The role of the intraspecific variability of hydraulic traits for modelling the plant water use in different European forest ecosystems

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Key Points:

- We explore the impact of the intraspecific variability of plant hydraulic traits on the simulated transpiration by CLM5.

- We find that a choice of plant hydraulic traits that reproduces observed plant transpiration also reduces simulated water stress.

- We demonstrate the critical role of the maximum xylem conductance in the model and its dependency on factors other than vegetation type.

17 **Abstract**

18 The drought resilience of forest ecosystems is generally believed to depend on the dominant tree
19 species' hydraulic traits. These traits define the maximum water transport capacity and the
20 degree of vulnerability to hydraulic failure of a tree species. This work evaluates the effect of the
21 intraspecific variability of hydraulic traits on the simulated tree water use in the Community
22 Land Model (CLM, version 5.0). We selected two contrasting broadleaved tree species and
23 performed a series of numerical experiments by modifying the parameters of the plant
24 vulnerability curve and the maximum xylem hydraulic conductance accounting for the variability
25 within each species. Our prescribed parameter sets represent vulnerable and resistant tree
26 responses to the water deficit. At sites with an ample water supply, the resistant configuration
27 simulates reduced water stress and increased transpiration compared to the vulnerable
28 configuration. Meanwhile, the model results are counter-intuitive at temporarily dry sites when
29 water availability is the limiting factor. The numerical experiments demonstrate the emergent
30 role of the maximum xylem conductance as a modulator of the plant water use strategy and the
31 simulated transpiration within the model. Using the default value for maximum xylem
32 conductance, the model tends to overestimate the early summer transpiration at drier sites,
33 forcing the vegetation to experience unrealistic water stress later in the year. Our findings
34 suggest that the parameterization of maximum xylem conductance is an important yet unresolved
35 problem in the CLM and similar land surface models.

36

37 **Plain Language Summary**

38 The survival of trees in drought conditions depends on their ability to adapt to water scarcity.
39 Part of this adaptation is characterized by specific plant traits, which are an important component
40 of Land Surface Models and largely determine the relationship between soil moisture and canopy
41 gas exchange. Our study explores how the variability of specific plant traits of individual tree
42 species may affect the selected model's ability to reproduce the water use observed in forest
43 stands in Europe. For climates with a pronounced summer dry period, we found that the default
44 model settings overestimated the vegetation water use in the early growing season, when water is
45 abundant, resulting in severe water stress and underestimated transpiration as the dry season
46 progressed. We specifically demonstrate that a rarely considered plant trait, representing the
47 maximum water transport capacity, plays an essential role in controlling the magnitude of
48 simulated water use and that adjustments to this parameter greatly help to reproduce the
49 vegetation water use observed in seasonally dry climates.

50 **1 Introduction**

51 The recent worldwide increase in drought incidence and severity (He et al., 2020) has
52 been associated with alterations in the soil carbon and nitrogen dynamics (Deng et al., 2021),
53 high rates of tree mortality (Powers et al., 2020; Senf et al., 2020) as a consequence of the high
54 atmospheric water demand (Hammond et al., 2022; McDowell et al., 2022), and a diminution in
55 forest evaporation (Lansu et al., 2020; Lindroth et al., 2020). The severity of drought impacts on
56 forest ecosystems and the spatial extent of them depends on the vegetation resistance and
57 resilience to water scarcity. The latter reflects the admixed vulnerability of individual trees
58 (Haberstroh & Werner, 2022) and is partly driven by the safety mechanisms used to overcome
59 disturbances in the whole tree hydraulic system (Arend et al., 2022).

60   The resistance of a species to water stress is commonly expressed in the plant
61 vulnerability curve and the recovery patterns shown by the tree species (Klein et al., 2018). This
62 curve postulates a continuous decline of plant organ conductance (e.g., roots, branches, leaves)
63 with declining water pressure in the plant organ (Sperry & Love, 2015; Venturas et al., 2017).
64 The parameters of the curve differ among and within tree species (Rosner et al., 2019), and are
65 influenced by the provenance of the species (Hajek et al., 2016; Lobo et al., 2018) and xylem
66 features (Pereira et al., 2018). The plant vulnerability curve has been analyzed across species and
67 biomes (Choat et al., 2012), allowing the degree of vulnerability to hydraulic failure to be
68 quantified (Venturas et al., 2017). Vulnerable trees commonly have a low wood density, an early
69 loss of conductance, and a small threshold between unstressed conditions and the occurrence of
70 hydraulic damage. Resistant trees have vulnerability curves ranging from gradual to steep
71 responses at lower water potentials. These trees commonly have large safety margins and high
72 wood density (Johnson et al., 2012; Meinzer & McCulloh, 2013; Mrad et al., 2019). The degree
73 of vulnerability to hydraulic failure has been related to the trade-off between xylem safety and
74 efficiency for many tree species (U. G. Hacke et al., 2006; Venturas et al., 2017). This trade-off
75 requires the coordination of the plant hydraulic traits and water use strategy (WUS), which
76 ranges from aggressive to conservative (Flo et al., 2021; Mrad et al., 2019). Also, the WUS is
77 influenced by the stomatal regulation capacity of the tree species (Konings & Gentine, 2017),
78 modulated by the vapor pressure deficit (Novick et al., 2019), and driven partially by the soil
79 water content (Fu et al., 2022). However, the current evidence does not allow the generalization
80 of this trade-off for all plant species (Gleason et al., 2016).

81   The plant hydraulic theory is numerically implemented in models using either a plant
82 pipe model, a porous media model, an electrical analogy model (Li et al., 2021), or optimality-
83 based models (Sabot et al., 2020; H. Xu et al., 2021). Plant pipe models follow the Hagen-
84 Poiseulle law and require the use of allometric scaling laws (Li et al., 2021; Mrad et al., 2018),
85 whereas porous media models are based on Richards equation assuming that water movement
86 through the xylem mimics an unsaturated porous media flow (Christoffersen et al., 2016; Li et
87 al., 2021). The electrical analogy models resemble an electrical circuit with resistance and
88 capacitance parameters that control the water flow following Darcy's law (Bonan et al., 2014; Li
89 et al., 2021). Finally, optimality-based models do not prescribe hydraulic traits based on
90 observations, but assume that vegetation finds its "optimum" by maximizing a carbon-related
91 goal function for given environmental conditions (Joshi et al., 2022; Sabot et al., 2020). An
92 electrical analogy model has low to moderate computational requirements making it a suitable
93 model for implementation in large scale Land Surface Models (LSMs). For example, the
94 Community Land Model 5.0 (CLM5, Lawrence et al., 2019) implements an electrical analogy
95 model using the plant vulnerability curve to downscale the segment conductance according to the
96 percent loss of conductance (PLC) (Kennedy et al., 2019). Given its recent implementation, the
97 simulated plant hydraulic response (e.g., vulnerability to hydraulic failure) of CLM5 during
98 drought conditions and across different forested ecosystems has not yet been evaluated in detail.
99 Specifically, it has never been examined in detail to what extent the current (and default) plant
100 hydraulic formulation and parameterization of the model reproduces realistic transpiration rates
101 and plant water status under varying soil moisture availability and atmospheric water demand.

102   The implementation of plant hydraulic formulations for LSMs from a species-specific
103 perspective is rare (e.g., De Kauwe et al., 2022; Sabot et al., 2020), and most of the current
104 LSMs rely on the definition of plant hydraulic traits within the Plant Functional Type (PFT)
105 classification framework (e.g., D. M. Lawrence et al., 2019; Zhang et al., 2022) . This

106 classification assumes that hydraulic traits are spatially homogeneous and temporally fixed
107 within predefined vegetation categories, which is equivalent to assuming the same drought
108 sensitivity within the same PFT class. Several studies have addressed the implications associated
109 with the loss of diversity in the PFT classification in terms of water and carbon dynamics by
110 using deterministic or stochastic coordinated plant attributes (Christoffersen et al., 2016; Pappas
111 et al., 2016; Y. P. Wang et al., 2012; X. Xu et al., 2016) or by exploiting plant trait-climate
112 relationships (Verheijen et al., 2013). A recent study by Butler et al. (2022) showed that the
113 aggregation of allocation and hydraulic traits into PFTs reduces the productivity of the modelled
114 ecosystem with respect to the flux data of sites with a strong dependency on vegetation
115 phenology. Liu et al. (2021) argues the negative impacts of generalizing the plant hydraulic traits
116 at PFT level and proposed a set of hydraulic functional types as an alternative to current PFTs.
117 Overall, representing the plant trait inter- and intraspecific diversity within the PFT broad
118 classification scheme remains a challenging task requiring the characterization of the emergent
119 plant response by coordinating water use strategies with the xylem vulnerability (Skelton et al.,
120 2015). While the trade-off between plant hydraulic traits has been addressed in previous studies
121 using detailed plant hydrodynamic models (e.g., Mirfenderesgi et al. (2019)), the coordination
122 between hydraulic traits and water use strategies remains largely unexplored, with some attempts
123 carried out aiming to improve the plant hydraulic framework implemented in LSMs (e.g., Eller et
124 al., 2020; Sabot et al., 2020). Addressing this issue may provide an opportunity to define optimal
125 strategies for large-scale parameterizations of key plant hydraulic traits (e.g., maximum xylem
126 conductance), which are rarely documented in existing hydraulic trait databases (Liu et al.,
127 2021).

128   This manuscript aims to evaluate the effect of the intraspecific variability of plant
129 hydraulic traits on the simulated transpiration response of two contrasting tree species in CLM5.
130 The intraspecific variability of plant hydraulic traits defines the spectrum of vulnerability
131 responses to hydraulic failure and the water use strategies of each species. This spectrum
132 considers that individual tree species have different boundaries determining their degree of
133 vulnerability to hydraulic failure. Our hypothesis is that vulnerable trees transpire more than
134 resistant trees under unstressed water conditions and strongly reduce transpiration during dry
135 periods. On the other hand, resistant trees maintain low transpiration rates but experience less
136 stress on the plant hydraulic system. This hypothesis is evaluated for two broadleaved tree
137 species, *Quercus ilex* L. and *Fagus sylvatica* L., with contrasting phenologies and provenances.
138 For each species, we distinguish between a resistant and vulnerable hydraulic trait configuration
139 by extracting from the reported parameter sets for that species the plant vulnerability curves with
140 the minimum and maximum xylem pressure inducing 50% loss of hydraulic conductance (Ψ_{p50})
141 value, respectively. The results of point-scale numerical experiments with CLM5 based on each
142 parameterization are compared to the sap flux observed at four experimental sites across Europe.
143 The representation of the simulated vulnerability to hydraulic failure and the water use strategy
144 of each species are interpreted using the simulated leaf water stress factor (β) and percent loss of
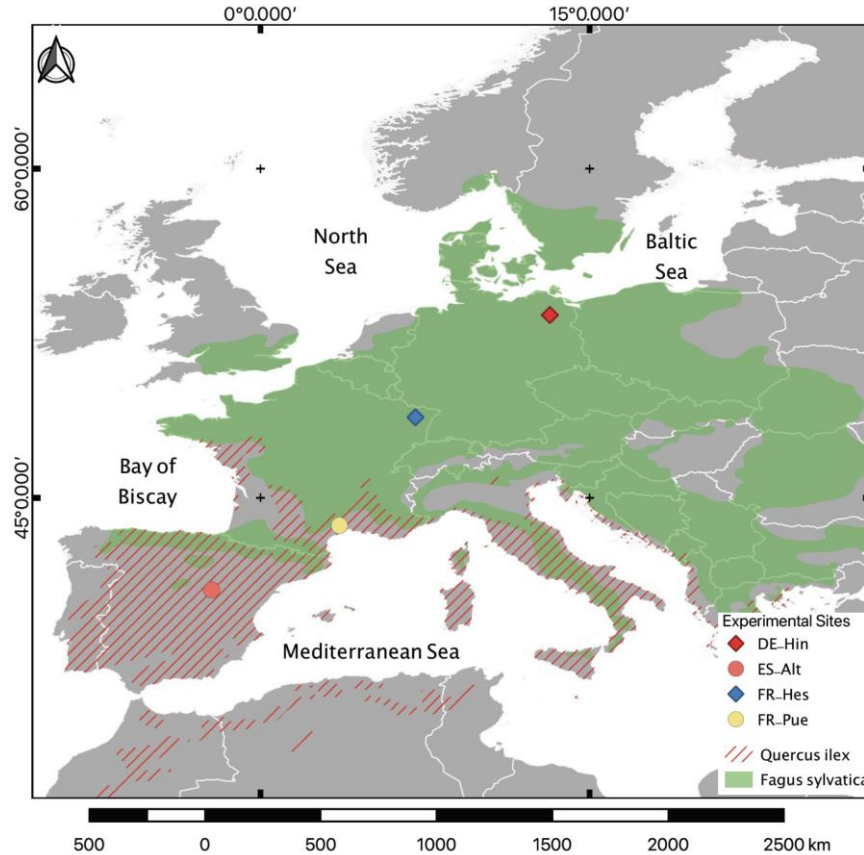145 conductance (PLC) in different plant organs.

