For each pair of variables, the table reports the mean value of the residuals, which represents the strength of the interaction. In bold: the interactions considered strong, thus significant, by the model.

to 1100  $\mu mol * s^{-1} * m^{-2}$ ), sap velocity responded more readily to daily average VPD and vice versa. Few PAR values were recorded above 1100  $\mu mol * s^{-1} * m^{-2}$ and the apparent decrease in sap velocity after this value was not interpretable. At the one-day time scale, no interaction between mean soil VWC and VPD was detected by model 2 (Tab. 2), similarly to what was found at the sub-daily timescale (Tab. 1).

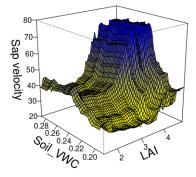


Figure 10: Three-dimensional partial dependence plots for the interaction between soil VWC  $(m^3 * m^{-3})$  and LAI  $(m^2[leaf] * m^{-2}[soil])$  in the BRT model 2 for daily average sap velocity  $(cm * h^{-1})$  in cacao trees. All variables except those plotted are held constant at their mean values.

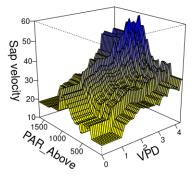


Figure 11: Three-dimensional partial dependence plots for the interaction between PAR  $(\mu mol * s^{-1} * m^{-2})$  and VPD (kPa) in the BRT model 2 for daily average sap velocity  $(cm * h^{-1})$  in cacao trees. All variables except those plotted are held constant at their mean values.

## 527 4. Discussion

# 528 4.1. Climate

The measured climate of the 2019-2020 dry season (Fig. 2) was representative for the conditions experienced by cacao during this time of the year in the Eastern region of Ghana. As soil VWC approached the wilting point in the control plot, air humidity dropped even below 20% and temperatures were suboptimal (Fig. 2, Fig. A.1), on a daily basis, the measured conditions were found to be stressful according to the definitions given for soil and atmospheric stresses (see Section 1).

#### 536 4.2. Canopy

Despite the occurrence of high temperatures, the trees were able to cope 53 with the climate in December. We argue so because we did not observe any 538 decline in LAI in either watering treatment, in line with previous studies with 539 no atmospheric stress [67, 19]. Leaf shedding is a common response to drought of 540 tropical and subtropical species [68, 69], as it allows trees to reduce transpiration 541 rates and, hence, avoid hydraulic failure. In line with the idea that cacao trees 542 apply this drought-coping strategy, during the more severe part of the dry season 54 **3** (from January onwards), LAI decreased in both watering treatments, although 544 at a faster rate in the control plot (Fig. 3), supporting the idea that leaf shedding 545 helps cacao to cope with soil drought. However, given that LAI also decreased 546 in the irrigated plot throughout the experimental period, it is likely that trees 547 in the irrigated plot also suffered from water stress. It could be argued that the 548 irrigation level (60 L/tree on alternate days, equivalent to approximately 9.6 mm 549 of rain) was not sufficient to completely mitigate soil drought (Fig. 2), despite 550 being in line with the water requirement of a mature tree found in literature 551 [37]. 552

Alternatively, it can be interpreted as the atmospheric drought having contributed to leaf shedding. This latter interpretation is supported by the more abundant litter production observed in December, when the soil VWC was significantly higher in the irrigated plot than in the control but air humidity and temperature reached values above stressful thresholds (Fig. A.1, B.2). The timing of the drop in LAI suggests that the key atmospheric condition impacting the canopy was the extremely low air humidity to which cacao leaves might not be adapted, especially considering that cacao developed under very high relative air humidity [70].

