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1 Assessing the combined effects of forest management and climate change
2 on carbon and water fluxes in European beech forests

3

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17

18 **Abstract**

19 The consequences of climate change continue to threaten European forests, particularly
20 for species located at the edges of their latitudinal and altitudinal ranges. While
21 extensively studied in Central Europe, European beech forests require further
22 investigation to understand how climate change will affect these ecosystems in
23 Mediterranean areas. Proposed silvicultural options increasingly aim at sustainable
24 management to reduce biotic and abiotic stresses and enhance these forest ecosystems'
25 resistance and resilience mechanisms. Process-based models (PBMs) can help us to
26 simulate such phenomena and capture early stress signals while considering the effect
27 of different management approaches. In this study, we focus on estimating sensitivity of
28 two state-of-the-art PBMs forest models by simulating carbon and water fluxes at the
29 stand level to assess productivity changes and feedback resulting from different climatic
30 forcings as well as different management regimes. We applied the 3D-CMCC-FEM and

31 MEDFATE forest models for carbon (C) and water (H₂O) fluxes in two sites of the Italian
32 peninsula, Cansiglio in the north and Mongiana in the south, under managed vs.
33 unmanaged scenarios and under current climate and different climatic scenarios (RCP4.5
34 and RCP8.5). To ensure confidence in the models' results, we preliminary evaluated their
35 performance in simulating C and H₂O flux in three additional beech forests of the
36 FLUXNET network along a latitudinal gradient spanning from Denmark to central Italy.
37 The 3D-CMCC-FEM model achieved R^2 values of 0.83 and 0.86 with RMSEs of 2.53 and
38 2.05 for C and H₂O fluxes, respectively. MEDFATE showed R^2 values of 0.76 and 0.69
39 with RMSEs of 2.54 and 3.01. At the Cansiglio site in northern Italy, both models simulated
40 a general increase in C and H₂O fluxes under the RCP8.5 climate scenario compared to
41 the current climate. Still, no benefit in managed plots compared to unmanaged ones, as
42 the site does not have water availability limitations, and thus, competition for water is
43 low. At the Mongiana site in southern Italy, both models predict a decrease in C and H₂O
44 fluxes and sensitivity to the different climatic forcing compared to the current climate;
45 and an increase in C and H₂O fluxes when considering specific management regimes
46 compared to unmanaged scenarios. Conversely, under unmanaged scenarios plots are
47 simulated to experience first signals of mortality prematurely due to water stress
48 (MEDFATE) and carbon starvation (3D-CMCC-FEM) scenarios. In conclusion, while
49 management interventions may be considered a viable solution for the conservation of
50 beech forests under future climate conditions at moister sites like Cansiglio, in drier sites
51 like Mongiana conservation may not lie in management interventions alone.

52

53 **Keywords:** Climate change sensitivity, *Fagus sylvatica* L., Forest management sensitivity,
54 Carbon fluxes, Water fluxes, Stress mitigation, Process-based models

55

56

57 **1. Introduction**

58 Predicting the future evolution of European forests is essential to continue to benefit from
59 the ecosystem services they provide for human well-being. Forests offer, for instance,
60 climate change mitigation through their ability to store atmospheric carbon dioxide in
61 biomass and soil (Augusto and Bo, 2022; Pan et al., 2024). In 2020, the European Green
62 Deal prioritized the vital role of forests and the forestry sector in attaining sustainability
63 objectives, such as promoting sustainable forest management, enhancing forest
64 resilience, and climate change mitigation (European Commission, 2021). Technological
65 advances and studies of forest ecosystem responses to management practices continue
66 to promote the evolution of strategies that maintain or enhance forest ecosystem
67 services, such as promoting biological diversity, water resources, soil protection, or
68 carbon sequestration (Pukkala, 2016). Different forest management systems have been
69 adopted in Europe over the years (e.g., clear-cutting or shelterwood) depending, among
70 others, on the wood product desired, the stand age, and structure (Brunet et al., 2010).
71 Forest management can be a key element in mitigating the effects of climate warming,
72 maintaining the current primary productivity and the current distribution of tree species,
73 or altering forest composition to promote more suited and productive species (Nolè et
74 al., 2015; Bosela et al., 2016). Indeed, the carbon sequestration capacity and productivity
75 of forests are dependent, primarily, on species composition, site conditions as well as on
76 stand age (Rötzer et al., 2010; Vangi et al., 2024a, b), which are affected by past and
77 present forest management activities. According to Collalti et al. (2018) and Dalmonech
78 et al. (2022), monospecific forests in Europe would appear unable to further increase
79 current rates of carbon storage and biomass production under future climate scenarios,
80 considering current management practices, but at the same time demonstrating that
81 managing under Business as Usual (BAU) practices still allows forests to accumulate
82 biomass at higher rates compared to stands left to develop undisturbed.

83 European beech (*Fagus sylvatica* L.) is an important deciduous tree species widely
84 distributed in Europe, from southern Scandinavia to Sicily and Spain to northwest Turkey
85 (Durrant et al., 2016). In Italy, according to the National Forest Inventory (INFC, 2015),
86 beech forests cover a total area of 1,053,183 hectares, accounting for about 11.7% of the
87 country's overall forested land. European beech forests demonstrate susceptibility to
88 temperature and precipitation fluctuations. For instance, a warmer environment and less
89 precipitation are forcing shifts in distribution area or the onset of loss of canopy
90 greenness (Axer et al., 2021; Noce et al., 2017, 2023; Zuccarini et al., 2023; Rezaie et al.,
91 2018). According to Skrk et al. (2023), the decline in growth of the beech forests primarily
92 occurs in the dry and warm marginal conditions prevalent near the geographical edge of
93 its distribution with a sub-Mediterranean climatic regime, posing a threat to the survival
94 of beech populations in those areas. However, tree ring analyses have also revealed an
95 unexpected increase in growth in the south Mediterranean region of Albania and
96 Macedonia beech forests at the end of the 20th century, challenging the presumed
97 decline of forest ecosystems due to drought (Tegel et al., 2014). Puchi et al. (2024)
98 additionally shed light on the susceptibility to extreme drought events of beech forests
99 found at higher latitudes compared to those found at lower latitudes in the Italian
100 peninsula by highlighting an increase, for the latter, in growth related to the abundance
101 of precipitation. In this context, it is important to minimize the uncertainty surrounding
102 the response of the carbon, water, and energy cycles within beech forest ecosystems,
103 especially as they have been shown to adapt to varying environmental drivers (De
104 Burman et al., 2024).

105 Process-based models (PBMs) are useful tools for studying forest dynamics, as well as
106 water (H₂O) and carbon (C) use efficiency, and carbon stocks as key variables of forest
107 mitigation potential (Vacchiano et al., 2012; Pilli et al., 2022; Testolin et al., 2023;
108 Morichetti et al., 2024). Forest modelling has been widely used by forest ecologists for

109 tackling numerous applied research questions, and the field is continuously evolving to
110 improve process representation to achieve higher realism and predictive capacity under
111 warmer climate and forest management scenarios (Riviere et al., 2020; Kimmins et al.,
112 2008; Nolè et al., 2013; Maréchaux et al., 2021). By comparing the predictive
113 performance of different models under current environmental conditions, it is possible
114 to gain confidence in their predictions of future trends and make informed decisions in
115 forest ecosystem management and planning processes (Huber et al., 2013; Mahnken et
116 al., 2022).