2 Materials and Methods

2.1 Tree Species and Experimental Sites

148   The tree species selected for this study, *Fagus sylvatica* L. and *Quercus ilex* L., belong to
149 the same botanical family (Fagaceae) but differ in their phenology and spatial distribution in

150   Europe (Figure 1). *Fagus sylvatica* is a deciduous broadleaved tree distributed in Central and
151   Western Europe, from Southern Italy to Southern Norway. This tree species grows from sea level
152   to 1000 m a.s.l., with a higher upper elevation limit in dryer regions. It does not survive in
153   locations with poor drainage or stagnant water, and its relatively shallow root system makes it
154   susceptible to emerging hot droughts and high temperatures (Houston Durrant et al., 2016; von
155   Wuehlisch, 2008). *Quercus ilex* is a broadleaved evergreen species that grows as a tree or shrub.
156   It inhabits the Mediterranean basin from the coast up to 1800 m a.s.l. This species can survive
157   low temperatures and its sclerophyllous character increases its resistance to drought by reducing
158   water loss during dry periods (de Rigo & Caudullo, 2016; Schirone et al., 2019).

159   Two experimental sites for each species were selected from the SAPFLUXNET database
160   (Poyatos et al., 2020). *Fagus sylvatica* is the dominant tree species in Hesse (France, FR-Hes)
161   and Hinnensee (Germany, DE-Hin), over the sampling periods of 2001-2005 and 2012-2014,
162   respectively (Table 1). Both sites have a temperate oceanic climate (Cfb) according to Köppen-
163   Geiger's climate classification (Beck et al., 2018), with no significant intraseasonal precipitation
164   variability. The stand age marks the main difference between these two sites; trees in FR-Hes
165   were 34 years old during the selected measurement period while those in DE-Hin were more
166   than 200 years old. The mean tree diameter reflects this age difference, with 12.9 cm at FR-Hes
167   and 43.6 cm at DE-Hin. *Quercus ilex* is the dominant tree species in Puechabon (France, FR-
168   Pue) and Alto Tajo (Spain, ES-Alt). These sites cover the monitoring periods 2001-2005 and
169   2012-2014, respectively. The climate differs slightly between these two sites; FR-Pue has a hot-
170   summer Mediterranean climate (Csa) while ES-Alt has a warm-summer Mediterranean climate
171   (Csb). The different elevations of the sites explain the differences in climate classification (Table
172   1). Despite a lack of differences in the stand age between these two sites, the diameter recorded
173   for the trees in FR-Pue (9.1 cm) is much smaller than the diameter in ES-Alt (24.4 cm).

174

Figure 1. Geographical location of the selected experimental sites and the spatial distribution of *Fagus sylvatica* L. (diamonds) and *Quercus ilex* L. (circles) and their spatial distribution across Europe. The spatial distribution of the tree species is based on Mauri et al. (2022).

2.2 Model Setup

The Community Land Model version 5.0 (CLM5, Lawrence et al. (2019)) was applied at each experimental site using point-scale setups. Hourly atmospheric forcing was retrieved from the SAPFLUXNET dataset. This dataset includes precipitation, wind speed, air temperature, relative humidity, and incoming shortwave radiation. The incoming longwave radiation was calculated according to An et al. (2017) using the vapor pressure deficit and temperature. The COSMO-REA6 reanalysis product (Bollmeyer et al., 2015) with a temporal and spatial resolution of one hour and 0.05°, respectively, was used to fill in the missing variables (i.e., atmospheric pressure) and temporal data gaps for each site. The monthly leaf area index (LAI) in m^2 m^{-2} is based on the 8-days' time-series of the Global Land Surface Satellite (GLASS) product (Liang et al., 2013, 2014) that has a spatial resolution of 0.05° for the different periods under analysis. The monthly stem area index (SAI) in m^2 m^{-2} was retrieved from the global surface dataset of the model as described in Lawrence & Chase (2010). The LAI of the sites covered with *Fagus sylvatica* trees was forced to 0.0 $m^2 m^{-2}$ from October to April. This decision was based on the observed phenology at FR-Hes (Q. Wang et al., 2005) and DE-Hin (Blume et al., 2022). The root area index (RAI) in $m^2 m^{-2}$ is calculated in the model (see Equation 2.11.15 of the technical documentation (UCAR, 2020)) based on plant functional type-specific parameters such as the LAI, SAI, root fraction in each soil layer, and the root-to-shoot ratio. The main soil

196 characteristics (e.g., soil texture, organic matter content) were taken from Bonan et al. (2002),
197 while the depth to bedrock was taken from Pelletier et al. (2016). Multi-year spin-up runs were
198 performed for each experimental site by reinitializing soil moisture and soil temperature until a
199 dynamic equilibrium condition was reached. The tree species at the selected sites pertain to two
200 distinctive plant functional types (PFTs), with *Fagus sylvatica* representing the Temperate
201 Broadleaf Deciduous Tree (BDT) in FR-Hes and DE-Hin and *Quercus ilex* representing the
202 Temperate Broadleaf Evergreen Tree (BET) in FR-Pue and ES-Alt; see Table 2 for the default
203 plant hydraulic configuration (DC) of these two PFTs.

204   2.3 Plant Vulnerability Curve

205   The plant vulnerability curve (PVC) implemented in CLM5 (Equation 1) determines the
206 plant segment specific hydraulic conductance k (mm_{H2O} $mm_{H2O}^{-1}s^{-1}$) based on three parameters:
207 the xylem pressure inducing 50% loss of hydraulic conductance (Ψ_{p50}, MPa), the non-
208 dimensional sigmoidal shape parameter of the curve (c_k), and the maximum plant hydraulic
209 conductance (k_{max}, mm_{H2O} $mm_{H2O}^{-1}s^{-1}$). The CLM5 plant hydraulic routine uses a plant
210 segmentation that differentiates between roots, stems, shaded and sunlit leaves. Each plant
211 segment uses k_{max}, Ψ_{p50} and c_k as static parameters that can be adjusted and may differ between
212 plant segments (i.e., root, xylem, and sunlit and shaded leaf) and PFTs. The plant hydraulic
213 system of CLM5 uses k to determine the flux per plant segment by applying a Darcy's law
214 equation, where the reference area varies between plant segments: the leaf area index (LAI, m^2m^{-2})
215 for the stem-to-leaf, the stem area index (SAI, m^2m^{-2}) for the root-to-stem, and the root area
216 index (RAI, m^2m^{-2}) for the soil-to-root segment. A detailed description of the equations used by
217 the plant hydraulic system of CLM5 is provided in Kennedy et al. (2019) and Lawrence et al.
218 (2019).

219

$$k = k_{max}2^{-\left(\frac{\Psi}{\Psi_{p50}}\right)^{c_k}}$$	Equation 1

220   2.4. Intraspecific Variability of Plant Hydraulic Traits

221 The intraspecific variability of both tree species was determined based on the loss of hydraulic
222 conductance by 12%, 50%, 88%, and in some cases at 10% (Ψ_{p12}, Ψ_{p50}, Ψ_{p88}, and Ψ_{p10},
223 respectively), as reported in the Xylem Functional Traits (XFT) database (Choat et al., 2012).
224 The limited data for *Fagus sylvatica* in the XFT database led to the selection of additional
225 experimental data to improve the representation of the species. Meanwhile, the data set of
226 *Quercus ilex* was subject to an additional screening of the individual records due to concerns
227 about the measurement of the PVC in different species of Oaks (Cochard et al., 2013) and in
228 particular to the open-vessels artefact issue affecting the measurements in *Quercus ilex* (Martin-
229 StPaul et al., 2014). Therefore, we conducted a literature review and selected those experiments
230 that implemented procedures to prevent such artefacts (e.g., excised under water). The complete
231 list of references used to retrieve the additional data for *Fagus sylvatica* and *Quercus ilex* is
232 available in Table S1. The c_k parameter of each dataset was determined by converting the
233 reported slope of the vulnerability curve at Ψ_{p50} to c_k or by solving the CLM vulnerability curve
234 for c_k and inserting any provided combination of PLC and Ψ_{p10}, or Ψ_{p12}, or Ψ_{p88} values reported
235 in the XFT database, with a preference for Ψ_{p10} or Ψ_{p12} if available. The procedure to determine
236 the c_k parameter assumes that Equation 1 follows the Weibull distribution, allowing the

237   vulnerability curve formulation from Domec and Gartner (2001) to be used. From this
238   formulation, we derived Equation 2 to calculate the c_k parameter based on the Ψ_{p50}, the slope of
239   the curve (s) at Ψ_{p50} (Pa^{-1}), and V as a constant dimensionless value of 34.66. To determine V, we
240   deduced Equation 3 from Domec and Gartner (2001) and inserted the percent loss of
241   conductivity (τ_{50}) of 50%. Equation 4 is used to calculate s (Pa^{-1}) using the slope at any specific
242   loss of conductivity (τ). This indicator is calculated with Equation 4 using τ in %, Ψ_{p50}, and Ψ_x
243   that represents the matric potential at the selected τ. Finally, the two curves with the highest and
244   lowest Ψ_{p50} values were selected for each species to represent the vulnerable (VC) and resistant
245   (RC) configuration, respectively (Figure 2); see Table 2 for more details on the obtained values.

246

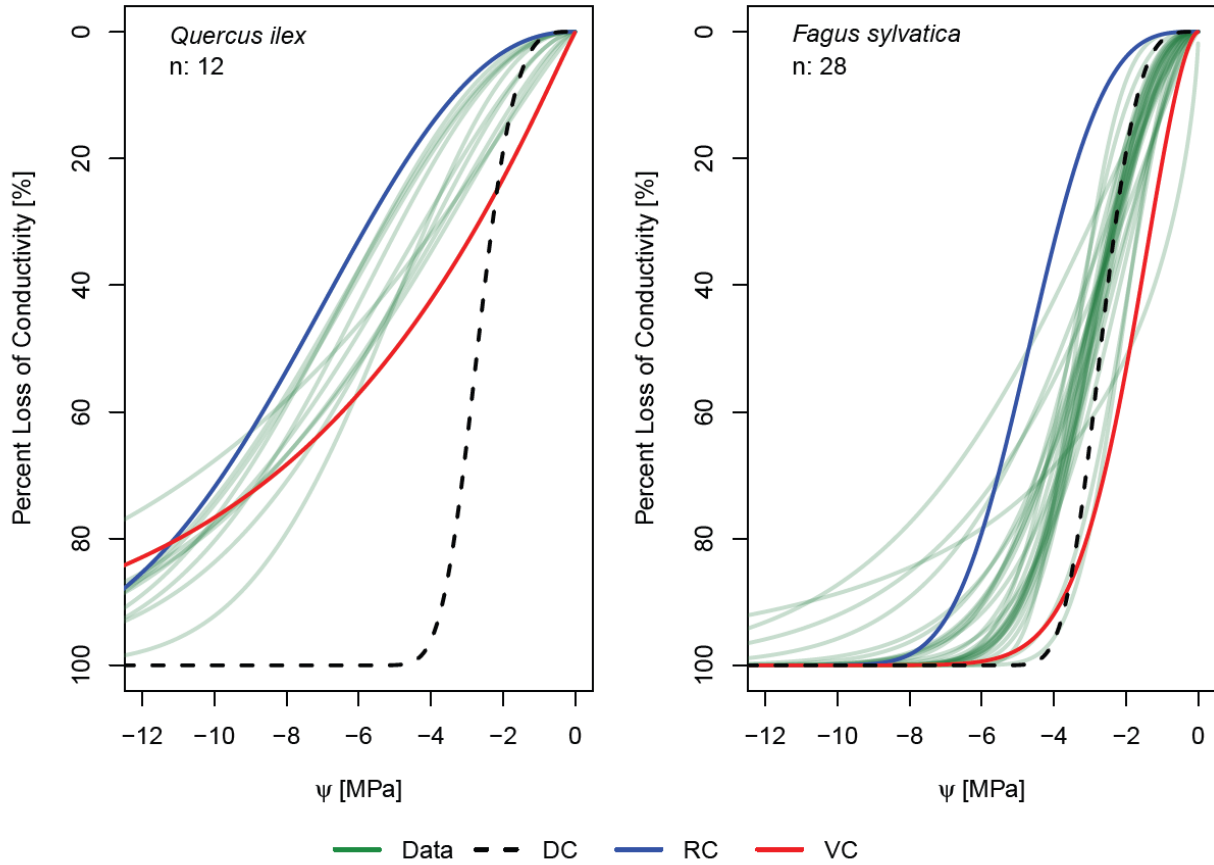| $$c_k = \frac{\Psi_{p50} \cdot s}{V}$$	Equation 2

247

$$V = (\tau_{50} - 100) \cdot \ln\left(1 - \frac{\tau_{50}}{100}\right)$$	Equation 3

248

$$s = -25 \frac{\log\left(\frac{100 - \tau}{\tau}\right)}{\Psi_x - \Psi_{p50}}$$	Equation 4

249

250   The xylem water potentials of *Fagus sylvatica* have a narrow distribution, with the Ψ_{p12},
251   Ψ_{p50}, and Ψ_{p88} values ranging from -2.0 MPa to -5.0 MPa (Figure 2). The two extreme curves
252   obtained from this dataset have a steep decline of hydraulic conductance with the diminution of
253   water potentials, with a small range in c_k (1.73 to 3.33) and Ψ_{p50} (-1.9 MPa to -4.7 MPa) values
254   (Table 2). *Quercus ilex* has a more negative range of xylem water potentials than *Fagus*
255   *sylvatica*, ranging from -4.97 MPa to -7.66 MPa. Therefore, the VC of *Quercus ilex* is described
256   by a Ψ_{p50} of -4.97 MPa and a c_k value of 1.06; meanwhile, its RC has a Ψ_{p50} of -7.66 MPa and a
257   c_k of 2.27.