By the end of the dry season, the control plot had lost two thirds of its 562 foliage, twice the loss in the irrigated plot. In light of the lack of interaction 563 between VPD and soil VWC in explaining sap velocity (Tab. 1, 2) it could 56 be assumed that the effects of atmospheric and soil stress on LAI were inde-565 pendent. If this assumption is accepted, the difference in LAI drop between 566 the two plots must have been driven by the difference in soil VWC (Fig. 2, 3). 567 The fact that the net loss in LAI in the control relative to that in the irrigated 568 treatment (-62.5% vs. -32.5% over the course of the experiment, respectively)569 was much larger than the difference in litterfall between the two treatments (309 570  $g[dry] * m^{-2}$  and 247  $g[dry] * m^{-2}$ , respectively - a relative difference of 25%) 571 suggests that some leaf flushing might have occurred in the irrigated treatment. 572 This hypothetical leaf flushing in the irrigated treatment never sufficed to coun-573 terbalance the leaf shedding, possibly due to a limitation in carbon to invest in 574 leaf production, resulting in a drop in LAI. Further, more specific, studies to 575 investigate the effects of atmospheric and soil drought on leaf flushing dynamics 576 would be needed in order to properly characterize such effects. 577

It is acknowledged that the irrigation might have been sub-optimal but, based 578 on the above reasoning, the loss of canopy density in the irrigated plot was 579 imputable mostly to the atmospheric stress. In the control treatment the ad-580 ditional loss was caused by the soil stress and the interaction between the two 581 stresses, if present. Thus, our results on the LAI evolution and litterfall produc-582 tion show that the impact of atmospheric stress (VPD) on the canopy during 583 the dry Harmattan season may be of similar magnitude, if not greater, than 584 that of soil stress (soil VWC), regardless of the possible interaction between the 585 two stresses (Fig. 3). If the two stresses were independent, then the depressing 586 effects of soil and atmospheric stress were of the same magnitude, one third 587

of the initial canopy density each. Alternatively, the existence of an interac-588 tion between the two would mean that the soil stress impacted the canopy less 589 than the atmospheric one. The major role found for the aerial stress is in line 590 with Hutcheon at al. (1973) [71], who concluded that the beneficial effect of 591 irrigation on cacao total biomass production was limited when in presence of 592 atmospheric stress. Other authors reported the same [17, 37] but, to our knowl-593 edge, this was the first attempt to quantify the effects under field conditions in 594 West Africa. Our results contrast with those from a study in which LAI did 595 not change in response to a 73% reduction in precipitation over 13 months [19]. 596 It must be noted that the experimental conditions between this experimental 597 site in Sulawesi, Indonesia and ours in eastern Ghana differ greatly. For the site 598 in Sulawesi, it was suggested that the reduction in incoming precipitation was 599 insufficient to impose stressful conditions due to the high mean annual rainfall 600 (2,844 mm) [72]. Furthermore, at the Indonesian site, daily mean relative air 601 humidity never dropped below 69%, in contrast to our study site, where relative 602 air humidity often fell below 40%, and where we found a significant drop in LAI, 603 irrespective of the irrigation regime. 604

#### 605 4.3. Sap velocity

The two models developed (half-hourly: BRT model 1, daily: BRT model 2) 606 allowed us to obtain a good estimate of the environmental and climatic drivers of 607 sap velocity on a half-hourly and daily scale. This allowed an effective investiga-608 tion of the impacts of atmospheric and soil water stress on cacao during Harmat-609 tan season. Up to now, in cacao, the response of transpiration to atmospheric 61 0 drought stress had only been studied under controlled conditions, in young in-611 dividuals and never in combination with limited water availability [13, 14, 15]. 61 2 Previous studies addressing the effects of climatic factors (VPD, soil VWC and 61 3 PAR mainly) on transpiration did not find clear patterns [13, 17]. Fraga et al. 614 2020 [13] argued that large temporal and spatial heterogeneity, mainly in VPD 615 and soil water, could account for some of the unexplained variability in transpi-616 ration. We argue that our approach of separating the instantaneous half-hourly 61 7

effect from the integrated daily effects allowed us to disentangle some of these patterns. Furthermore, to our knowledge, this is the first study that addressed the effect of the interactions among variables on sap velocity.