117 The main goal of the present study is to evaluate the impact of forest management
118 regimes and climate change scenarios on European beech forests using two state-of-
119 the-science PBMs: 3D-CMCC-FEM (Collalti et al., 2014) and MEDFATE (De Cáceres et al.,
120 2023). More specifically, the study aims to provide deeper insights into the carbon (C)
121 and water (H₂O) fluxes of this species under varying management practices and changing
122 environmental conditions. The MEDFATE model is capable of simulating the complex
123 water dynamics linking the soil-vegetation-atmosphere continuum. The performance of
124 MEDFATE in simulating soil moisture dynamics and plant transpiration has been
125 extensively evaluated across various scales and different stand structures, particularly in
126 Mediterranean environments (De Cáceres et al., 2015, 2021; Sánchez-Dávila et al., 2024).
127 Complementarily, the 3D-CMCC-FEM model has been extensively validated and shown
128 to effectively capture the spatial and temporal variability of carbon and water fluxes,
129 while accounting for ecological heterogeneity and integrating forest management
130 practices across a wide range of scales (Collalti et al., 2018; Dalmonech et al., 2022, 2024;
131 Mahnken et al., 2022).

132 Since the study sites vary in terms of environmental factors that can affect gross primary
133 productivity (GPP), as well as latent heat (LE), which are the two variables considered in
134 this analysis, the use of two PBMs can provide the highest reliability in capturing the

135 complex dynamics of these variables under diverse environmental conditions. By
136 leveraging the strengths of both models, we can achieve a more robust and
137 comprehensive understanding of C and H₂O across the different sites. Specifically, we
138 tested: (i) to what extent different forest management options can influence C and H₂O
139 fluxes under the present-day climate; and, (ii) how harsher climate conditions may affect
140 the C and H₂O fluxes under different management options. To gain confidence on the
141 models predictive capacity we preliminary parameterized and evaluated models'
142 performances for C and H₂O fluxes at three forest stands dominated by beech forests:
143 the Sorø (DK-Sor), Hesse (FR-Hes), and Collelongo (IT-Col) sites, which are included in
144 the PROFOUND Database (PROFOUND DB) (Reyer et al., 2020a, b) and makes part of
145 the FLUXNET Network (Pastorello et al., 2020). To address the questions, we assessed
146 the C and H₂O fluxes at two target and independent beech forest sites in Italy (Cansiglio
147 and Mongiana) by simulating their development under various management options and
148 evaluating their (model) sensitivity to current and more severe climate conditions.

149

150 **2. Material and methods**

151 **2.1 3D-CMCC-FEM model**

152 The 3D-CMCC-FEM v.5.6 ('Three-Dimensional - Coupled Model Carbon Cycle - Forest
153 Ecosystem Module') (Collalti et al., 2024, and references therein; Marconi et al., 2017;
154 Dalmonech et al., 2022, 2024; Vangi et al., 2024a, b; Morichetti et al., 2024) is a C-based,
155 eco-physiological, biogeochemical and biophysical model. The model simulates C and
156 H₂O fluxes occurring within forest ecosystems daily, monthly, or annually, depending on
157 the processes to simulate, with a common spatial scale of one hectare (Collalti et al.,
158 2016). Photosynthesis is simulated using the biochemical model of Farquhar-von
159 Caemmerer-Berry (Farquhar et al., 1980), integrating the sunlit and shaded leaves of the
160 canopy (De Pury and Farquhar, 1997). For the temperature dependence of the Michaelis-

161 Menten coefficient for Rubisco and the CO₂ compensation point without mitochondrial
162 respiration, the model adopts the parameterization described in Bernacchi et al. (2001,
163 2003). The net balance at the autotrophic level is represented by net primary production
164 in Eq. 1:

$$165 \quad \text{NPP} = \text{GPP} - R_a \quad (1)$$

166 where R_a includes both maintenance respiration (R_m) and growth respiration (R_g). When
167 R_m exceeds GPP, resulting in a negative NPP, the trees utilize their non-structural carbon
168 reserves (NSC) (i.e., soluble sugars and starch, undistinguished) to meet the carbon
169 demand (Merganičová et al., 2019; Collalti et al., 2020a). In deciduous trees, NSC is used
170 to create new leaves during the bud-burst phase, replenishing during the growing
171 season under favourable photosynthetic conditions, and finally remobilising to other
172 tissues to prepare trees for dormancy at the end of the growth phase. The model
173 assumes that NSC reserves are actively mobilized to meet metabolic demands during
174 periods of stress or carbon deficits, such as drought. For instance, during periods of
175 negative carbon balance, the model allocates stored NSC to sustain key physiological
176 processes (e.g., maintenance respiration, leaf and fine root formation). The allocation
177 scheme ensures that NSC replenishment is prioritized before supporting growth
178 demands (i.e., wood growth), consistent with evidence showing that carbon flows are
179 first directed toward restoring NSC reserves until critical thresholds are reached
180 (Hartmann & Trumbore, 2016). Replenishment of non-structural carbon reserves is
181 essential to achieve the minimum safety threshold (i.e., 11% of sapwood dry mass for
182 deciduous trees; Schwalm and Ek, 2004). Failure to meet these thresholds may trigger
183 at first remobilization from leaves and fine root and subsequently to defoliation
184 mechanisms, while complete depletion of reserves (e.g., during prolonged stress
185 periods) could lead to the death of the entire cohort of trees through carbon starvation.
186 In 3D-CMCC-FEM stomatal conductance g_s is calculated using the Jarvis equation (Jarvis,

187 1976). The equation includes a species-specific parameter $g_{s,max}$ (i.e., maximum stomatal
188 conductance) controlled by factors such as light, atmospheric CO₂ concentration, air
189 temperature, soil water content, vapour pressure deficit (VPD), and stand age (Collalti et
190 al., 2019). According to Waring and Running (2007) and Monteith and Unsworth (2008),
191 the Penman-Monteith equation is used to calculate the latent heat (LE) fluxes of
192 evaporation as a function of incoming radiation, VPD, and conductances at a daily scale,
193 summing up the canopy, soil, and snow (if any) latent heat flux expressed as $W \cdot m^{-2}$ or
194 $MJ \cdot m^{-2} \cdot time^{-1}$.

195 The 3D-CMCC-FEM accounts for forest stand dynamics, including growth, competition
196 for light, and tree mortality under different climatic conditions, considering both the CO₂
197 fertilization effects and temperature acclimation (Collalti et al., 2018, 2019; Kattge and
198 Knorr, 2007). Several mortality routines are considered in the model, such as age-
199 dependent mortality, background mortality (stochastic mortality), self-thinning
200 mortality, and the aforementioned mortality due to carbon starvation. In addition to
201 mortality, biomass removal in 3D-CMCC-FEM results from forest management
202 practices, such as thinning and final harvest (Collalti et al., 2018; Dalmonech et al., 2022;
203 Testolin et al., 2023). The required model input data include stand age, average DBH
204 (Diameter at Breast Height), stand density, and tree height (Collalti et al., 2014). The soil
205 compartment is represented using one single bucket layer, in which the available soil
206 water (ASW, in mm) is updated every day considering the water inflows (precipitation
207 and, if provided, irrigation) and outflows (evapotranspiration, i.e., the sum of evaporation
208 from the soil and transpiration of the canopy). The remaining water between these two
209 opposite (in sign) fluxes that exceeds the site-specific soil water holding capacity is
210 considered lost as runoff. For a full 3D-CMCC-FEM description, refer to Collalti et al.
211 (2024).