258   The k_{max} values used by default in CLM5 are assumed constant for the different PFTs and
259   homogeneous across the different plant organs (i.e., root, xylem, and leaves). k_{max} values for each
260   plant segment can be determined based on the experimental specific hydraulic conductance (k_s,
261   kg m^{-2}MPa^{-1}s^{-1}), which is defined as the flow rate per cross sectional area per unit of pressure
262   difference along a plant segment (kg m^{-2}MPa^{-1}s^{-1}) (Eamus et al., 2016). However, a standard
263   procedure for determining k_{max} for its use in CLM5 (i.e., at PFT level and for each plant segment)
264   from tree- and plant organ-specific information existing in literature has not been specified yet.
265   Therefore, considering the large uncertainty in estimating this parameter and the unknown effect
266   of its variability, we arbitrarily choose a range of values between one order of magnitude above
267   (2.0 x 10^{-7} mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}) and below (2.0 x 10^{-9} mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}) the default value (2.0 x
268   10^{-8} mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}) of the model. The upper and lower values of this variability range are
269   referred to from now as high (Hk_{max}) and low (Lk_{max}) xylem conductance, respectively.

270

271

Figure 2. Spectrum of the vulnerability curves of *Quercus ilex* L. and *Fagus sylvatica* L. The solid blue, red, and dashed black lines represent the resistant (RC), vulnerable (VC), and default (DC) vulnerability curves used in the numerical experiments, respectively. The vulnerability curves were not differentiated between plant organs within the same tree species. The solid green lines show the full data set of vulnerability curves used for each species.

278

279   2.5. Numerical Experiments

280   The role of the intraspecific variability of plant hydraulic traits in contrasting tree species
281 was examined based on a series of numerical experiments. These experiments aimed to assess to
282 what extent the plant hydraulics representation of CLM5 reproduces the measured transpiration
283 of each experimental site based on the spectrum of vulnerability to the hydraulic failure of each
284 tree species (Table 2). All the experiments considered that the plant segments (i.e., roots, stems,
285 leaves) of a given plant functional type (PFT) have the same plant hydraulic parameterization
286 (i.e., k_{max}, Ψ_{p50}, and c_k); allowing the same degree of vulnerability to hydraulic failure between
287 the distal portions of the modeled vegetation.

288   The first set of experiments compared the effect of the PVC shape on the distribution of
289 PLC values, leaf water stress (β), and the transpiration simulated by the model. In this
290 experiment we only changed the shape of the PVC without modifying the default k_{max} value

291 among configurations. The PVC shape parameters determine the steepness of the hydraulic
292 response (c_k) and the range of water potentials at which the plant will start experiencing extreme
293 water stress (Ψ_{p50}). We hypothesized that the RC describes a plant response less affected by low
294 soil water potentials, while the VC describes a plant response with a high susceptibility to
295 hydraulic failure at low water potentials.

296   The second set of experiments explored the role of k_{max} in constraining the whole plant
297 water use strategy of the different tree species. This was achieved by changing only the k_{max}
298 value of the default configuration (DCk_{max}) to the high (Hk_{max}) and low (Lk_{max}) xylem
299 conductance while keeping the default model configuration for the shape parameters of the
300 vulnerability curve (Table 2). Finally, two additional intermediate values were added to this
301 experiment representing the midpoint between the boundaries and the default k_{max} (1.1 x 10^{-7}
302 mm$_{H2O}$ mm$_{H2O}^{-1}$s^{-1} and 1.1 x 10^{-8} mm$_{H2O}$ mm$_{H2O}^{-1}$s^{-1}) and are referred to as intermediate-high
303 (IHk_{max}) and intermediate-low (ILk_{max}) xylem conductance, respectively. The optimal choice of
304 the defined k_{max} configurations was selected based on the maximum value of the index of
305 agreement (see section 2.6.4) and is referred to in the manuscript as the 'optimal' k_{max}. For sites
306 where the index of agreement of the 'optimal' k_{max} is lower than 0.6 we used an additional k_{max}
307 configuration corresponding to the midpoint between the best performing k_{max} and the following
308 k_{max} until we achieve an index value > 0.6.

309   The third set of experiments aimed to evaluate the role of coordinated changes in safety
310 (i.e., shape parameters) and transport capacity (i.e., maximum xylem conductance). We analyzed
311 the plant hydraulic response simulated by CLM5 using the k_{max} configuration that has the largest
312 index of agreement (see section 2.6.4) obtained for each site in the second set of experiments
313 together with both Ψ_{p50} and c_k values used in the first set of experiments (Table 2). We
314 hypothesized that a more appropriate k_{max} value would lead to the expected pattern of vulnerable
315 hydraulic configurations experiencing more water stress. The stress was evaluated based on the
316 variation of the percent loss of conductance (PLC) and the leaf water stress factor (β) described
317 later in section 2.6.3.

318   2.6. Data Analysis

319   2.6.1. Reference Evaporation

320   Equation 5 is based on Equation 6 from Allen et al. (1998), and calculates the reference
321 evaporation (E_o) used as a descriptive variable of the atmospheric water demand for each
322 experimental site but has no relevance for the model simulations. Equation 5 assumed a
323 reference crop of 0.12 m height, a surface resistance of 70 s m^{-1}, and an albedo of 0.23. This
324 equation requires wind speed (u) in m s^{-1}, net radiation (R_n) and ground heat flux (G) both in MJ
325 m^{-2}d^{-1}, air temperature (T) in C, and the actual and saturated vapor pressures (e_a and e_s,
326 respectively) in kPa. G was extracted from the modeled results of the default configuration of
327 each experimental site. The slope of the saturation vapor pressure curve at air temperature (Δ,
328 kPa K^{-1}) was computed using Equation 6, based on Equation 13 from Allen et al. (1998). The
329 psychrometric constant (γ) was estimated with Equation 7 based on Equation 8 from Allen et al.
330 (1998), where λ is the latent heat of vaporization (2.45 MJ kg^{-1}), c_p is the specific heat at constant
331 pressure (1.013 x 10^{-3} MJ kg^{-1} K^{-1}), p is the atmospheric pressure (kPa), and ϵ is the molecular
332 weight ratio of water vapor and dry air (0.622).

333

$$E_o = \frac{0.408 \cdot \Delta \cdot (R_n - G) + \gamma \frac{900}{T + 273} \cdot u \cdot (e_s - e_a)}{\Delta + \gamma \cdot (1 + 0.34 \cdot u)}$$

Equation 5

334

$$\Delta = \frac{4098 \cdot \left(0.6108 \cdot exp^{\left(\frac{17.27 \cdot T}{T + 273.3}\right)}\right)}{(T + 237.3)^2}$$

Equation 6

335

$$\gamma = \frac{c_p p}{\epsilon \lambda}$$

Equation 7

336   2.6.2. Upscaled Transpiration

337   Observed forest transpiration (E_T) in mm hr^{-1} was calculated based on the hourly and sub-
338 hourly sap flux of individual trees (Q_{tree}) in cm^3 hr^{-1} available on the SAPFLUXNET data set
339 (Poyatos et al., 2020). We used equation 8 to obtain E_T and summarized it in daily time steps
340 following the recommendations of Nelson et al. (2020). Equation 8 requires Q_{tree} aggregated in
341 hourly fluxes per tree (m^3 hr^{-1}tree^{-1}), the basal tree area (Ω_{tree}) in m^2 tree^{-1}, the stand basal area
342 (Ω_{stand}) in m^2 m^{-2}, and the number of measured trees (n). All the information required in Equation
343 8 is available on the SAPFLUXNET data set for each site. The stand basal area of DE-Hin was
344 missing in the SAPFLUXNET data set, so we obtained it from Moreno et al. (2017) according to
345 the geographical location of the plot.

346

$$E_T = \frac{\Omega_{stand}}{n \cdot 10^3} \cdot \sum_{tree=1}^{n} \frac{Q_{tree}}{\Omega_{tree}}$$

Equation 8

347   2.6.3 Plant Water Stress

348   The plant water stress was evaluated by comparing the percentage loss of hydraulic
349 conductance (PLC) and the transpiration water stress parameter (β). The PLC was calculated
350 using Equation 9 at the root-stem (hereafter named stem) and stem-leaf (hereafter named leaf) plant
351 segments. This equation uses the simulated (k) and the maximum (k_{max}) plant organ conductance, where
352 low PLC values represent a stressed plant segment. The leaf water stress factor (β_x) of each
353 component (i.e., sunlit and shaded leaf) is used to down-regulate the photosynthesis and stomatal
354 conductance (D. M. Lawrence et al., 2019) and ranges from 0 (fully stressed component) to 1
355 (non-stressed component). The β_x is calculated as the ratio of the actual stomatal conductance
356 ($g_{x.s}$) over the unstressed stomatal conductance ($g_{x.max}$) (Equation 10). The canopy water stress
357 factor (β) is calculated as the weighted average of shade and sunlit components according to their
358 corresponding LAI components (Equation 11). Further details on the mathematical formulation of β
359 factor of CLM5 are provided in Kennedy et al. (2019).

360

$$PLC = 100 \cdot \left(1 - \frac{k}{k_{max}}\right)$$

Equation 9

361

$$\beta_x = \frac{g_{x.s}}{g_{x.max}}$$	Equation 10

362

$$\beta = \frac{\beta_{sunlit} \cdot LAI_{sunlit} + \beta_{shaded} \cdot LAI_{shaded}}{LAI_{sunlit} + LAI_{shaded}}$$	Equation 11

363

364   We used the modeled PLC as a proxy of realistic plant responses considering that when
365 the PLC is closer to 12%, the stomatal control of the plants prevents the trees from reducing the
366 internal water storage and dehydrating further. When the PLC is closer to 50% the hydraulic
367 stress in the plants triggers a series of negative effects such as leaf shedding or partial dieback of
368 branches, and the recovery process of the plant does not necessarily reach a full recovery.
369 Meanwhile, when the PLC is closer to 88% it is considered that most of the trees are dying or are
370 already dead (Choat et al., 2018; Johnson et al., 2012; Meinzer et al., 2009; Preisler et al., 2022).

371   2.6.4 Index of Agreement

372   The data analysis across sites and model configurations focused on comparing the
373 measured and modeled transpiration during extended summer periods lasting from May to
374 September using the Index of Agreement (Γ) proposed by Duveiller et al., (2016). Equation 11
375 determines Γ as the product between an α coefficient and the Pearson correlation coefficient (r).
376 The α coefficient represents any bias existing between measured and modeled daily transpiration
377 rates, where a value of 1.0 represents a perfect agreement between both data sets and a 0 value
378 means no agreement between them. This coefficient is determined using the standard deviation
379 of measured and modeled transpiration (σ_X and σ_Y, respectively) and their mean values ($\bar{X}$ and
380 $\bar{Y}$).