Soil VWC and LAI did not change significantly at the time scale of one day. 621 Despite their lack of variation at the sub-daily level, soil VWC and LAI varied 622 over the season, setting different daily conditions for sap velocity variation in 623 response to the variability of the remaining variables (PAR, VPD) (Fig. 4,5,6). 624 The BRT model 1 showed that, of these two conditions, soil VWC mattered 625 more than LAI (Fig. 6). This suggests that water availability matters more 626 than the total leaf area available for transpiration in determining sap velocity. 627 In cacao the transpiration is largely performed in the outer crown, while the 628 shade leaves are far less active [17, 67]. The loss of foliage probably interested 629 mostly sun leaves [67] but the trees were probably able to adapt the lower strata 630 of the canopy to the new conditions [73, 74], minimizing the direct effect of LAI 631 on sub-daily sap velocity. This was reflected in the flat response of sap velocity 632 to LAI in model 1 (Fig. 6). Instead, the different soil VWC over the season 633 appeared to matter more in determining the sub-daily sap velocity, possibly 634 hinting that cacao might be more limited by its root water uptake and by the 635 conductivity of the vascular system rather than by the canopy conductance, in 636 line with [16]. 637

Within a day, sap velocity responded mainly to the variation of PAR and VPD 638 and their synergistic effects (Fig. 7). At the sub-daily time scale, PAR was the 639 principal driver for sap velocity, as long as VPD did not exceed the mean daily 640 cycle by more than 2 kPa. Beyond this value for the anomaly, it is possible 641 that for most part of the day stomatal regulation was no longer effective in 642 dealing with the high water demand from the atmosphere. This is supported 643 by the fact that irrigation was not able to limit water loss when VPD was ex-644 tremely high and the irrigated trees ended up transpiring as much as the trees 64 5 in the control plot (Fig. 5 A). Moreover, the observed peak in sap velocity up 646 to 20 cm/h in the morning cannot be explained by the extremely low values of 647 VPD and PAR (Fig. 5 A, B). However, said peak may be explained as a phe-648

nomenon of recovery from embolism [75]. It has been described in other plants 649 that, following embolism due to hydraulic stress, the plant applies a positive 650 root pressure to force the gas to dissolve. Such recovery strategy of embolism 65 removal is often put in place concurrently to the start of transpiration in the 652 morning [76, 77, 75]. The control trees, constantly stressed by the low VWC, 653 may have had to adopt this strategy regardless of the VPD level, as the early 654 morning peak in sap velocity suggests (Fig. 5 A,B). The irrigated trees did not 65 5 present the same peak in the early morning in days with moderate-low average 656 VPD but irrigation might have not prevented embolism when VPD was high, 657 leading to an early morning sap velocity peak compatible with the refilling of 658 vases (Fig. 5 A, B). This behaviour is congruent with the idea that irrigation 659 is not sufficient when the atmospheric stress is too strong. 660

Cacao plants naturally occur in the understorey of tropical forests, where light 661 is limited and VPD is rarely high [78]. Hence, we could expect that stomatal 662 behavior is finely-tuned in cacao to respond to variations in light availability, 663 to maximize photosynthesis [74, 79], but it might not be adapted to regulate 664 water loss under increasing VPD, as we observed in our experiment (Fig. 4). 665 Before this study, the atmospheric component of water stress had rarely been 666 taken into account because such conditions are seldom met in most of the cacao-667 growing areas worldwide [18, 72]. Nonetheless, most of the West African cacao 668 belt undergoes atmospheric stress on a quasi-seasonal basis under the influence 669 of the Harmattan winds [23] and the chances of harsher atmospheric stress in 670 the region will increase with climate change. Given that West Africa includes 671 the two leaders of cacao production worldwide (Ivory Coast and Ghana) [80], 672 our study highlights the importance of recognizing the due importance of the 673 atmospheric stress. 674

At the seasonal time scale (effects on daily averages), the effects of PAR and VPD, as well as their interaction, were maintained. Together with PAR and VPD, at the seasonal time scale soil VWC played a major role as well (Fig. 9, 11). The response curve of sap velocity to soil VWC in figure 9 highlights that, under our experimental conditions, the soil reached a critical soil VWC at which

cacao water extraction capacity was challenged  $(0.24 \ m^3 * m^{-3})$ , approximately 680 -0.16 MPa for our soil). A significantly lower critical value of soil water poten-681 tial for water extraction (-0.079 MPa) has previously been reported for young 682 cacao plants under field conditions in Brazil [13]. The difference between the 683 Brazilian study and ours is possibly related to the differences in genetics, age, 684 and rooting depth considered in the two experiments. Here, we studied mature 685 trees subject to water stress on a regular basis every year. These two elements 686 (age and prevailing climate) imply that our trees would have had a more devel-687 oped rooting system, capable of exploring a larger soil volume. 688