212

213 2.2 MEDFATE model

214 MEDFATE v.4.2.0 is an R-based modelling framework that allows the simulation of the
215 function of forest ecosystems, with a specific emphasis on drought impacts under
216 Mediterranean conditions (De Cáceres et al., 2021, 2023). MEDFATE calculates energy
217 balance, photosynthesis, stomatal regulation, and plant transpiration of gas exchange
218 separately at sub-daily steps. Like 3D-CMCC-FEM, MEDFATE also simulates
219 photosynthesis at the leaf level using the biochemical model of Farquhar-von
220 Caemmerer-Berry (Farquhar et al., 1980) for sunlit and shaded leaves (De Pury and
221 Farquhar, 1997). MEDFATE can simulate plant hydraulics and stomatal regulation
222 according to two different approaches: (a) steady-state plant hydraulics and optimality-
223 based stomatal regulation (Sperry et al., 1998; Sperry et al., 2017); and (b) transient plant
224 hydraulics including water compartments and empirical stomatal regulation (Sureau-
225 ECOS; Ruffault et al., 2022). In this work, we took the second approach, i.e., Sureau-
226 ECOS (Ruffault et al., 2022).

227 The hydraulic architecture of the Sureau-ECOS module comprises arbitrary soil layers,
228 where the rhizosphere containing coarse and fine root biomass is calculated for each
229 layer. The total root xylem conductance is determined by factors such as root length
230 (limited by soil depth), weight, and distribution across the different layers. In addition,
231 the resistance to water flow is dependent on two plant compartments (leaf and stem,
232 each composed of symplasm and apoplasm). Overall, plant conductance is defined by
233 the sum of resistances across the hydraulic network (i.e., soil, stem, and leaves), taking
234 into account processes such as plant capacitance effects (i.e., the variation of
235 symplasmic water reservoirs in the stem and leaves) and cavitation flows (i.e., water
236 released to the streamflow from cavitated cells to non-cavitated cells during cavitation)
237 (Hölttä et al., 2009). To withstand drought stress, adjacent conduits (tracheids or
238 vessels) and/or living cells (e.g., parenchyma) release water to the xylem and may

239 subsequently be refilled. In the event of embolization, cavitated xylem conduits release
 240 their water to the non-cavitated parts of the xylem, which then transfer it to adjacent
 241 compartments. Each element (roots, stem, leaves) of the hydraulic network has a
 242 vulnerability curve k_{ψ} , that declines as water pressure becomes more negative. The
 243 xylem vulnerability curve is modelled using a sigmoid function, defined by the equation:

$$244 \quad k_{\psi} = \frac{k_{\max}}{1 + \exp\left(\frac{\text{slope}}{25}\left(\frac{\psi}{\psi_{50}}\right)\right)} \quad (2)$$

246 where k_{\max} is the maximum hydraulic conductance, ψ_{50} is the water potential
 247 corresponding to 50% of conductance, and "slope" is the slope of the curve at that point.
 248 The stem vulnerability curve can be used to determine the proportion of stem
 249 conductance loss (PLC_{stem}) associated with vessel embolism. This embolism reduces
 250 overall tree transpiration and photosynthesis. Plant hydraulic failure and tree death can
 251 occur if the PLC_{stem} exceeds the 50% threshold.

252 Gas exchange in the Sureau-ECOS module depends on stomatal conductance (which
 253 depends on light, water availability, and air temperature) and leaf cuticular conductance,
 254 which changes with leaf temperature due to changes in the permeability of the
 255 epidermis. Stomatal regulation, unlike the 3D-CMCC-FEM, follows the Baldocchi (1994)
 256 approach, which allows coupling leaf photosynthesis with water losses. In addition, a
 257 multiplicative factor depending on leaf water potential is used to decrease stomatal
 258 conductance under drought conditions, following a sigmoidal function similar to stem
 259 vulnerability.

260 Soil water balance is computed daily. MEDFATE can consider an arbitrary number of soil
 261 layers with varying depths in which the water movement within the soil follows a dual-
 262 permeability model (Jarvis et al., 1991; Larsbo et al., 2005). Soil water content (ΔV_{soil} , in
 263 mm) is calculated taking into account variables such as infiltration, capillarity rise, deep

264 drainage, saturation effect, evaporation from the soil surface, transpiration of the
265 herbaceous plant, and woody plant water uptake. A full MEDFATE description is
266 available at <https://emf-creaf.github.io/medfatebook/index.html>.

267

268 **2.3 Evaluation sites**

269 Model evaluation was performed in three PROFOUND and FLUXNET Network European
270 beech sites, i.e., Sorø (DK-Sor, Denmark), Hesse (FR-Hes, France), and Collelongo (IT-
271 Col, Italy), in which we retrieved information on soil texture, soil depth, and stand
272 inventory data of forest structure for model initialization (Collalti et al., 2016; Marconi et
273 al., 2017; Reyer et al., 2020a, b; <https://fluxnet.org/>). These sites are equipped with the
274 Eddy Covariance towers (EC; Pastorello et al., 2020) for long-term continuous
275 monitoring of atmospheric carbon, water, and energy fluxes of the forests (Fig. 1). The
276 DK-Sor site is located in the forest Lille Bogeskov on the island of Zealand in Denmark.
277 FR-Hes is situated in the northeastern region of France and lies on the plain at the base
278 of the Vosges Mountains. IT-Col (Selva Piana stand) is a permanent experimental plot
279 installed in 1991 and situated in a mountainous area of the Abruzzo region, centre of Italy.
280 The pedological characterization of soils exhibits distinct variations across the studied
281 sites. The soil at the DK-Sor site is predominantly classified as either Alfisols or Mollisols.
282 The FR-Hes site showcases an intermediary nature, displaying characteristics akin to
283 both luvisols and stannic luvisols. At the IT-Col site, the prevailing soil type is identified
284 as Humic alisols, according to the USDA soil classification system. Full details of these
285 sites are reported in Table 1.

286 The variables accounted for in the evaluation were obtained from the Fluxdata website
287 (<http://fluxnet.fluxdata.org/data/fluxnet2015-dataset/>) from the FLUXNET2015
288 database (Pastorello et al., 2020). The variables considered are the daily GPP, estimated
289 from Net Ecosystem Exchange (NEE) measurements and quality checked using the
290 constant USTAR turbulence correction according to Papale et al. (2006) and the Latent

291 Heat flux (LE) with energy balance closure correction (i.e., 'LE_CORR') (Pastorello et al.,
292 2020).

293

294 **2.4 Study sites**

295 The two target sites considered in this study are Cansiglio and Mongiana Forests (Fig. 1)
296 (De Cinti et al., 2016). Each site consists of nine long-term monitored plots of differently
297 managed beech stands, with a spatial extension for each area above 3 ha, for about 27
298 ha of the experimental area. Three different silvicultural treatments were applied (see
299 Figs. S1-S2). For each site, three of the nine plots considered were left unmanaged (i.e.,
300 no cutting and leaving the stands to natural development), defined as 'Control' plots,
301 three plots were managed following the historical shelterwood system ('Traditional'),
302 and three with innovative cutting ('Innovative'). In Cansiglio, considering the
303 developmental stage of the stand was an establishment cut to open growing space in
304 the canopy for the establishment of regeneration. The 'Innovative' cutting consisted of
305 selecting a non-fixed number of scattered, well-shaped trees (the 'candidate trees') and
306 a thinning of neighbouring competitors to reduce competition and promote better
307 growth. In Mongiana, 'Traditional' silvicultural treatment was the first preparatory cut to
308 increase the vitality and health of the intended residual trees in the stand. The 'Innovative'
309 option was the identification of 45–50 as 'candidate trees' per hectare and removing only
310 direct competitors.