381

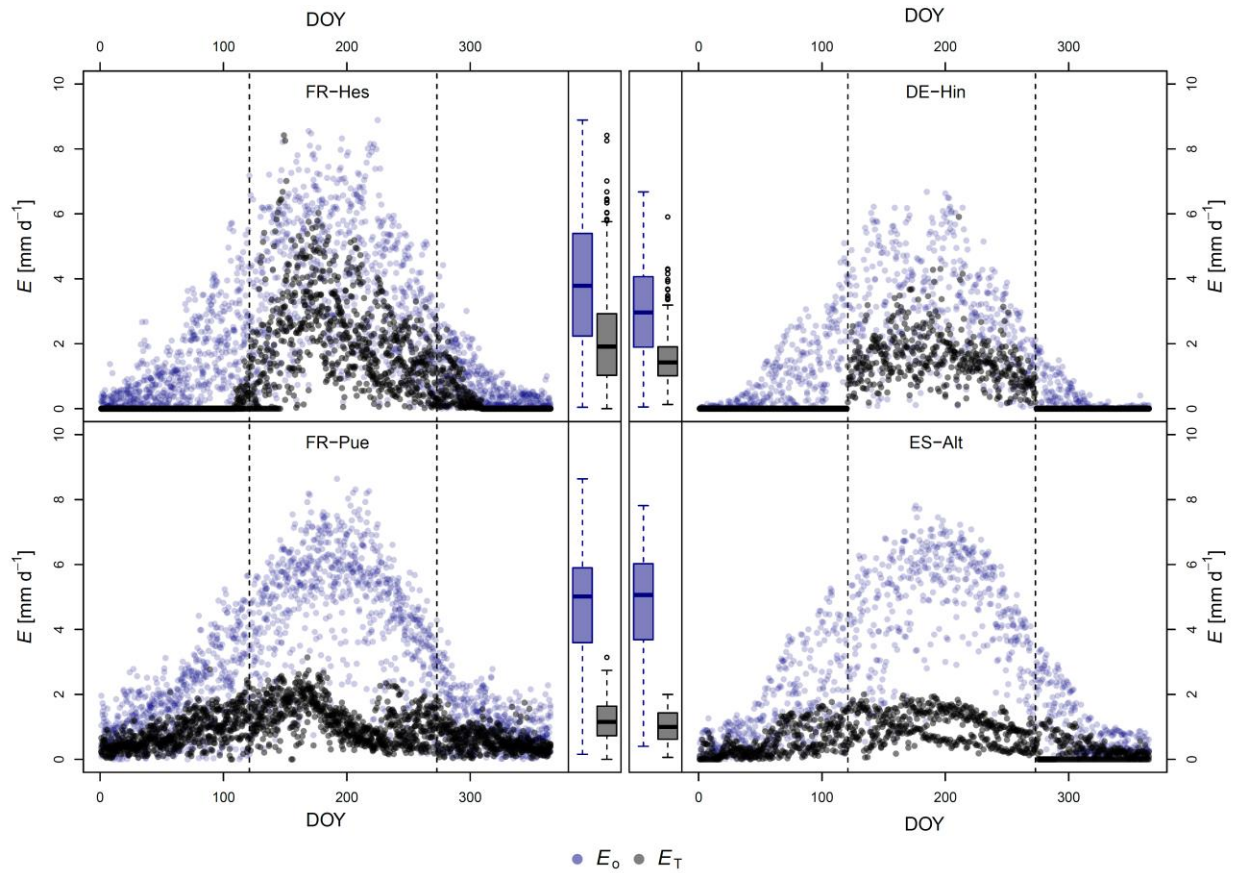| $$\Gamma = \alpha \cdot r \quad \text{where } \alpha = f(x) = \begin{cases} 0, & \text{if } r \leq 0 \\ \dfrac{2}{\left(\dfrac{\sigma_X}{\sigma_Y} + \dfrac{\sigma_Y}{\sigma_X} + \dfrac{(\bar{X} - \bar{Y})^2}{\sigma_X \cdot \sigma_Y}\right)}, & \text{otherwise} \end{cases}$$	Equation 12

382

3. Results

384   The impact of the different plant hydraulic parametrizations was investigated by
385 comparing the simulated time series of transpiration (E_{Tm}) to the upscaled sap flux measurements
386 (E_T). Furthermore, a comprehensive insight into the simulated plant hydraulic response was
387 gained by analyzing the temporal evolution and probability density of PLC, the transpiration
388 water stress parameter (β), and the water potentials across the soil-vegetation continuum (Ψ).

389   3.1. Reference Evaporation and Measured Transpiration

390   Figure 3 shows the multiannual variability of E_o and E_T for each site. During the
391 extended summer period, the atmospheric water demand is two- and four-times larger than the

392   E_T in FR-Hes/DE-Hin and FR-Pue/ES-Alt, respectively. It is worth noting that despite belonging
393   to the same climate classification, the DE-Hin and FR-Hes sites have a large difference in E_o
394   values. This difference is linked to lower temperatures at DE-Hin compared to FR-Hes.
395   E_T patterns differ among species, with sites dominated by *Fagus sylvatica* (i.e., FR-Hes and DE-
396   Hin) showing maximum E_T values of more than 4 mm d^{-1} during summer and 0 mm d^{-1} in spring
397   and autumn due to the deciduousness of the forest species. In contrast, the evergreen *Quercus*
398   *ilex* at FR-Pue and ES-Alt keep maximum E_T values lower than 4 mm d^{-1}, have smaller intra-
399   seasonal variations with greater spring and autumn E_T, but smaller values in summer compared
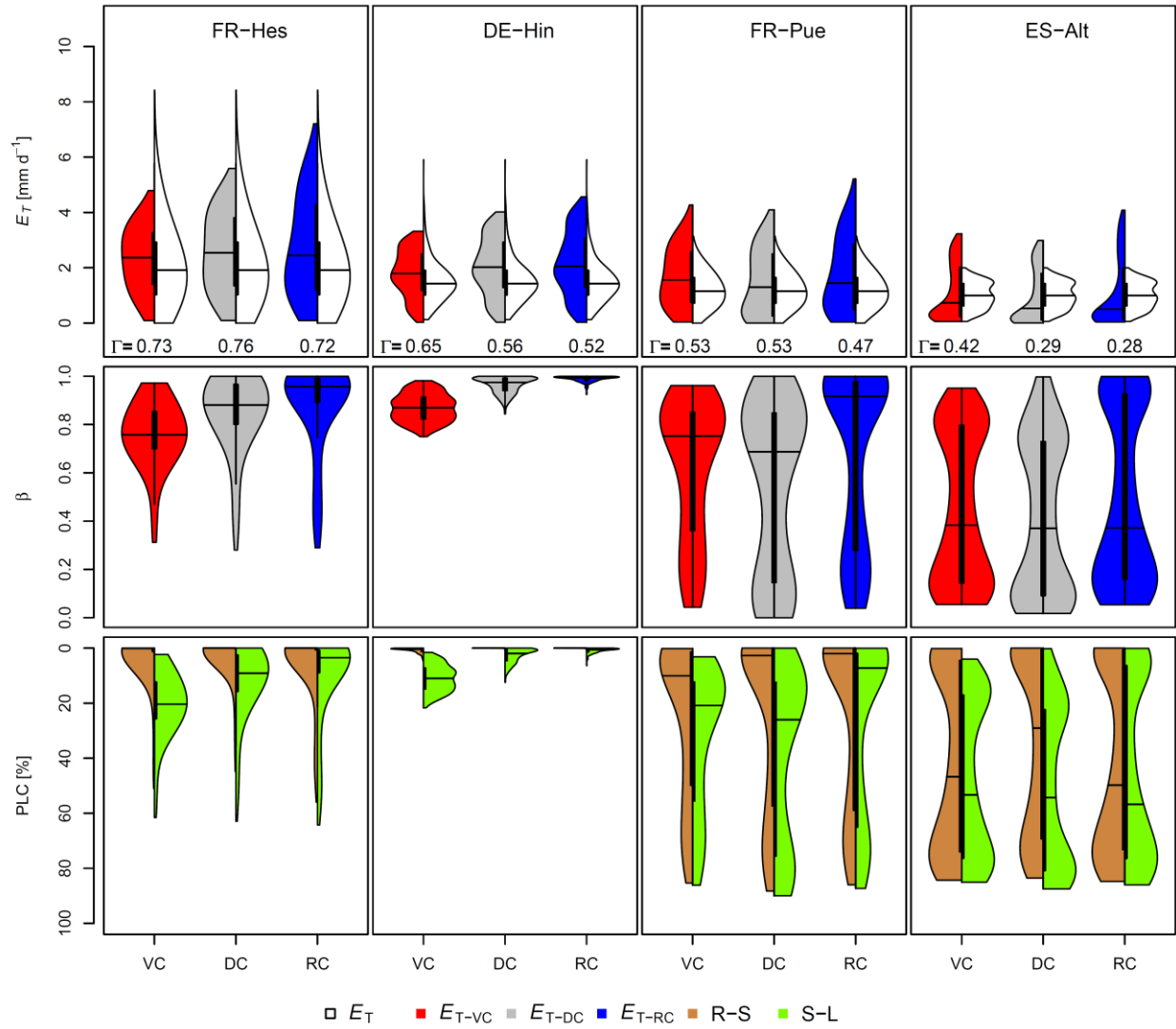400   to the *Fagus sylvatica* sites.

401

402

403   **Figure 3**. Multi-annual variation of reference evaporation (E_o) and measured daily transpiration
404   (E_T) of the four forested sites in Europe. The dashed vertical lines show the extended summer
405   period (from May to September) used for the current analysis of model results. The box plots
406   show the difference between E_o and E_T of the extended summer period for each site.

407

408   3.2. Effects of Changing the Shape of the Vulnerability Curve

409   Looking first at the deciduous sites, the experimental site FR-Hes has similar measured
410   and modeled transpiration estimates with Γ values larger than 0.7 (Figure 4), with the DC having
411   the best E_T representation. Nonetheless, the data distribution (i.e., interquartile range) of the VC
412   has a better match than the DC with the observed transpiration despite its slightly lower Γ value

413   (VC: 0.73, DC: 0.76).. The Γ values in DE-Hin are larger than 0.5 with the VC as the best
414   representation of the E_T in this site (Γ= 0.65), with minor differences during the end of the
415   extended summer period (see Figure S1). The RC and DC show a significant over-estimation of
416   E_T during May (see Figure S1). Figure 4 indicates that there are marginal differences of the PLC
417   at the root-stem segment at both sites (i.e., FR-Hes and DE-Hin) when representing a vulnerable
418   (VC) and resistant (RC) shape of the vulnerability curve. As expected, the VC tends to produce
419   lower transpiration rates and higher water stress conditions represented by low β values. These
420   stress conditions are mainly found at the stem-leaf level with the median of the PLC values going
421   beyond 12% while those at the root-stem level remain close to zero. The comparison of the
422   distribution of the PLC values at different plant levels (i.e., root-stem and stem-leaf) with those
423   of the β stress factor provides some additional insights into the relative effect of stomata and
424   plant hydraulics on the simulated transpiration response. For example, the low root-stem values
425   at FR-Hes and DE-Hin show that β is influenced by environmental stressors at leaf level, because
426   there is no strong reduction of the plant hydraulic conductance at the root-stem segment (Figure
427   4).

428   The effects of changing the shape of the vulnerability curve are remarkably different at
429   the evergreen sites (i.e., FR-Pue and ES-Alt) populated with *Quercus ilex* species (Figure 4). At
430   these sites, all configurations overestimate the transpiration response in May, which leads to a
431   strong underestimation of E_T during prolonged dry conditions of summer and followed by a slow
432   recovery in September (see Figure S1). Counterintuitively, the resistant configuration (RC) does
433   not alleviate the canopy stress as expected; meanwhile, the vulnerable configuration (VC)
434   simulates higher transpiration rates than the default configuration (DC) during most of the
435   summer. The unexpected model response is confirmed by the distribution of the simulated water
436   stress factor and PLC values, with the response of the RC and VC reflecting a higher level of
437   hydraulic failure compared to DC. The sites FR-Pue and ES-Alt show that β is partly influenced
438   by the reduction of the plant hydraulic conductance for part of the extended summer, because
439   both sites have a large period experiencing low PLC values (PLC < 50%) at the root-stem
440   compartment.