Contrary to our initial expectations, our results from the BRT models did not 689 clearly show an interactive effect between atmospheric and soil drought on ca-690 cao transpiration, represented in the models by VPD and soil VWC (Tab. 1, 691 2). Yet, our results cannot completely rule out our initial hypothesis as, for 692 example, Model 1 highlighted a not significant but strong link between VPD 693 and soil VWC (Tab. 1). The BRT models could have failed at capturing such 694 interaction because of the paucity of observations when atmospheric and soil 695 drought co-occurred in the control plot, a key requirement to train the model. 696 Furthermore, the provided level of irrigation might have not been sufficient to 697 completely alleviate the soil stress in the irrigated part. If this was the case, 698 the small  $\Delta$  VWC between plots might have made more complicated for the 699 model to capture an interaction between soil and atmospheric drought across 700 treatments as well. 701

At the seasonal time-scale (daily average measurements), we found that the re-702 sponse of sap velocity to both atmospheric and soil drought was modulated by 703 the interaction with LAI. We found that beyond certain threshold values of both 704 VWC (0.24  $m^3 * m^{-3}$ ) and LAI (3.5  $m^2[leaf] * m^{-2}[soil]$ ), sap velocity did not 705 respond to further increases in either variable. The daily average sap velocity 706 measured under these conditions, high VWC and LAI(76  $cm * h^{-1}$ ), could be 707 the maximum supported by the root and vascular system of cacao, although 708 this should be further tested in other climates. When soil VWC was below 0.24 709  $m^3 * m^{-3}$ , sap velocity was strongly responsive to soil VWC variations, regard-71 0

less of the LAI, whereas for a given soil VWC, the response of sap velocity to 711 changes in LAI was less pronounced (Fig. 10). This behaviour is compatible 71 2 with the higher importance of soil VWC over LAI found in model 1 (Fig. 6) 71 3 and with the more dynamic shape for the response of sap velocity to soil VWC 714 and flatter one for LAI in model 1 (Fig. 6) and for most of the range (1.5 to 3.5) 715 in model 2 (Fig. 9). Furthermore, the irrigated trees only showed signs of prob-716 able embolism recovery in the mornings of days with a high VPD, suggesting 717 that they were not able to uptake enough water from the soil to avoid embolism 71 8 (Fig. 5 A), while in the control the morning peak in sap velocity, that we pro-71 9 posed as a sign of embolism recovery, was present with approximately the same 720 magnitude under high and low VPD (Fig. 5 A,B). This suggests that the main 721 limitation for water transportation in the soil-plant-atmosphere continuum was 722 found in the ability of the plant to extract water from the soil. The results dis-723 cussed in this paragraph support the theory that cacao's transpiration is mostly 724 limited by the root water uptake capacity rather than by the conductivity of 72! the vascular system or the total leaf area, in line with [16]. 726

#### 727 5. Conclusions

For the first time, in this study, we assessed the effects of soil and atmo-728 spheric water stress on canopy transpiration of adult cocoa trees. We showed 729 that under high soil and atmospheric water stress, irrigation decreased leaf shed-730 ding in response to limited water availability in the soil. Nonetheless, under a 731 climate change scenario with harsher conditions experienced by cacao under the 732 influence of the Harmattan winds, irrigation might not suffice to sustain cacao 733 production. In fact, our study highlights that cacao transpiration increases with 734 high VPD, which could further compromise soil water availability and eventu-735 ally aggravate soil stress. The use of shade nets or of shade trees with a deep 736 rooting system, as previously suggested [81, 82], could be a key requirement for 737 cacao farming in the future to partially alleviate atmospheric drought stress. 738 Deep-rooted vegetation and/or shading nets diminish the detrimental impact 739

of increased atmospheric drought by buffering temperature and relative air hu-74 0 midity. This prevents stressful values for temperature and relative air humidity 741 without adding competition for water extraction. Still, to predict the actual 74 2 impact of increased drought severity under climate change on cacao production, 74 **3** it will be necessary to look beyond the effects of soil and atmospheric drought 744 on transpiration and leaf area, and study how these conditions affect flower, 74 5 fruit and seed production in the field. Furthermore, other similar studies are 746 needed to validate and strengthen these conclusions, which rely on data with 747 limited replicates that may limit the possibility of generalizing these findings for 748 other situations. In any case, this study has confirmed in the field what previous 749 studies had found in controlled environments and proposed new results that are 750 in line with the behaviour in similar environments of cacao and other species. 751