311 The Cansiglio site is situated in a mountainous area in the Veneto region, northern Italy.

312 Mongiana site is located in a mountainous area in the Calabria region of southern Italy.

313 The latter shows higher mean annual temperature (MAT, C°) and lower mean annual
314 precipitation (MAP, mm·year⁻¹) (i.e., drier conditions) than the Cansiglio site located at
315 higher latitudes (Table 1). Data on forest structure and soil texture were collected during
316 the field campaigns conducted in 2011 and 2019 (Cansiglio) and in 2012 and 2019

317 (Mongiana). At the Cansiglio site, soils are identified as Haplic luvisols, whereas at
318 Mongiana, the predominant soil classifications consist of Inceptisols and Entisols,
319 according to the USDA soil classification system. The variables analyzed in these sites,
320 like in the evaluation sites, were GPP and LE. A summary of these sites is reported in
321 Table 1.

322

323 **2.5 Meteorological data**

324 For the evaluation sites (i.e., DK-Sor, FR-Hes, IT-Col) observed meteorological data were
325 retrieved from the harmonized PROFOUND database (Reyer et al., 2020a, b) and
326 FLUXNET2015 database (Pastorello et al., 2020; <https://data.icos-cp.eu/>).

327 For the Mongiana and Cansiglio sites, meteorological data for 2010–2022 were obtained
328 at daily temporal resolution from the relevant region's Regional Environmental Protection
329 Italian Agencies (ARPAs), which are responsible for monitoring climate variables with
330 weather stations. The choice of thermo-pluviometric weather station was based on the
331 minimum distance from the study area (between 2 and 9 km away from the study sites,
332 respectively) and on the data availability and integration with other weather stations in
333 the proximity, representing the best available and obtainable meteorological observed
334 data for these sites. The Bagnouls–Gausson graph (Fig. S3) shows the mean monthly
335 precipitation (mm) and air temperature (°C) recorded for every station inside the
336 catchment.

337 Climate scenarios used as inputs for the two models at the Cansiglio and Mongiana sites
338 were from the COSMO-CLM simulation at a spatial resolution of approximately 2.2 km
339 over Italy (Raffa et al., 2023).

340 The daily variables considered for 3D-CMCC-FEM were mean solar radiation ($\text{MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$),
341 maximum and minimum air temperature (°C), precipitation ($\text{mm}\cdot\text{day}^{-1}$), and the
342 mean relative air humidity (%). In contrast, the MEDFATE model uses mean solar

343 radiation, maximum and minimum air temperature, precipitation, the daily maximum
344 and minimum relative air humidity, and wind speed ($\text{m}\cdot\text{s}^{-1}$).

345

346 **2.6 Modelling set-up**

347 A set of parameters specific for *Fagus sylvatica* L. was provided as input to the model
348 3D-CMCC-FEM as described in Collalti et al. (2023) while for MEDFATE as in De Cáceres
349 et al. (2023). To remove any confounding factors related to parameterization, the
350 parameters related to photosynthesis and stomatal conductance were kept constant
351 between the two models (see Table 2). A complete list of parameters and their values
352 for both models as adopted in the present study can be found in supplementary
353 materials (Table S2).

354 We then used the LAI and Available Soil Water (AWS) values obtained from the 3D
355 CMCC-FEM outputs as input for running simulations with the MEDFATE model given that
356 the model function used here does not prognostically simulate LAI and Available Soil
357 Water (ASW). Precisely, here we used MEDFATE to simulate C and H_2O fluxes only while
358 considering plant hydraulics (De Cáceres et al., 2021), from the forest structure
359 predicted, in terms of LAI, by 3D-CMCC-FEM. For MEDFATE water balance, LAI values
360 determine the competition for light and also drive the competition for soil water, along
361 with the root distribution across soil layers.

362

363 **2.7 Model evaluation**

364 To evaluate the performance of the two models across the different sites examined and
365 under varying environmental conditions, we first assessed GPP and LE daily fluxes along
366 a latitudinal gradient at sites equipped with EC towers. The reliability of the two models
367 subsequently allowed us to coherently simulate fluxes at sites where EC towers were
368 absent. Both models were run for five years on the evaluation sites, with the simulation
369 period determined by the availability of observed data provided, as already mentioned,

370 from the PROFOUND database, specifically, from 2006 to 2010 at DK-Sor and IT-Col
371 sites while for FR-Hes starting from 2014 to 2018. The performance metrics of the results
372 of the evaluation for each site for the GPP and LE variables were the coefficient of
373 determination (R^2), Root Mean Square Error (RMSE), and Mean Absolute Error (MAE).

374

375 **2.8 Model application in managed sites**

376 In the managed sites (i.e., Cansiglio and Mongiana), simulations were performed using
377 Historical climate ('Hist') and, to analyse models' sensitivities to climate change, under
378 two Representative Concentration Pathways 4.5 and 8.5 ('Moderate' and 'Hot Climate'),
379 respectively. The 'Hist' climate was used to run simulations at the Cansiglio site from 2011
380 to 2022 and the Mongiana site from 2012 to 2022. In contrast, simulations using RCP4.5
381 and RCP8.5 climate ran accounting for the same period, that is, eleven years for the
382 Cansiglio site and ten years for the Mongiana site, but considering the last years of the
383 climate change scenarios (i.e., 2059–2070 and 2060–2070, respectively) to create
384 harsher temperature and precipitation conditions, but with an increased atmospheric
385 CO₂ concentration (in $\mu\text{mol}\cdot\text{mol}^{-1}$).

386 For each of the nine sampled areas, in the Cansiglio and Mongiana sites, we considered
387 a representative area of one hectare for each type of plot: 'Control', 'Traditional', and
388 'Innovative'. At the beginning of the simulations, each site thus included a total of 9 plots,
389 each one hectare in size—comprising three 'Control' plots, three 'Traditional' plots, and
390 three 'Innovative' plots. This setup resulted in a total of nine hectares being simulated
391 per site where the model 3D-CMCC-FEM removed a certain percentage of the Basal
392 Area (BA) according to the LIFE-ManFor project (see Table S1). 'Traditional' and
393 'Innovative' cutting took place for the first time in 2012 (Cansiglio) and 2013 (Mongiana),
394 respectively. Following preliminary results, since the Mongiana site experienced a lighter
395 thinning intensity compared to the Cansiglio site (refer to Table S1), consequently, for the

396 Mongiana site, we considered an alternative management option involving the removal
397 of 40% of the BA. This was done to evaluate whether a more intensive management
398 approach ('SM') could have influenced models' results on GPP and LE fluxes related to
399 the reduction in competition and enhanced water availability.

400

401 3. Results

402 3.1 Model evaluation

403

404 The GPP at DK-Sor, FR-Hes, and IT-Col sites estimated from EC and simulated by 3D-
405 CMCC-FEM and MEDFATE are shown in Fig. 2. At the DK-Sor site, the 3D-CMCC-FEM
406 simulates a mean daily GPP of $5.14 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$, while MEDFATE $5.13 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; and
407 EC $5.54 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; at the FR-Hes site, 3D-CMCC-FEM mean daily GPP of 6.18
408 $\text{gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$ compared to MEDFATE $4.82 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$, and EC $4.99 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; lastly
409 at the IT-Col site, 3D-CMCC-FEM mean daily GPP of $4.88 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$ compared to
410 MEDFATE $4.19 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; and EC $4.11 \text{ gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$. Additionally, at the DK-Sor site,
411 the 3D-CMCC-FEM simulated a mean daily LE of $2.83 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$, while the MEDFATE
412 simulated a mean value of $2.22 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; and EC $3.19 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; at the FR-Hes
413 site, 3D-CMCC-FEM simulated a mean daily LE of $4.16 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$ compared to
414 MEDFATE $3.01 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; and EC $4.47 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; in the end at the IT-Col site, 3D-
415 CMCC-FEM simulated a mean daily LE of $2.02 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$ compared to MEDFATE 2.57
416 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$; while EC $3.93 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$. The GPP predicted by 3D-CMCC-FEM has
417 shown higher values of R^2 (0.92) at DK-Sor and the lowest value at FR-Hes site ($R^2 = 0.76$)
418 whilst a value of $R^2 = 0.83$ at IT-Col site, respectively. For the MEDFATE model, the
419 highest predicted GPP value of R^2 (0.85) was at DK-Sor, the lowest ($R^2 = 0.68$) at IT-Col,
420 and at FR-Hes $R^2 = 0.76$, the same showed for the 3D-CMCC-FEM model, respectively.
421 Differently, the highest R^2 (0.89) value for 3D-CMCC-FEM considering LE predicted vs.