441

442

Figure 4. Distributions of daily transpiration (E_T), leaf water stress factor (β) and percentage loss of hydraulic conductance (PLC) during extended summer periods for different hydraulic vulnerability configurations. Each configuration represents a specific combination of the shape parameters (i.e., Ψ_{p50} and c_k) of the plant vulnerability curve. Each violin plot contains the multi-annual data for each site. The distribution of measured E_T is contained in the upper row plots of each site, with the observations shown on the right side of each violin plot (in dark green), and the simulations using the vulnerable (VC), default (DC), and resistant (RC) configurations shown on the left side, in red, grey, and blue, respectively. The PLC values per configuration and site are split between the plant compartments root-stem (R-S) and stem-leaf (S-L) colored in brown and lime green, respectively.
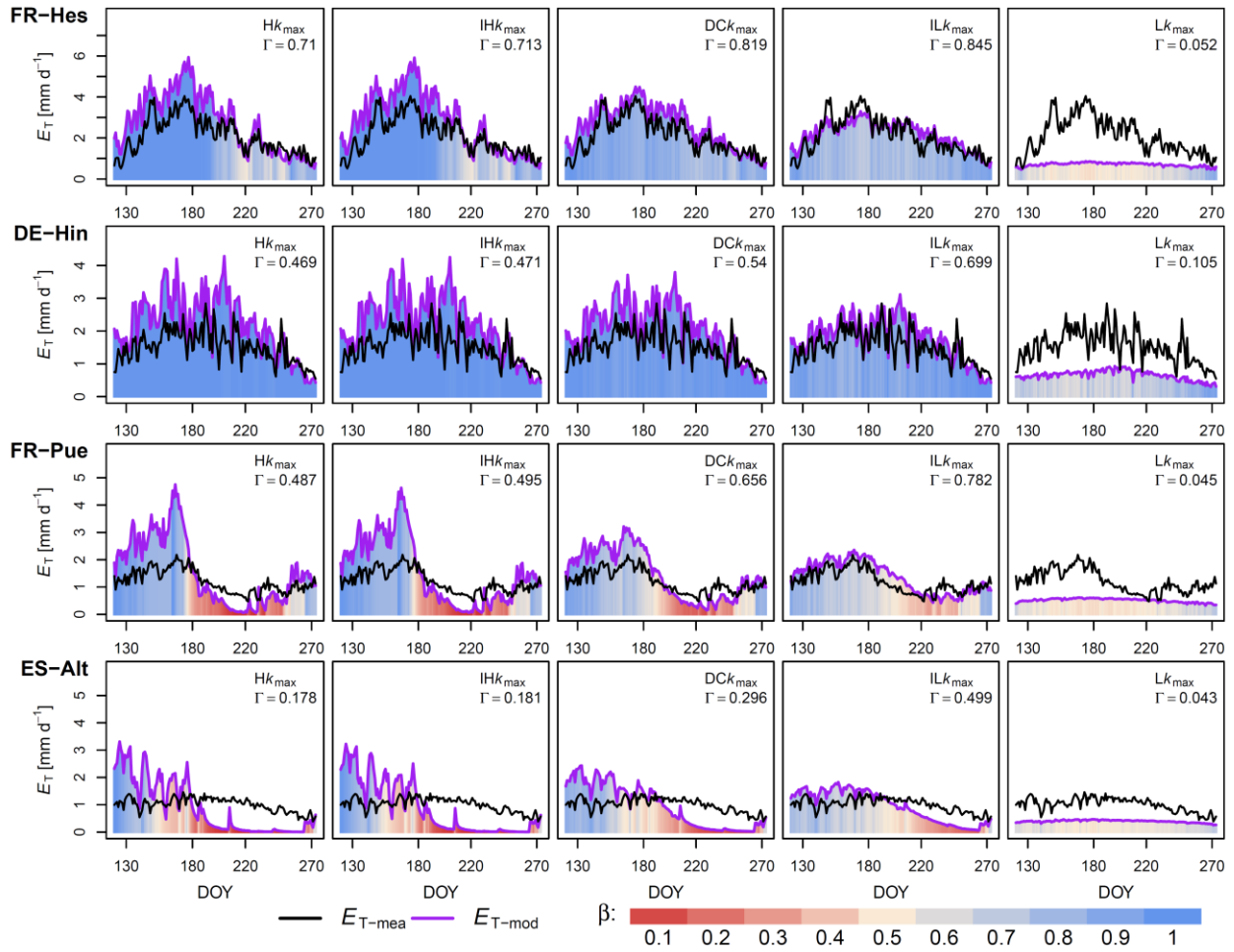
453

454   3.3. The Regulating Effects of Xylem Conductance

455   The second set of numerical experiments investigated the extent to which the water use
456 strategy can be modified in the model through gradual changes over a spectrum of k_{max} values.
457 Figure 5 illustrates how the default k_{max} configuration (DCk_{max}) has the relatively best
458 performance at FR-Hes (Γ=0.818) compared to other sites (Γ<0.7). Also, this figure shows that
459 moving from the high k_{max} (Hk_{max}) towards the low k_{max} (Lk_{max}) configuration, the simulated
460 transpiration can be 'adjusted' to better capture the differences in transpiration seasonality at the
461 four selected sites. The Hk_{max} and intermediate high k_{max} (IHk_{max}) did not show differences in the
462 simulated E_T for the selected sites, as within this range of k_{max} values transpiration rates are
463 limited by the atmospheric water demand. Most effective changes in the simulated E_T values
464 occur in the range between the default k_{max} (DCk_{max}) and low k_{max} (Lk_{max}), with the best
465 correspondence between observed sap flow and simulated transpiration rates achieved by ILk_{max}
466 for all the sites. Here it is interesting to note that at ES-Alt, the model performances can be
467 further improved by increasing the sampled k_{max} values between ILk_{max} and Lk_{max} (see Figure
468 S4), with an 'optimal' k_{max} value of 6.5 x 10^{-8} mm$_{H2O}$ mm$_{H2O}^{-1}$s^{-1}.

469   We found that gradual changes in k_{max} systematically affected the soil matric potential
470 (Ψ_{soil}) across all sites (Figure S2). This tendency shows the impact of transpiration on the soil
471 water reservoir by increasing the plant water acquisition. Higher k_{max} values allow more water to
472 be extracted from the soil and hence a reduction in the soil moisture. In contrast, reduced k_{max}
473 compared to the default value results in a reduced water transport capacity and diminishing soil
474 water acquisition. The Lk_{max} configuration restricts the plant water transport at all sites to a point
475 where the soil matric potential is close to 0 all year round (Figure S2).

476   Sites covered with *Fagus sylvatica* do not experience extreme transpiration stress (β<0.5)
477 even when the E_T is overestimated as in the Hk_{max}, IHk_{max}, and DC configurations (Figure 5).
478 The increment of leaf water stress with the Lk_{max} configuration at these two sites (i.e., FR-Hes
479 and DE-Hin) does not go beyond 0.5. This is the result of a constrained water transport within
480 the plant due to the limitation created by an extremely low k_{max}. Therefore, the stomatal
481 conductance used to determine the β values is reduced, increasing the difference between the
482 stomatal conductance (g_s) and the maximum g_s (g_{max}). The sites with *Quercus ilex* (i.e., FR-Pue
483 and ES-Alt) experience a more significant leaf water stress in summer when the k_{max}
484 overestimates the transpiration in spring (Hk_{max}, IHk_{max}, and DC). The use of smaller k_{max} values
485 at these drier sites triggers a more restricted vegetation water use under wet conditions (i.e.,
486 spring and early summer). Using a smaller k_{max} at sites with stronger dry seasons enables the
487 vegetation to not use all the water in spring, allowing the soil water reservoir to supply the
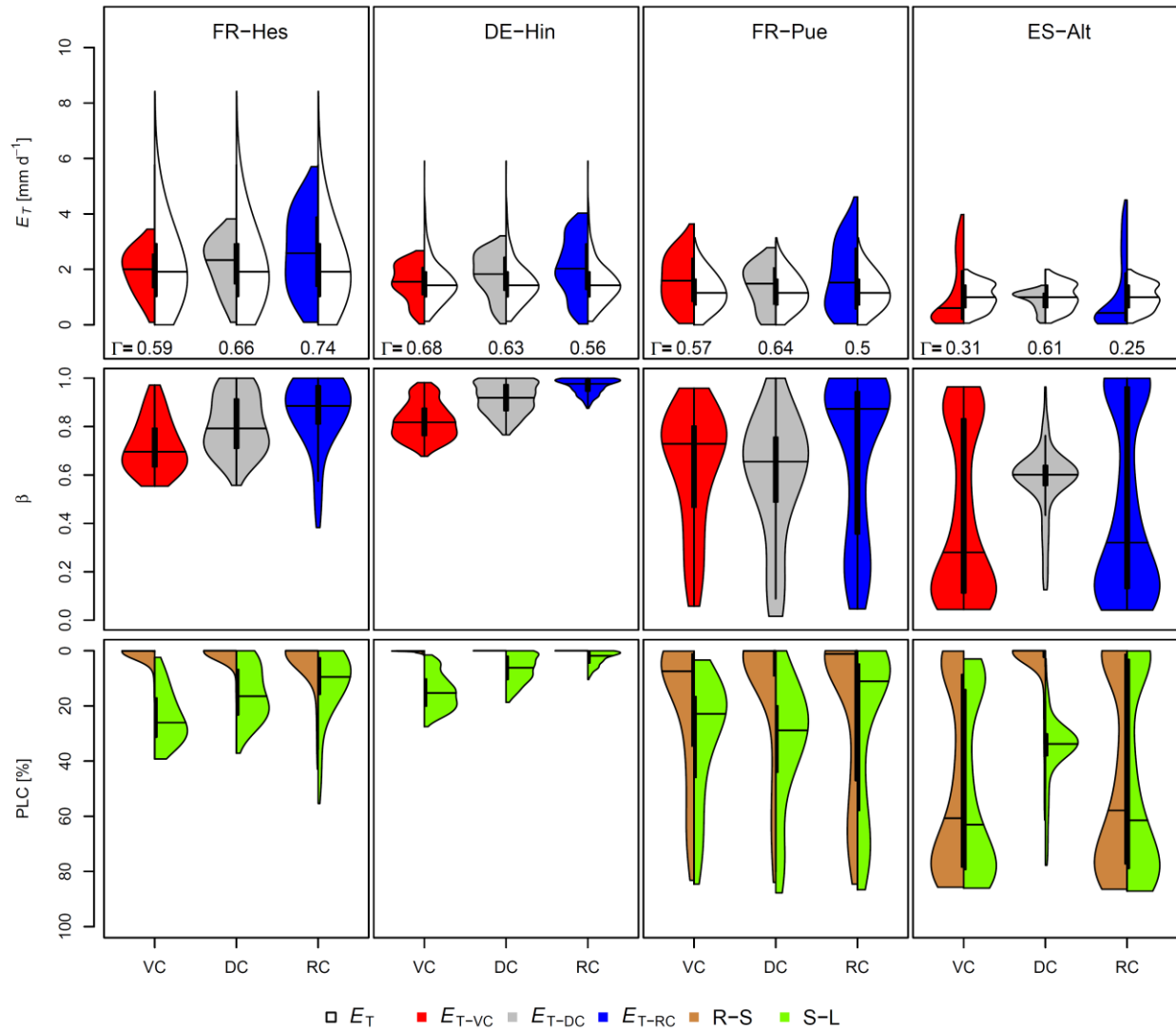488 moisture needed in summer.

489

490

Figure 5. Temporal variation of measured transpiration ($E_{T\text{-mea}}$), modelled transpiration ($E_{T\text{-mod}}$), and leaf water stress factor (β) to gradual changes of maximum xylem conductance (k_{max}) at each experimental site. The plots for each site represent a decrease of maximum xylem conductance from left (larger k_{max}) to right (low k_{max}). The leaf water stress (β) in each plot is represented by daily bars and tends towards reddish colors when β falls below 0.5 (extreme stress), while the blueish colors represent unstressed leaf conditions (β above 0.5). The index of agreement (Γ) is used for comparing the different model configurations per site.