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# 762 Appendix A. Atmospheric stressors

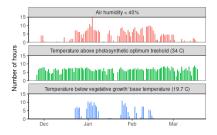


Figure A.1: Number of hours per day under stressful climatic conditions for cacao trees according to values reported in [17, 26, 27, 28, 21]; each bar is one day. Top: air humidity below 40%; middle: temperature above photosynthetic optimum threshold (34 °C); bottom: temperature below the base temperature for vegetative growth (19.7 °C).

### 763 Appendix B. Leaf area index

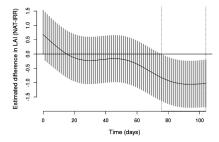


Figure B.1: Difference in estimated leaf area index (LAI) between the irrigated and the control plot. The fitted line is a generalized additive model and the shaded area is the 95% prediction interval. A shaded area non-overlapping with the zero-line indicates a significant difference between watering treatments.

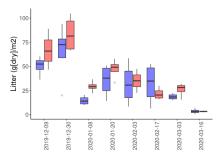


Figure B.2: Litter production of cacao trees during the dry Harmattan season, in presence (blue) and absence (red) of irrigation. Box-plots represent variations between the four litter traps per treatment and date. Bold lines represent means per plot and date.

## 764 Appendix C. Boosted Regression Trees analysis

### 765 Appendix C.1. parameterization of the models

	Step	Family	tl	bg	nt	tc	lr
Model 1	30 min	Gauss	0.01	0.75	100	3	0.1
Model 2	1 day	Gauss	0.001	0.75	100	3	0.0

Table C.3: Parametrization of the BRT models to predict the transpiration during the day at a 30-minute time interval (Model 1) and at a daily time step (Model 2). A BRT model requires the definition of a set of hyperparameters:

family - the type of statistical error distribution characterizing the data, in our case Gaussian; tolerance (tl) - the threshold in variance change under which to stop the model reiteration;

bag fraction (bg) - the fraction of dataset to use for the training of each tree. The remaining data is used at each step to cross validate the set of relations found;

number of trees (nt) - the number of trees necessary for optimal prediction. It is determined based on tl and bg;

tree complexity (tc) - the maximum level of interaction between variables to consider in a tree;

learning rate (lr) - determines the contribution of each tree to the growing model. The seed number used was - 210920.

## 766 Appendix C.2. Evaluation of BRT models on the test data

To allow an independent evaluation of the BRT models, 25% of the field data was set aside to be used as a test subset of data and was not used at any <sup>769</sup> point in the training of the BRT models. This section of the appendix presents the evaluation of the models (Model 1, Model 2) on this test data subset.

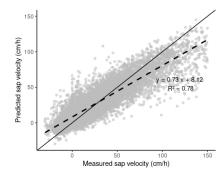


Figure C.1: Observed (abscissa) vs. predicted (ordinates) sap velocity, the corresponding linear regression (dashed line) and y=x line (full black line) for the BRT half-hourly model (model 1) test data.

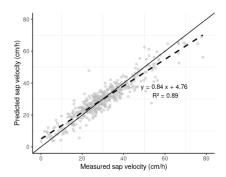


Figure C.2: Observed (abscissa) vs. predicted (ordinates) sap velocity, the corresponding linear regression (dashed line) and y=x line (full black line) for the BRT daily model (model 2) test data.

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# Highlights

Assessment of atmospheric and soil water stress impact on a tropical crop: the case of *Theobroma cacao* under Harmattan conditions in eastern Ghana.

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- Soil and atmospheric water stresses' effect on cacao's transpiration are independent
- Atmospheric stress reduced the canopy density by one third regardless of irrigation
- An extra third of the canopy density was lost due to soil water stress in the control
- Soil moisture and air vapour pressure are the key drivers of cacao sap velocity
- Cacao transpiration appears more limited by root uptake than canopy conductivity