422 observed was at FR-Hes site and almost the same values for DK-Sor and IT-Col sites (R^2
423 = 0.85 and 0.84, respectively). MEDFATE, for predicted vs. observed LE variable, has
424 shown the highest R^2 (0.77) at IT-Col site, lower R^2 (0.69) value at FR-Hes site and the
425 lowest R^2 (0.62) value at DK-Sor site, respectively. In general, both the Root Mean Square
426 Error (RMSE) and Mean Absolute Error (MAE) values in all sites were reasonably low,
427 falling within the ranges of 3.31 to 2.02 $\text{gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$ and 2.46 to 1.47 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$, for
428 both models and for both the variables. In Fig. 3 and Table 3 the summary of the
429 evaluation metrics performance results.

430

431

432 3.2 Simulation results at Cansiglio

433

434 Fig. 4 shows the simulation results using the 3D-CMCC-FEM and MEDFATE models in
435 the Cansiglio site. For the 3D-CMCC-FEM, the 'Control' plot exhibited the lowest GPP
436 values under 'Hist' climate conditions, averaging 1,681 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. These values
437 increased slightly to 1,982 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under the RCP4.5 climate and further rose to
438 2,204 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under the RCP8.5 climate. Similarly, for plots managed with
439 'Traditional' methods, the trends were consistent with the 'Control' plot, showing
440 average GPP values of 1,603, 1,942, and 2,141 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under 'Hist', RCP4.5 and
441 RCP8.5 climate, respectively. However, 'Innovative' management showed lower GPP
442 fluxes across all three climate scenarios, with average values of 1,534, 1,882, and 2,075
443 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under 'Hist' RCP4.5 and RCP8.5 climate, respectively. The MEDFATE model
444 showed higher mean absolute GPP increases than the 3D-CMCC-FEM model under
445 RCP4.5 and RCP8.5 climates, respectively. Under the 'Hist' climate and all treatments,
446 the mean GPP values were about 1,638 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$, whereas under the RCP4.5 climate,
447 they rose to 2,516 and 2,995 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under RCP8.5 climate. Analyzing in Fig. 4 the
448 trends of LE for the 3D-CMCC-FEM model, these trends closely follow those of GPP

449 concerning management treatments. The 3D-CMCC-FEM LE values for the 'Control'
450 plots, similar to GPP, were lowest for the 'Hist' climate with an average value over the
451 simulation years of 1,200 and 1,501 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ for the RCP4.5 climate, and 1,391
452 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ for the RCP8.5 climate, respectively. The LE of the 'Traditional'
453 management predicts values of 1,129 in the 'Hist' climate, 1,440 in the RCP4.5 climate, and
454 1,338 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ in the RCP8.5 climate, respectively. For the 'Innovative' management,
455 the mean LE values were 1,121 in the 'Hist' climate, 1,403 for the RCP4.5 climate, and 1,306
456 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ for the RCP8.5 climate, respectively. Similar to the GPP fluxes, the
457 MEDFATE model simulated reductions in LE fluxes among the treatments and higher
458 values across the climates. The mean LE value modelled in the 'Hist' climate, grouped by
459 treatments (because of slight differences among managements), was about 920, 1,419 in
460 the RCP4.5 climate, and 1,456 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ in the RCP8.5 climate, respectively (Fig. S12).
461 MEDFATE simulated a stem xylem conductance loss of approximately 40% in the
462 seventh, eighth, and twelfth years of simulation for the RCP8.5 climate scenario in the
463 'Control' plot. In contrast, this loss was predicted only in the seventh year for the
464 managed plots. Conversely, near-zero or negligible stem embolism were simulated
465 under the 'Hist' and RCP4.5 climate scenarios. The 3D-CMCC-FEM simulated higher
466 values, albeit in a small percentage (i.e., between 8-10%) of NSC, increasing
467 proportionally to the intensity of basal area removed, better observable in the graph at
468 the tree level.

469

470 3.3 Simulation results at Mongiana

471 Simulation results at the (drier) Mongiana site well depicted the differences with the rainy
472 Cansiglio site (Fig. 5). The 3D-CMCC-FEM model showed no significant differences in the
473 mean values of GPP among various management interventions under 'Hist' climate
474 conditions, with a mean value of 2,151 $\text{gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. Compared to the Cansiglio site,

475 Mongiana exhibited lower average GPP values. Under RCP4.5 climate conditions, the GPP
476 for the 'Control' plot was $1,864 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. In contrast, under the current climate, the
477 'Traditional' and 'Innovative' management interventions yielded higher average GPP
478 values of 2,115 and $2,086 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. The GPP values under the more intensive
479 management ('SM') and RCP4.5 climate decreased even further than those of the
480 'Control' plot, with an average value of $1,650 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. Under the RCP8.5, no
481 differences in GPP were observed among management strategies, with values of about
482 $1,525 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$. Moreover, under the RCP8.5, the 'Control' plot experienced
483 complete mortality after five years of simulations. The MEDFATE model predicted
484 slightly higher average values of GPP in the 'Control' plots ($2,099 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$)
485 compared to the managed plots ($2,087 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$, encompassing both 'Traditional'
486 and 'Innovative' of two management strategies), with no significant differences observed
487 among the management strategies and under 'Hist' climate. Under the RCP4.5 and
488 RCP8.5, the GPP values were 1,608 and $1,935 \text{ gC}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$, respectively (Fig. S11). The
489 PLC_{stem} graph in Fig. 5 indicated very high xylem embolism levels (i.e., reaching 100%
490 every year) under RCP4.5 and RCP8.5 already in the first year of simulations. A
491 pronounced embolism event was observed under the 'Hist' climate in 2017, 2018, 2019,
492 and 2022 in a 30%–45% range for the 'Control' plots, while the managed plots
493 experienced a maximum embolism of approximately 40% in 2017. Conversely, the 3D-
494 CMCC-FEM model did not report any significant differences between managed and
495 unmanaged plots for the LE. The average LE value for the 'Hist' climate was $1,796$
496 $\text{MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$, which decreased to $1,220 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$ under the RCP4.5 and $1,190 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$
497 under the RCP8.5 in managed plots. As previously described, the 'Control' plot
498 under the RCP8.5 experienced mortality in the sixth year of simulation. Similarly to the
499 previously described GPP fluxes, the MEDFATE model reported a slight difference in LE
500 fluxes between the 'Control' plot under historical climate conditions ($1,623 \text{ MJ}\cdot\text{m}^{-2}\cdot\text{year}^{-1}$

501 ¹) and the managed plots (1,603 MJ·m⁻²·year⁻¹). For the RCP4.5 and RCP8.5, the LE values
502 were 1,100 and 1,089 MJ·m⁻²·year⁻¹, respectively (Fig. S12). The LE values of the 'Control'
503 plots are not reported either for RCP4.5 or for RCP8.5 because of the mortality
504 experienced for the simulation years.