498

3.4. The Combined Effects of Changing the Shape of the Vulnerability Curve and the Maximum Xylem Conductance

The third set of experiments was designed to evaluate the sensitivity of E_T to the plant vulnerability curve (PVC) parameters with the modified k_{max}. We expect that the use of best-fitted k_{max} values obtained in the second set of experiments allows a better evaluation of the impacts of the coordinated changes between c_k and Ψ_{p50}. As compared to Figure 4, the simulated E_T is much closer to the observed at all sites for DC, also avoiding extreme stress at xylem level (PLC < 50%) for extended periods (Figure 6). Furthermore, the coordinated changes of the shape

507   parameters with the k_{max} enable the simulation of a more realistic hydraulic response of the root-
508   stem and stem-leaf segment to dry season conditions across the four selected sites (Figure 6).
509   Similarly to the experiment 1, the results indicate that at sites populated by *Fagus sylvatica*, the
510   severe hydraulic failure events (PLC > 50%) simulated by the model are much less frequent at
511   FR-Hes and are completely absent at DE-Hin. Meanwhile, the Mediterranean sites (i.e., FR-Pue
512   and ES-Alt) are characterized by low PLC values (<20 %) for the root-stem plant segment, while
513   more severe PLC values are simulated at the stem-leaf level. Only the RC of these evergreen
514   sites shows a more severe stress response in summer, where the root-stem compartment
515   experiences PLC values larger than 20% for half of the time (second half of the violin plot of
516   Figure 6). At the same time, the xylem-leaf compartment also shows a strong reduction of
517   conductance (PLC > 50%). The impact of the RC also affects the xylem-leaf compartment,
518   where the bimodal distribution depicts the problem of the reduced provision of water for
519   vegetation during summer due to a more aggressive soil water extraction in late spring or early
520   summer.

521

522

523 **Figure 6**. Distributions of daily transpiration (E_T), leaf water stress factor (β) and percent loss of
524 conductance (PLC) during the extended summer period for different hydraulic vulnerability
525 configurations with optimal k_{max}. Each configuration represents the specific combination of the
526 shape parameters (i.e., Ψ_{p50} and c_k) of the plant vulnerability curve and the optimum k_{max} per site
527 from Experiment 2. Each violin plot contains the multiannual data for each site. The distribution
528 of measured E_T is contained in the upper plot of each site (solid dark green). The vulnerable
529 (VC), default (DC), and resistant (RC) configurations are represented by red, grey, and blue solid
530 colors, respectively. The PLC values per configuration and site are split between the plant
531 compartments root-stem (R-S) and stem-leaf (S-L) colored in brown and lime green,
532 respectively.

4. Discussion

4.1. What is known about the selected tree species?

535 *Fagus sylvatica* and *Quercus ilex* are two tree species with contrasting responses to dry
536 periods. These responses depend on physiological adaptations and the cumulative exposure to
537 specific environmental conditions that shape the WUS of each species. *Fagus sylvatica* relies on
538 a small water reservoir because of its shallow root system (Houston Durrant et al., 2016; Kirchen
539 et al., 2017; Leuschner, 2020). This has been documented at FR-Hes and DE-Hin (Granier et al.,
540 2000; Heinrich et al., 2018), where the species were found to be more susceptible to reductions
541 of soil water availability due to dry spells and droughts. This tree species prefers to grow in
542 environments with abundant precipitation, with no water stagnation on the ground or prolonged
543 dry periods (Houston Durrant et al., 2016; von Wuehlisch, 2008). This could be why *Fagus*
544 *sylvatica,* keeps significant transpiration rates as leaf water potentials decline, but it is also
545 frequently observed to shed leaves prematurely under extreme drought, which could be due to
546 reduce water loss and hydraulic failure or due to hydraulic failure (Leuschner, 2020).

547 *Quercus ilex* can grow deep roots, increasing the accessible water reservoir and allowing
548 the trees to withstand long dry periods (Peñuelas & Filella, 2003; Zapater et al., 2011), as has
549 been shown at FR-Pue and ES-Alt (Baldocchi et al., 2010; Forner et al., 2018). Its evergreen
550 character is maintained during summer thanks to its physiological adaptations such as
551 sclerophyllous leaves, summer growth reduction, and strong stomatal control (Barbeta &
552 Peñuelas, 2016; Terradas & Savé, 1992). The high wood density of oak is linked to its reduced
553 porosity, allowing it to resist lower matric potentials during summer, reducing its susceptibility
554 to hydraulic failure (Terradas & Savé, 1992). The strong stomatal control of this species
555 classifies it as the most isohydric species of the *Quercus* genus (Barbeta & Peñuelas, 2016). This
556 process is clearly visible in summer at FR-Pue and ES-Alt, where precipitation is scarce, and the
557 trees reduce transpiration rates by closing their stomata. Overall, the difference between the two
558 selected species relies on the degree of vulnerability to hydraulic failure and the WUS, with
559 *Fagus sylvatica* showing a vulnerable response and aggressive WUS, while *Quercus ilex* is more
560 resistant to hydraulic failure with a conservative WUS.

4.2. Some unexpected effects of the vulnerability curve shape parameters

562 The plant vulnerability curve (PVC) is widely used to model the plant water use response
563 to water stress from single trees up to the ecosystem scales (Kennedy et al., 2019; Li et al., 2021;
564 Mackay et al., 2015; Mencuccini et al., 2019; Sloan et al., 2021). However, linking a PVC to a
565 vegetation classification framework based on plant form and phenology, such as the PFT system,

566 introduces biases linked to the homogenization of highly diverse plant responses (Matheny et al.,
567 2017). Despite the existence of some species-specific LSM studies focused on the impact of
568 varying the PVC parameterizations for different species within the same PFT (e.g., De Kauwe et
569 al., 2022; Sabot et al., 2020), to our knowledge, the impact of characterizing the shape
570 parameters of the PVC (Ψ_{p50} and c_k) for the same species or plant functional types (PFTs) has not
571 been investigated in detail for the current implementations of the PVC in land surface models
572 (LSMs). In CLM5, the default parameterization of the plant hydraulic traits is the same for the
573 PFTs under analysis at the four experimental sites. This parameterization does not agree with the
574 current evidence showing a high degree of variability for parameters such as Ψ_{p50} (e.g., Eller et
575 al., 2020; Lu et al., 2022; Xie et al., 2023). Nonetheless, this homogeneity in the
576 parameterization of the plant hydraulic traits of CLM5 provided the opportunity to evaluate the
577 effect of the environmental conditions, namely the dynamics of atmospheric water demand and
578 soil water availability, on the simulated plant hydraulic response. FR-Hes and DE-Hin are sites
579 with a continuous water supply during summer due to the low intra-seasonal variability of
580 precipitation (Blume et al., 2022; Granier et al., 2008). Regular precipitation prevents the drying
581 out of the soil water reservoir during summer, allowing the vegetation to operate at low to
582 moderate levels of water stress throughout the year. The default plant hydraulic parameterization
583 of CLM5 reproduces an aggressive water use strategy (WUS) of *Fagus sylvatica* at FR-Hes,
584 allowing the vegetation to transpire at rates close to the atmospheric water demand. However, the
585 use of the same plant hydraulic parameterization across the selected PFTs (Table 2) does not
586 reflect the conservative WUS expected at Mediterranean sites, such as FR-Pue and ES-Alt,
587 which are inhabited by *Quercus ilex*. These two sites have a strong atmospheric water demand
588 but receive very little precipitation in summer (Allard et al., 2008; Lorenzo-Lacruz et al., 2010),
589 resulting in extremely negative soil water potentials and severe plant water stress in the default
590 model simulations (Jiménez-Rodríguez et al., 2022). Although, the inclusion of water uptake
591 from deeper soil reservoirs can also reduce the severity of simulated water stress and under-
592 estimation of transpiration rates in the model during summer periods, as shown by Jiménez-
593 Rodríguez et al. (2022), the need to regulate the extreme plant water consumption during un-
594 stressed wet periods prevailed.

595   The limitations underscored by the default plant hydraulic parameterization of CLM5 in
596 reproducing the aggressive and conservative WUS persist when changes are applied only to the
597 Ψ_{p50} and c_k parameters. That is, the model response is dominated by the instantaneous
598 atmospheric water demand and restricted by the soil water availability. Therefore, at sites where
599 water supply is continuous throughout the year (e.g., FR-Hes and DE-Hin) the decrease of Ψ_{p50}
600 with the resistant configuration (RC) allows more water to be extracted under given
601 meteorological conditions while reducing plant water stress (PLC and β) as expected (Knüver et
602 al., 2022; Walthert et al., 2021). However, the deciduous character of both sites with leaf absence
603 during the first part of the year does not prevent the RC configuration from overestimating
604 transpiration rates (E_T) for both sites (Figure 4 and Figure S1) during the first part of the
605 extended summer. This pattern of the model response illustrates the dominant role of plant
606 hydraulics over stomatal control of E_T. Under seasonally limited soil water supply, as is the case
607 at FR-Pue and ES-Alt during summer, the model simulates a counter-intuitive response when
608 changing the shape parameters of the PVC, with the resistant configuration (RC) suffering more
609 water stress and a reduced E_T than the default or vulnerable configurations (DC and VC,
610 respectively) (Figure 4). The entire intraspecific variability in PVC shape parameters for
611 *Quercus ilex* does not reproduce the conservative WUS in the model that would be expected of a

612   tree species able to withstand significant water shortage conditions (Barbeta & Peñuelas, 2016;
613   Terradas & Savé, 1992). On the contrary, the choice of more resistant PVC shape parameters
614   diminished the soil water availability simulated at FR-Pue and ES-Alt during summer due to
615   over-use of water in spring, magnifying the overall vegetation water stress.

616   4.3. Uncovering the role of maximum xylem conductance

617   The results of the second set of numerical experiments highlight the role of the maximum
618   xylem conductance (k_{max}) in determining the transpiration rates under ample water supply and
619   therefore shaping the seasonal water use strategy. Here we found that reducing the plant
620   hydraulic conductance can improve both, the over-estimation of transpiration in the early
621   growing season and the under-estimation in late summer, due to more carry-over soil resources
622   from the early to the late season. These results illustrate the effective role of k_{max} in constraining
623   the water use at sites with seasonal water limitations (i.e., FR-Pue and ES-Alt). The maximum
624   hydraulic conductance is a parameter highly influenced by plant segment age (Weithmann et al.,
625   2022) and local environmental conditions rather than genetics (Hochberg et al., 2018; Lu et al.,
626   2022). However, the complex interactions between environmental conditions and individual
627   species made it difficult to identify individual environmental drivers for temperate tree species
628   such as *Acer platanoides* L., *Carpinus betulus* L., and *Tilia cordata* Mill (Fuchs et al., 2021).
629   The response of species response of hydraulic conductance to contrasting soil conditions is not
630   fixed among species where *Fraxinus ornus* L. had shown a strong sensitivity to soil dryness
631   (Gortan et al., 2009) while other species such as *Pinus sylvestris* are not affected by such soil
632   gradients (Jackson et al., 1995). The high environmental plasticity has been documented for
633   *Fagus sylvatica* (Weithmann et al., 2022) pointing out the age of the plant segments as the
634   principal driver defining the hydraulic conductance. This characteristic is represented by the
635   range of k_{max} values observed for *Fagus sylvatica* and *Quercus ilex* (Figure S5). The observed
636   k_{max} values vary by two orders of magnitude for *Fagus sylvatica* (BDT) and five orders of
637   magnitude for *Quercus ilex* (BET), with similar maximum values for both.

638   The primary role of k_{max} for the plant hydraulic system of CLM5 is in constraining the
639   water transport during unstressed conditions and thereby determining the magnitude of plant
640   water use and how much water is left in the ground, some of which might be available later. At
641   FR-Hes, larger k_{max} values compared to the default value increase the water transport in the
642   model, allowing to match the atmospheric water demand and transpiration measurements. In
643   contrast, smaller k_{max} values are needed at ES-Alt and FR-Pue to prevent the vegetation from
644   depleting the soil water reservoir in spring and therefore enable continued plant water use under
645   moderate stress during the dry summer. A lower k_{max} depicts a transport limitation allowing to
646   reduce the water stress on the plant in the model, while a larger k_{max} allows the model to transpire
647   at higher rates, mimicking an aggressive WUS. The large influence that k_{max} has in controlling
648   the water acquisition in CLM5, despite the use of the Medlyn slope for controlling the stomatal
649   conductance, makes the model to work differently than other numerical models that rely on
650   stomatal conductance to control or mimic the WUS (Sloan et al., 2021). Therefore, in CLM5, an
651   adequate selection of k_{max} plays the role of restraining the vegetation from transpiring
652   excessively in spring and at the beginning of summer to ensure an adequate water supply as
653   summer progresses in a Mediterranean (summer-dry) climate. Note that in the Darcy's law
654   equation used in the plant hydraulics system of CLM5 a certain sensitivity in the simulated
655   transpiration fluxes could be expected by also changing the cross-sectional area of the different
656   plant segments (e.g., SAI). However, there is no direct correspondence between the prescribed

657 SAI values in the model (defined as the sum of all non-photosynthetic vegetation, including
658 stems, branches, and dead leaves (P. J. Lawrence & Chase, 2007)) and the basal area reported
659 from the selected sites (Table 1). Therefore, the uncertainty added using this variable in the
660 simulated transpiration fluxes cannot be explained by structural properties of the forest.