505

506 **4. Discussions**

507 First, this study evaluated the performances of two different process-based models in
508 simulating diverse beech stands across Europe, starting to the north of Europe and
509 moving towards the south under different environmental conditions. Secondly, the study
510 has focused on the models' sensitivity and the relative impacts of different management
511 options and different climatic conditions in two independent beech forest stands in the
512 north and south of the Italian peninsula.

513

514 **4.1 Model evaluation**

515 To assess the models' accuracy in predicting C and H₂O fluxes, we compared daily GPP
516 and LE data obtained from the EC towers. Both models predicted daily GPP and LE
517 accurately and ensured a good range of general applicability of both models (Kramer et
518 al., 2002; Verbeeck et al., 2008). The 3D-CMCC-FEM model seems to slightly
519 overestimate GPP daily values along latitudinal gradients starting from the north (DK-
520 Sor) to the south (IT-Col), as already found in Collalti et al. (2016). MEDFATE, in contrast,
521 showed a slight overestimation of GPP only at IT-Col site. The LE predicted by 3D-
522 CMCC-FEM is more accurate than MEDFATE prediction for DK-Sor and FR-Hes sites but
523 not in IT-Col site in which 3D-CMCC-FEM has shown to underestimate compared to the
524 observed EC values. For MEDFATE the underestimation of LE was observed in all the
525 evaluation sites.

526 The spread observed for the GPP and LE fluxes between the two models may be
527 attributed to the different assumptions that govern stomatal regulation since both

528 models use the Farquhar-von Caemmerer-Berry biochemical model to calculate
529 photosynthesis. The over or underestimation of the flows estimated by the models both
530 for GPP and the LE compared to the data observed from the EC towers can be attributed
531 either to the presence of the understory (although commonly sporadic in mature beech
532 stands), which was not considered in the simulations by both models and to errors on
533 daily measurement by EC technique (Loescher et al., 2006) or because a not perfect fit
534 in the modeled seasonality (i.e., the begin and the end of the growing season)
535 (Richardson et al., 2010). However, the overall leaf phenological pattern of the European
536 beech in these sites is well represented by the two models in almost all of the years
537 according to EC data as shown in supplementary materials (see Figs. S4, S5, S6, S7, S8,
538 S9, and S10). It is important to note that we did not specifically calibrate the models'
539 parameters at each site separately. Instead, both models were parameterized using
540 existing values taken from the literature, therefore with one single set of parameter
541 values for all sites.

542

543 **4.2 Climate change and forest management at the Cansiglio and Mongiana site**

544 The pre-Alpine site of Cansiglio showed slight differences in the fluxes (i.e., GPP and LE)
545 between the three different management practices and the three climate scenarios (i.e.,
546 no climate change, RCP4.5. and RCP8.5) used (Figs. S11–S12). Future climate is expected
547 to be higher temperature if compared to the historical one, with MAT higher of about
548 3.93°C under RCP4.5 and 4.95°C under RCP8.5 for the Cansiglio site and 4.52°C under
549 RCP4.5 and 5.42°C under RCP8.5 at the Mongiana site. Similarly, MAP is expected to be
550 510 mm lower under RCP4.5 and 602 mm under RCP8.5 at the Cansiglio site,
551 respectively, while 902 mm lower under RCP4.5 and 914 mm under RCP8.5 at the
552 Mongiana site, respectively.

553 Regarding management, the response of the 3D-CMCC-FEM to the removal of a
554 percentage of the basal area from the stand led to a decrease in GPP in the 'Traditional'
555 cutting and an even greater extent, in the 'Innovative' cutting compared to the 'Control'
556 (i.e., no management). Similarly to 3D-CMCC-FEM, the MEDFATE model simulates slight
557 differences in fluxes amount (e.g., lower values for 'Traditional' and 'Innovative' cutting
558 than 'Control' plots) between the management regimes in the plots. These results align
559 with those of Guillemot et al. (2014), who observed a slight decrease in GPP in managed
560 compared to unmanaged temperate beech forests in France under different thinning
561 regimes. However, differences were observed in both models under the three different
562 climates used in the simulations. GPP increased from the 'Hist' climate to RCP4.5 and
563 reached the maximum for RCP8.5, respectively. This suggests a plastic response (e.g.,
564 photosynthesis and stomatal response) of the stands, as simulated by models, to harsher
565 conditions, indicating, potentially, a high drought acclimation capacity (Petrik et al., 2022)
566 and increased GPP because of the early budburst, and a prolonged vegetative season
567 (Peano et al., 2019), and the so-called 'atmospheric CO₂ fertilization' effect as also found
568 by de Wergifosse et al. (2022) and Reyer et al. (2013), especially in sites with no apparent
569 water limitation both under current and projected future climate conditions. The
570 anisohydric behavior of *Fagus sylvatica* L. results in prolonged stomatal opening relative
571 to isohydric species, although Puchi et al. (2024) recently found large variability in
572 European beech responses, maintaining prolonged photosynthetic activity. Still, this
573 response is modulated by summer precipitation and the availability of soil water storage
574 (Leuschner et al., 2021; Baudis et al., 2015). However, for high-altitude stands, growth
575 could be negatively affected under warmer conditions, as suggested by Chmura et al.
576 (2024). The LE results for the 3D-CMCC-FEM showed lower values over the simulation
577 period for managed stands than unmanaged ones showing lesser sensitivity to forest
578 management if compared to MEDFATE. Yet, under the RCP4.5, the LE values were higher

579 compared to both the 'Hist' climate and the RCP8.5 one due to greater annual cumulative
580 precipitation than the RCP8.5 and higher, on average, temperatures than the 'Hist'
581 scenario (Fig. S12).

582 Conversely, the MEDFATE model was shown to be more sensitive to climate, with a
583 clearer distinction between the 'Hist' climate, the RCP4.5 and RCP8.5 climates, with
584 higher and nearly equal values in the harsher conditions (i.e., RCP4.5 and RCP8.5
585 climates), with slight differences in the management treatments as obtained by 3D-
586 CMCC-FEM.

587 The Non-Structural Carbon (NSC) amount showed the highest values in 'Innovative'
588 plots, followed by 'Traditional' plots, and the lowest values in 'Control' plots, suggesting
589 a benefit in carbon stock accumulation with more carbon going for carbon biomass and
590 less for reserve-replenishment for these stands under management interventions.
591 Nevertheless, NSC levels remain nearly the same for the three climate scenarios
592 throughout all the simulation years. It is important to note that MEDFATE simulated an
593 initial loss of stem conductance under the climate scenarios, indicating a premature
594 onset of water stress for the stand. Although in RCP4.5 this is negligible, in RCP8.5 PLC_{stem}
595 values reach a maximum xylem cavitation value of about 40% in the eighth year of
596 simulation for managed plots while for 'Control' plots in the eighth, ninth, and twelfth
597 years, highlighting potential benefits of management to reduce drought stress because
598 of less rain interception and canopy evaporation and transpiration (Giuggiola et al., 2018;
599 Schmied et al., 2023).

600 The GPP at the southern Apennine site of Mongiana showed a decrease under RCP4.5
601 and RCP8.5 scenarios when simulated by the 3D-CMCC-FEM model as a result of
602 harsher environmental conditions, as also resulted in the study by Yu et al. (2022), in
603 which the productivity and then the growth of European beech in southern regions are
604 expected to decrease as affected by more severe climate conditions such as decreased

605 precipitation and increased air temperature (Tognetti et al., 2019). Indeed, the increase
606 in air temperature, a reduction in soil water availability, and the rise in vapor pressure
607 deficit (VPD) lead to earlier stomatal closure, increased mesophyll resistance, and
608 elevated abscisic acid production (Kane and McAdam, 2023), all of which contribute to
609 a decrease in the carbon assimilation rate (Priwitzer et al., 2014; Grossiord et al., 2020).
610 Specifically, GPP is higher under 'Hist' climate conditions, decreases under the RCP4.5,
611 and ultimately reaches even lower values under the RCP8.5. Under the RCP8.5 at the
612 fifth year of simulation, the stand in the 'Control' plot is simulated to die due to carbon
613 starvation. The annual decline in NSC (Fig. 5) due to an imbalance between carbon
614 uptake (photosynthesis) and the demands for growth and respiration suggests that the
615 trees are unable to replenish their carbon reserves. The depletion of NSC reserves may
616 ultimately disrupt processes such as osmoregulation and phenology (Martínez-Vilalta et
617 al., 2016), potentially leading to stand defoliation and/or mortality. The management
618 options did not show changes in GPP under the 'Hist' climate. However, the increase of
619 GPP was observed under the RCP4.5 in the plots where 'Innovative' and 'Traditional'
620 cutting occurred, although no differences were observed between them. For instance,
621 the same increase in GPP was reported by Fibbi et al. (2019) for other European beech
622 forests under climate change scenarios in Italy. The thinning reduces the leaf area and
623 then the LAI and increases the soil water availability, which positively influences stomatal
624 conductance and carbon assimilation, providing an acclimation mechanism to drought
625 during periods of water scarcity (Lüttschwager and Jochheim, 2020; Diaconu et al.,
626 2017).

627 In contrast, the more intense cutting exhibited even lower GPP values than the 'Control'
628 plots. This is likely due to the overly intense thinning, which contrasts the microclimate
629 effects within this forest stand, reducing the potential to offset climate warming at the
630 local scale (Rita et al., 2021). Heavy thinning, on the other hand, can increase light

631 penetration, soil evaporation, and wind speed, thereby heightening tree sensitivity to
632 vapor pressure deficit under dry conditions (Schmied et al., 2023; Simonin et al., 2007).
633 LE decreased with the decrease in precipitation under the RCP4.5 and RCP8.5 climate
634 scenarios compared to the 'Hist' climate. There were no significant differences in LE
635 among the various management regimes. For the MEDFATE model, negligible or no
636 differences in GPP were observed under all the climates among various management
637 options. Although the GPP values estimated by the MEDFATE model under the RCP4.5
638 and RCP8.5 are similar to those obtained from the 3D-CMCC-FEM model, a closer
639 analysis of the daily outputs (data not shown) reveals that trees photosynthesize until the
640 end of July, after which they experience significant embolism (i.e., maximum value of
641 100%), as indicated by the PLC_{stem} graph, indicating that the decrease in precipitation led
642 to summer soil moisture depletion and lethal drought stress levels.
643 Furthermore, the 'Control' plots experienced mortality even before reaching the summer
644 period. In recent decades, prolonged drought stress in Mediterranean mountain regions
645 has significantly reduced the productivity of beech forests, resulting in a decline in Basal
646 Area Increment (BAI) and overall growth (Piovesan et al., 2008). It is also important to
647 note that under 'Hist' climate conditions, the MEDFATE model indicated a stem
648 embolization loss ranging from approximately 10% to 45% during the drought period (i.e.,
649 2018–2020) in Europe as also highlighted in other study (Italiano et al., 2024; Thom et al.,
650 2023; Lombardi et al., 2023). The embolization was more pronounced and long-lasting
651 in the 'Control' plots than the managed ones. The same trends were obtained for LE.

652

653 **4.3 Uncertainties and factors influencing forest carbon and water dynamics**

654 Although there is scientific evidence of the positive effects of CO₂ fertilization effect on
655 forest primary productivity, uncertainties remain regarding the long-term persistence of
656 this positive feedback and the level at which this may saturate (Sperlich et al., 2020;

657 Wang et al., 2020). Furthermore, the down-regulation of this fertilization effect on
658 photosynthesis is influenced by interannual variations in meteorological parameters, as
659 well as by interactions within the carbon and nitrogen cycles. These factors must be
660 carefully assessed to improve the accuracy of flux projections under future climatic
661 conditions (Zaehle et al., 2014). It is also worth highlighting that this study does not
662 account for biotic disturbances such as pest outbreaks and diseases, nor for abiotic
663 disturbances like extreme climatic events (e.g., heatwaves, late frosts, and wildfires).
664 These factors could significantly alter carbon and water fluxes (Yu et al., 2022b),
665 potentially depleting carbon reserves faster and reducing the capacity for carbon
666 sequestration by these forest ecosystems as well as possibly necessitating the adoption
667 of different management strategies (Langer and Bußkamp, 2023; Margalef-Marrase et al.,
668 2020). Another critical factor is the depth of the root zone and soil, as well as its physico-
669 chemical composition. For example, it has been found that two-thirds of fine roots in
670 European beech are within the top 30 cm of soil, while coarse roots can extend beyond
671 depths of 240 cm (Meier et al., 2017). Although findings by Gessler et al. (2021) indicate
672 that, unlike oak forests, European beech forests cannot compensate for additional water
673 uptake from deeper soil layers during drought periods, Brinkmann et al. (2018) reported
674 contrasting results. In this study, we analyzed soil texture characteristics to a depth of
675 110 cm for Cansiglio and 40 cm for Mongiana. However, the limited understanding of
676 deeper layers may not fully capture the entire soil water reservoir and its dynamics.
677 Expanding knowledge of the deepest soil layers is essential to better understand root
678 development and, consequently, improve water storage capacity and drought resilience
679 in beech forests.

680

681 **5. Conclusions**

682 The two process-based models provide robust evidence for their application in
683 estimating fluxes, consistent with long-term EC tower measurements in European beech

684 forests. Despite the minimal parametrization effort to align the two models and the
685 avoidance of single site-specific parameters, reliable results can still be obtained, as
686 confirmed by the outputs from the Sorø, Hesse, and Collelongo sites. Regarding the sub-
687 Alpine Cansiglio site, although water limitation does not significantly impact fluxes or the
688 health of the forest under 'Moderate' climate conditions (RCP4.5), a potential concern is
689 the embolization predicted by the MEDFATE model under the 'Hot' climate (RCP8.5) at
690 this site, despite similar levels of precipitation. The high susceptibility of European beech
691 forests at the southern Apennine site of Mongiana to more severe (i.e., hotter and drier)
692 climatic conditions could lead to the collapse of this forest ecosystem, even with the
693 application of management options to reduce competition. However, it is crucial that
694 these seasonal droughts are not prolonged or intense enough to exceed the ecological
695 limits of the European beech. To avoid that European beech forests may necessitate of
696 strategic and specifically designed management planning at the single site level,
697 including the ability to project (e.g., with forest models) and evaluate future forest
698 conditions for better management schemes. However, the ability of these forests to
699 survive or resist the impacts of climate change may not depend solely on density
700 reduction interventions. Prioritizing the exploration of alternative sustainable
701 management strategies to promote carbon sequestration in both above-ground biomass
702 and soil is crucial for enhancing climate change mitigation efforts. Additionally,
703 evaluating silvicultural plans such as the introduction of complementary species can
704 improve the resilience of vulnerable European beech ecosystems. A modeling approach,
705 similar to the one used in this study, offers a valuable tool for assessing these alternative
706 strategies and refining forestry adaptive management practices. By integrating these
707 approaches, we can strengthen the long-term sustainability of forests while preserving
708 the ecological balance of vulnerable regions.