661   Most models describe the plant vascular factor by lumping the entire system into a single
662 term (Fatichi et al., 2016), omitting the large variability of the forest ecosystems related to tree
663 species and age (Weithmann et al., 2022). This is the case for k_{max} that is a key plant hydraulic
664 trait contributing to the control of the water transport capacity of vegetation (Eamus et al., 2016).
665 Within the plant hydraulic system of CLM5 the default k_{max} value is commonly used,
666 disregarding the variability of different tree species and stand density within the PFT
667 classification. This plant hydraulic trait (PHT) varies accordingly with the plant species,
668 environmental conditions, and tree size (Anfodillo & Olson, 2021; Domec et al., 2012; Domec
669 Jean-Christophe et al., 2008; Hochberg et al., 2018; Willigen et al., 2000), but previous studies
670 argued that k_{max} expresses the maximum xylem conductance of vegetation under the most
671 favorable environmental conditions (Sabot et al., 2020). However, to link k_{max} with the
672 experimental evidence based on measurements of specific xylem conductance (k_s) it is necessary
673 to include the forest structure (e.g., forest height, branch network) to upscale and better represent
674 the water flux within the vegetation.

675   The use of plant hydraulics in land surface modelling provides a framework to connect
676 the water stress with the stomatal response (Venturas et al., 2017), allowing a better control on
677 the simulated plant water use strategies. However, the site-specific character of k_{max} has largely
678 been ignored by the modelling community. In CLM5, the default value for k_{max} is the same for
679 all PFTs, and an order of magnitude lower than the lowest reported specific xylem conductance
680 (k_s) for *Fagus sylvatica*, whereas the reported values for *Quercus ilex* have two outliers, one and
681 three orders of magnitude below the default value (see Figure S5 for more details). The k_s values
682 of *Fagus sylvatica* do not match the range of k_{max} used in the second experiment, where the high
683 xylem conductance (Hk_{max}) is close to the lowest k_s value found for this tree species. However,
684 we found little difference in the simulations between the highest values of k_{max}, so exploring the
685 range of values where most observations lie would not improve the model simulations. The
686 differences between the reported k_s and model-default k_{max} highlights the complexity of defining
687 the k_{max} value for different plant functional types (PFTs) based on experimental data with a larger
688 number of species. The lack of detailed experimental k_s data of tree roots for different tree
689 species increases the difficulty to better understand the impact of the interaction between soil
690 type and whole plant k_{max}. The data availability constraint also extends to the continuous
691 monitoring of water potentials across and within ecosystems (Novick et al., 2022). There are
692 some studies providing discrete measurements of soil water potential (e.g., Zapater et al., 2011)
693 and predawn leaf water potentials (e.g., Lavoir et al., 2009; Peiffer et al., 2014) for the selected
694 study sites. However, the scarcity of continuous data sets does not allow the analysis of the most
695 stressful period during the day (i.e., noon) or the cumulative effect of water stress in the soil-
696 plant-atmosphere continuum. Nonetheless, the few data available for FR-Pue (i.e., Lavoir et al.,
697 2009) show a range changing from -1 MPa in early spring (i.e., DOY=90) to -5 MPa during the
698 peak of summer (i.e., DOY= 250). This range of predawn leaf water potentials agrees with the
699 leaf water potentials reproduced by the model during the same year and period (see Figure S7).
700 Here we show how important the correct parametrization of k_{max} is in CLM5 for capturing the
701 water use by vegetation in summer-dry climates. To progress, we need a better understanding of

702   how k_{max} is controlled by a complex set of growing conditions and co-ordination between the
703   root system and leaf area index (Aranda et al., 2015; Lemoine, Jacquemin, et al., 2002).

704   4.4. Understanding the impact of coordinated changes of plant hydraulic traits in CLM5

705   We found that the adjustment of the hydraulic vulnerability curve shape parameters (Ψ_{p50}
706   and c_k) alone did not enable the reproduction of the observed water use dynamics (Figure 4), as
707   choosing a parameterization that is more resistant to hydraulic failure (RC) resulted in even more
708   reduced dry season water use, if the maximum hydraulic conductance (k_{max}) was too high.
709   Indeed, the drastic effect caused by the more negative Ψ_{p50} of the RC is diminished by using a
710   smaller k_{max}, reducing the water extraction in spring, and letting the vegetation experience lower
711   PLC values in summer (Figure 6). Also, the fact that the VC of *Fagus sylvatica* results in low
712   PLC for the root-stem and more severe PLC for stem-leaf shows the model's ability to reproduce
713   important physiological processes along the PLC curve (Huber et al., 2019). These processes
714   may trigger different drought survival strategies depending on the species. For *Fagus sylvatica*,
715   water stress and loss in conductance may result in premature shedding of leaves during dry
716   conditions (Arend et al., 2022) or stomatal closure (Schuldt et al., 2016). The sites populated by
717   *Quercus ilex* are better simulated using low k_{max} values (Figures 5 and S4), which allow to better
718   reproduce the WUS of species adapted to water scarce environments (Terradas & Savé, 1992).
719   Nonetheless, extremely low Ψ_{p50} still trigger an excessive water uptake during the driest part of
720   the summer at Mediterranean sites (Figure 6), demonstrating the lack of stomatal regulation in
721   the model and its strong dependency on hydraulic limitations and soil water availability to
722   control the magnitude of E_T. Note that in our study, k_{max} was selected based on the default
723   vulnerability curve shape parameters, whereas the latter were adjusted in a second step, using the
724   previously selected k_{max}. The results could likely be improved by choosing an optimal
725   combination of k_{max}, Ψ_{p50} and c_k, but model calibration is not the goal of the present study.
726   Additionally, the fact that CLM5 does not consider the vegetation capacitance increases the need
727   to make use of the differentiation of k_{max} between plant segments, e.g. following the hydraulic
728   vulnerability segmentation hypothesis (Tyree & Ewers, 1991). This hypothesis proposes that the
729   stem should be the most resistant section of the tree, while the distal portions (i.e., roots and
730   leaves) should be the most vulnerable. However, the response of plant segmentation cannot be
731   generalized to all plant species because the effects of segmentation depends on the safety
732   modulation of other plant traits (Wilkening et al., 2023). Our study does not evaluate the effect
733   of different configurations per plant segment due to the lack of experimental data collected using
734   a consistent methodology across plant segments and tree species. Nonetheless, we expect that
735   any hydraulic differentiation between plant segments would lead to strong differences in the
736   plant water stress experienced at the root-stem segment.

737   The results of our study also demonstrate that generalizing the use of k_{max} as a
738   homogeneous parameter across PFTs in CLM5 prevents an adequate reproduction of the
739   magnitude and timing of E_T at sites in different climates. The fact that the same species (e.g.,
740   *Quercus ilex*) is not represented by the same k_{max} at different sites points out that xylem
741   conductance can be influenced by factors other than genetics (e.g., environmental conditions,
742   growth history). Also, the independence between stomatal control and hydraulic conductance in
743   the model is contradictory to what the existing evidence suggests (Franks, 2004). These aspects
744   magnify the effect that more negative Ψ_{p50} has on the water extraction when we change only the
745   curve shape parameters, something that was overlooked in previous studies (e.g., Bai et al.
746   (2021), Song et al. (2020)). The results of the coordinated changes in safety (i.e., the shape

747   parameters Ψ_{p50} and c_k) and efficiency (i.e., k_{max}) hydraulic traits underline how k_{max} rules the
748   WUS in the model, while Ψ_{p50} and c_k modulate the level of the hydraulic stress experienced. An
749   adequate parameterization of k_{max}, Ψ_{p50} and c_k in combination is critical for obtaining a simulated
750   plant hydraulic response that conforms with the plant water supply theory and the expected
751   physiological response of vegetation subjected to dry conditions. In addition, any issue linked to
752   plant hydraulic parameterization will also affect other variables such as carbon assimilation,
753   forest growth, or tree mortality, which depend on the plant water use by the forest and are highly
754   sensitive to extreme dry conditions (e.g., Joetzjer et al., 2022; Yao et al., 2022).
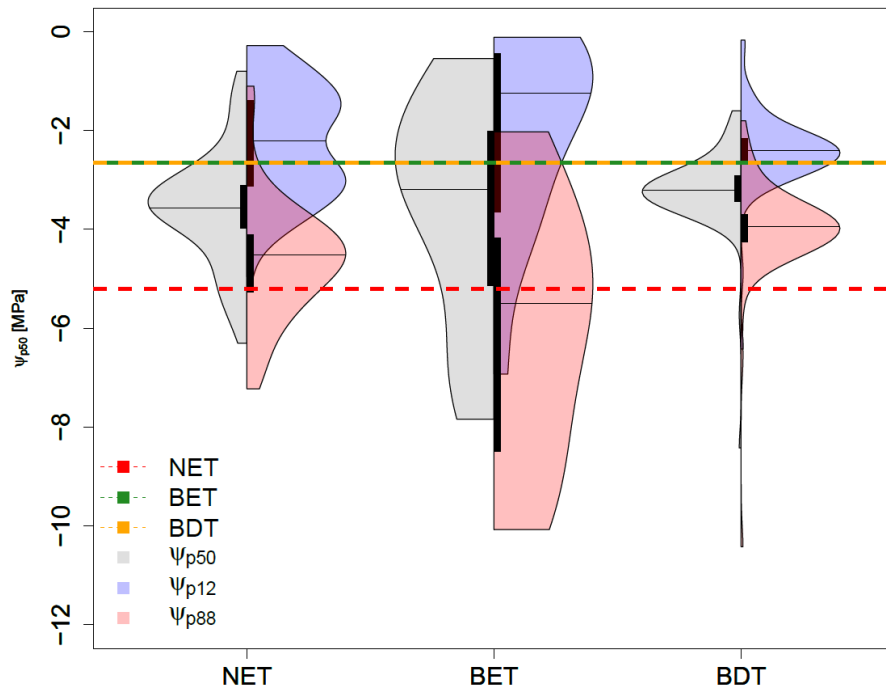
755   4.5. Addressing the plant hydraulic traits: homogeneity versus diversity

756   The plant functional type (PFT) classification system has been a valuable tool for
757   understanding drought resilience from an ecosystem perspective (Sturm et al., 2022). However,
758   the large variation in ecosystem functional properties related to the water cycle is insufficiently
759   explained by this classification system (Reichstein et al., 2014). Skelton et al. (2015) stressed the
760   need to characterize the plant response to drought by merging the current knowledge of the water
761   use strategies (WUS) with the xylem vulnerability. To fulfill this need, this classification system
762   requires the characterization of the physiological traits per vegetation type and growing stage.
763   Recent studies have shown the importance of individual tree characteristics (e.g., size, age) that
764   influence the tree water transport capacity (e.g., Bittencourt et al., 2023; Schoppach et al., 2021).
765   The combination of tree growth and environmental condition influence the overall plant
766   hydraulic performance (Fichtler & Worbes, 2012), which could explain the noted differences
767   between sites sharing the same species. The bias introduced by the heterogeneity within
768   individual tree species is enlarged by grouping per vegetation type, because the tree species
769   composition is a major driver when determining the transpiration of different vegetation types
770   (Bachofen et al., 2023). Nonetheless, these aspects are heavily homogenized using the current
771   PFT classification system, affecting the capacity to correctly predict the ecosystem water use
772   (Konings & Gentine, 2017) and leading to a poor predictive skill of the vulnerability to hydraulic
773   failure (Matheny et al., 2017).

774   *Fagus sylvatica* and *Quercus ilex* represent part of the variability of the plant hydraulic
775   traits (PHT) within the broadleaf deciduous (BDT) and broadleaf evergreen (BET) PFT classes.
776   Aiming to provide a broader context of the role of homogeneity versus diversity in plant
777   hydraulic trait studies, we sampled the XFT database (Choat et al., 2012) for a preselected set of
778   species per PFT in Europe (Buras & Menzel, 2019a; Fyllas et al., 2020; Leuschner & Meier,
779   2018) with the emphasis placed on the temperate BDT, BET, and adding the needleleaf
780   evergreen (NET) PFT to enrich the analysis (see Table S2 for details of the sampled species).
781   Figure 7 illustrates that the Ψ_{p50} used by default in CLM5 fails to capture the values of Ψ_{p50} for
782   NET and BDT in Europe, as the default values are not even close to the median values of the
783   distributions. The Ψ_{p50} of NET in CLM5 is way beyond the Ψ_{p88} for this PFT, representing an
784   extremely resistant tree with respect to the published data, while the Ψ_{p50} of BDT depicts a more
785   vulnerable tree closer to the reported median of Ψ_{p12}. The default Ψ_{p50} of BET in CLM5 matches
786   the median of the published data, but the large range showed by this PFT (-0.5 MPa to -9 MPa)
787   raises the question of how much of this variability is driven by geography or environmental
788   conditions. An even more important aspect to be considered is an adequate selection of k_{max} for
789   the PFTs, where the large variability showed for individual species does not agree with the best
790   fitted k_{max} of the model. This finding depicts the issue of considering k_{max} as constant among
791   PFTs when the variability of the species describing these PFTs is large (see Figure S5).

792   Liu et al. (2020) showed the benefit of including plant hydraulics in Earth System Models
793   improving the modeled vegetation response to climatic drivers, where the overestimation of
794   vegetation water use is a common issue when compared against the traditional processes relying
795   on soil moisture and VPD alone. They also recognize the ability of plant hydraulics to predict
796   vulnerability to droughts. In this regard, Kennedy et al. (2019) did the same for LSMs by
797   introducing the plant water stress routine in CLM5 that compartmentalized the PHT according to
798   PFT type. From an ecosystem perspective, simplifying hydraulic traits into single plant
799   functional types has additional repercussions. Matheny (2021) highlighted the importance of
800   incorporating flexible traits based on prevalent environmental stressors since tree species'
801   sensitivity to water stress is determined by their plasticity to the environment (Haberstroh &
802   Werner, 2022). This plasticity is exemplified by the different k_{max} values in *Fagus sylvatica* and
803   *Quercus ilex* in this manuscript, and the reported variability of xylem specific conductance
804   reported across many orders of magnitude for each species (e.g., Bär et al., 2018; Carevic et al.,
805   2014; Charra-Vaskou et al., 2012; Choat et al., 2012; David et al., 2007; Limousin et al., 2010;
806   Lübbe et al., 2022; Martínez-Vilalta et al., 2002; Tomasella et al., 2019). Flexible traits based on
807   the environmental stressors in CLM5 can be used by the spectrum of PVCs per PFT and
808   exploiting the role of k_{max} in regulating the WUS in the model. By adjusting the k_{max} to better
809   represent the transpiration response we may be able to identify the timing of important
810   physiological processes (e.g., leaf shedding) that differ between the VC and RC. In this way, we
811   may be able to better understand the significant changes in different ecosystem processes
812   triggered by intense dry periods (Oddi et al., 2022).

813

814 **Figure 7.** Violin plots describing the variability of the plant hydraulic traits (Ψ_{p12}, Ψ_{p50}, and Ψ_{p88})
815 within broadleaf evergreen (BET), broadleaf deciduous (BDT), and needleleaf evergreen (NET)
816 trees plant functional types (PFTs) representative of European forests (see Table S2 for the list of
817 pre-selected tree species per PFT). The red, green, and yellow dashed lines represent the NET,
818 BET, and BDT default values of Ψ_{p50} used by CLM5, respectively.

819

820   Grouping different tree species into the same PFTs in disregard of their physiological
821 adaptations may affect other parameters that also influence the plant water use within the model.
822 For example, CLM5 calculates the stomatal conductance (g_s) using the Medlyn model (Medlyn
823 et al., 2011). This model uses the minimum stomatal conductance (g_o) that is set to 100 μmol m$^-$
824 ^{2}s^{-1} in CLM5 (D. M. Lawrence et al., 2019) and a parameter that links stomatal conductance to
825 photosynthesis (g_1). These two parameters are segregated by PFT in CLM5 according to De
826 Kauwe et al. (2015). However, the g_1 values in the PFT parameterization can deviate greatly
827 from the species specific values reported by Lin et al. (2015). CLM5 uses a g_1 of 4.12 for the
828 broadleaf evergreen trees, which is much larger than the value of 1.73 reported for *Quercus ilex*.
829 This difference may induce an unconstrained transpiration at leaf level when using the default
830 configuration, compromising the coordination of root-stem-leaf conductance. Nonetheless, a
831 lower g_1 value will induce a more constrained transpiration affecting the vegetation capacity to
832 fulfill the atmospheric water requirements (see Figure S6). For the broadleaf deciduous sites (i.e.,
833 FR-Hes and DE-Hin), the differences are less pronounced, with the default g_1 value (4.45) being
834 closer to the reported value of 3.24 for *Fagus sylvatica* (Lin et al., 2015). This could be the
835 reason why E_T is over-estimated during wet conditions by the DC at FR-Pue and ES-Alt, but not
836 so much at FR-Hes and DE-Hin (Figure 5)

837   The use of PFTs in land surface modeling became a cornerstone enabling the reduction of
838 computing time and degrees of freedom thanks to the grouping of species based on plant form
839 and phenology (Colin Prentice et al., 1993). However, this grouping does not respect the large
840 variety in eco-physiological traits within a PFT, leading to strong misrepresentation of some
841 species. Therefore, a better approach for land surface modelling should target the description of
842 the vegetation according to their plant response similarities to CO_2 concentrations and
843 temperatures (Wullschleger et al., 2014), shared evolutionary and optimality principles (Franklin
844 et al., 2020), or by refining the current PFT classification based on key physiological processes
845 that determine the tree response to environmental conditions.

846

847 **5 Conclusions**

848   The intraspecific variability of the plant hydraulic traits of individual plant functional
849 types (PFT) allows to describe the spectrum of vulnerability to hydraulic failure from vulnerable
850 to resistant responses of different tree species. Understanding the importance of the right
851 selection of k_{max}, Ψ_{p50} and c_k from their large within-species variation requires a detailed
852 understanding of the role played within the model. This information is crucial for the modelling
853 community, where the parameter selection may induce considerable bias when assuming that all
854 tree species within the same PFT behave equally in different environmental conditions. The
855 adequate identification of dominant tree species per experimental site allows to narrow down the
856 variability of multiple species or by weighing the tree species contribution within the PFT but,

857   given the large variability in hydraulic traits even within a single species, a large uncertainty
858   prevails. Adjustments of the shape parameters of the hydraulic vulnerability curve (Ψ_{p50} and c_k)
859   alone do not enable the model to reproduce E_T during early summer at sites with seasonal water
860   deficits. The seasonal differences between measured and modelled transpiration illustrate the
861   importance of the maximum plant hydraulic conductance (k_{max}) for controlling the magnitude
862   and timing of E_T, i.e. the general water use strategy (WUS). A larger k_{max} allows the trees to
863   transpire larger amounts of water during favorable water conditions, quickly depleting the soil
864   water reservoir. In contrast, smaller k_{max} values limit the water transport and hence soil water
865   extraction rates, pushing the vegetation towards a more conservative WUS. Consequently, k_{max} is
866   a significant player in controlling the transpiration in CLM5 and allowing to mimic the WUS of
867   different species by limiting or enhancing the water transport. However, given the large within-
868   species variability in k_{max}, more research is needed to enable adequate parameterization of the
869   site-specific k_{max}. This work reveals the potential of plant hydraulic traits to mimic aggressive or
870   conservative WUS in CLM5, crucial for adequate reproduction of plant water use dynamics in
871   different climates. Given the large intraspecific variation in plant hydraulic traits and the
872   importance of the stand characteristics (e.g., tree height, stem area index) for limiting
873   transpiration rates in the model, a more fundamental understanding of the drivers for adjustments
874   in these parameters is needed.

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879

Open Research

881   The SAPFLUXNET data used for atmospheric forcing and transpiration estimates in the study
882   are available at ZENODO repository via https://doi.org/10.5281/zenodo.3971689 (Poyatos et al.,
883   2020) with a Creative Commons Attribution 4.0 International license for the files. The COSMO-
884   REA6 data used to complete the missing data for the atmospheric forcing in the study are
885   available from the opendata-FTP server at DWD
886   (https://opendata.dwd.de/climate_environment/REA/COSMO_REA6/) (COSMO-REA6, 2019).
887   CLM5.0 is publicly available through the Community Terrestrial System Model (CTSM) git
888   repository (Tag name: release-clm5.0.34) via https://github.com/ESCOMP/ctsm (CTSM,
889   2017/2022). The results of the numerical experiments, parameter files, and R scripts used during
890   the main data analysis of this manuscript are available at ZENODO repository via
891   https://doi.org/10.5281/zenodo.8019682 with a Creative Commons Attribution 4.0 International
892   license for the files (Jiménez-Rodríguez, 2023).

893

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895

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1490 **Table 1.** Summary of the environmental characteristics of each experimental site. All data is
1491 based on Poyatos et al. (2021) except those explicitly mentioning the source.

	DE-Hin	ES-Alt	FR-Hes	FR-Pue
Country	Germany	Spain	France	France
Site Name	Hinnensee	Alto Tajo	Hesse	Puechabon
Latitude (°)	53.33	40.802	48.674	43.741
Longitude (°)	13.192	-2.230	7.065	3.596
Elevation (m a.s.l.)	90	981	300	270
Mean Annual Precipitation (mm yr^{-1})	606.40	566.90	1003.48	1022.97
Mean Annual Temperature (°C)	8.68	11.74	9.97	13.80
Köppen-Geiger Climate Classification (Beck et al., 2018)	Cfb	Csb	Cfb	Csa
Slope (%)	2-5	5-10	0-2	0-2
Soil Texture	Sandy	n.a.	Silty	Clay-Loam
Soil Depth (cm)	n.a.	n.a.	120	52.5
Species under analysis	*Fagus sylvatica*	*Quercus ilex*	*Fagus sylvatica*	*Quercus ilex*
Stand Age (yr)	~200	59	34	58
Stand Basal Area (m^2 ha^{-1})	n.a.	13.1	19.7	28.1
Stand Height (m)	24.0	4.9	13.0	5.0
Period of Analysis	2012-2014	2012-2014	2001-2005	2001-2005

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1494 **Table 2.** Plant hydraulic parameterization (i.e., k_{max}, Ψ_{p50}, and c_k) used for the three numerical
1495 experiments carried out in each experimental site. The k_{max} used in experiment 3 was selected
1496 independently for each site after experiment 2. The plant hydraulic parametrization does not
1497 differ between plant segments (i.e., roots, stem, leaves) within the same tree species and
1498 configuration.

Experiment	Plant Hydraulic configuration	Parameter	Units	*Fagus sylvatica*	*Quercus ilex*
Experiment 1	Default Model Configuration (DC)	k_{max}	mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}		2 x 10^{-8}
		c_k	-		3.95
		Ψ_{p50}	MPa		-2.7
	Vulnerable Tree Configuration (VC)	k_{max}	mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}		2 x 10^{-8}
		c_k	-	1.73	1.06
		Ψ_{p50}	MPa	-1.9	-4.97
	Resistant Tree Configuration (RC)	k_{max}	mm$_{H2O}$ mm$_{H2O}$$^{-1}$s^{-1}		2 x 10^{-8}
		c_k	-	3.33	2.27
		Ψ_{p50}	MPa	-4.7	-7.66
Experiment 2	All configurations	c_k	-		3.95
		Ψ_{p50}	MPa		-2.7
	Hk_{max}	k_{max}	MPa		2.0 x 10^{-7}
	IHk_{max}	k_{max}	MPa		1.1 x 10^{-7}
	DCk_{max}	k_{max}	MPa		2.0 x 10^{-8}
	ILk_{max}	k_{max}	MPa		1.1 x 10^{-8}

		Lk_{max}	k_{max}	MPa		2.0×10^{-9}
Experiment 3	Default Model Configuration (DC)		k_{max}	$mm_{H2O} \, mm_{H2O}^{-1}s^{-1}$	1.1×10^{-8}	1.1×10^{-8} (FR-Pue) 6.5×10^{-8} (ES-Alt)
			c_k	-		3.95
			Ψ_{p50}	MPa		-2.7
	Vulnerable Tree Configuration (VC)		k_{max}	$mm_{H2O} \, mm_{H2O}^{-1}s^{-1}$	1.1×10^{-8}	1.1×10^{-8} (FR-Pue) 6.5×10^{-8} (ES-Alt)
			c_k	-	1.73	1.06
			Ψ_{p50}	MPa	-1.9	-4.97
	Resistant Tree Configuration (RC)		k_{max}	$mm_{H2O} \, mm_{H2O}^{-1}s^{-1}$	1.1×10^{-8}	1.1×10^{-8} (FR-Pue) 6.5×10^{-8} (ES-Alt)
			c_k	-	3.33	2.27
			Ψ_{p50}	MPa	-4.7	-7.66

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