709

710

711

712 **CRedit authorship contribution statement**

713 **Vincenzo Saponaro:** Conceptualization, Data curation, Formal analysis, Investigation,
714 Methodology, Resources, Software, Visualization, Writing – original draft, Writing –
715 review & editing. **Miquel De Càceres:** Conceptualization, Methodology, Software,
716 Supervision, Writing – review & editing. **Daniela Dalmonech:** Conceptualization,
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721 **Data availability**

722 Data will be made available on request.

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757

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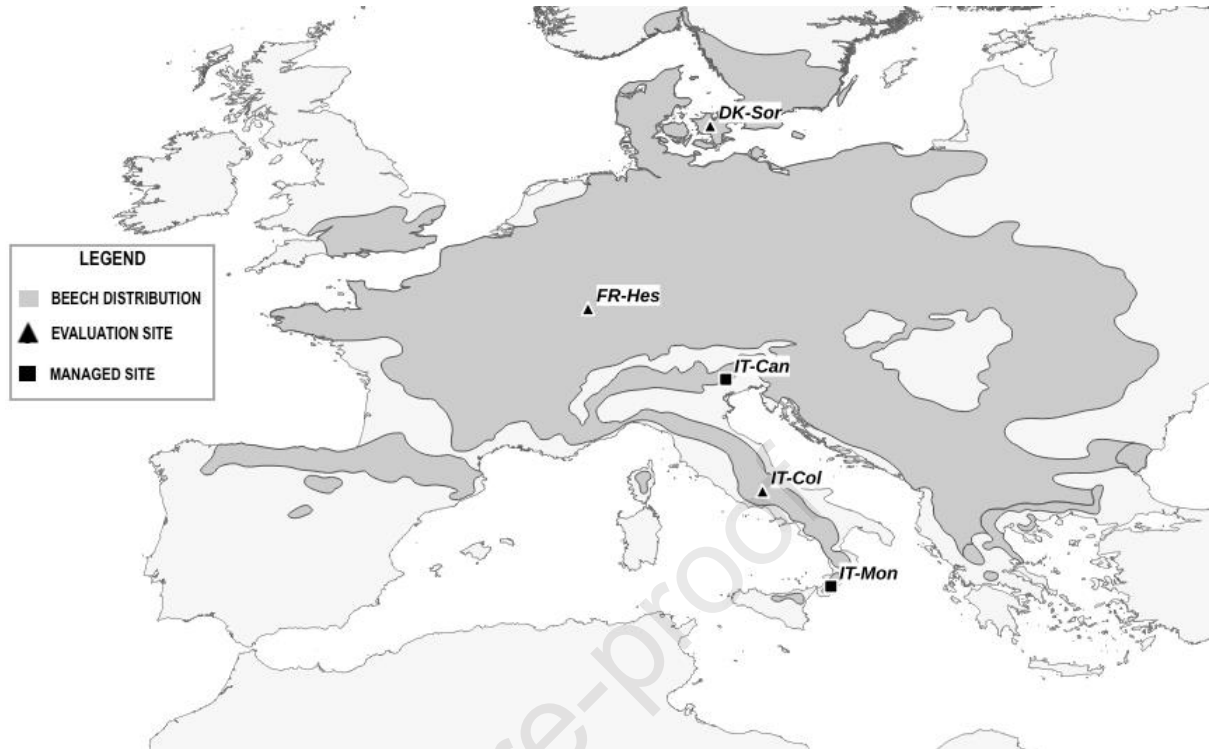
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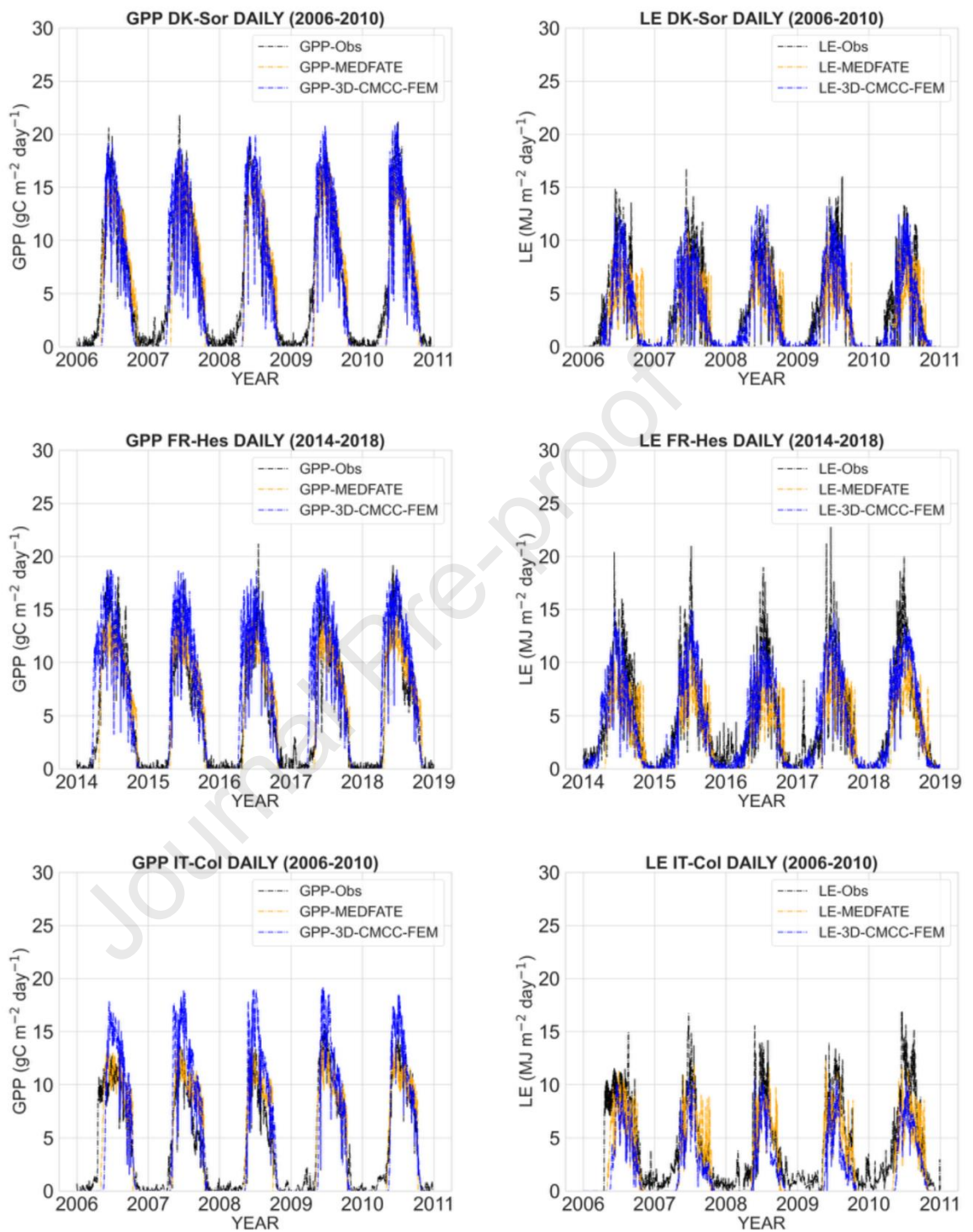
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Fig. 1. Map of the study sites. Triangles represent sites for validating fluxes, while the squares represent sites for management investigation.

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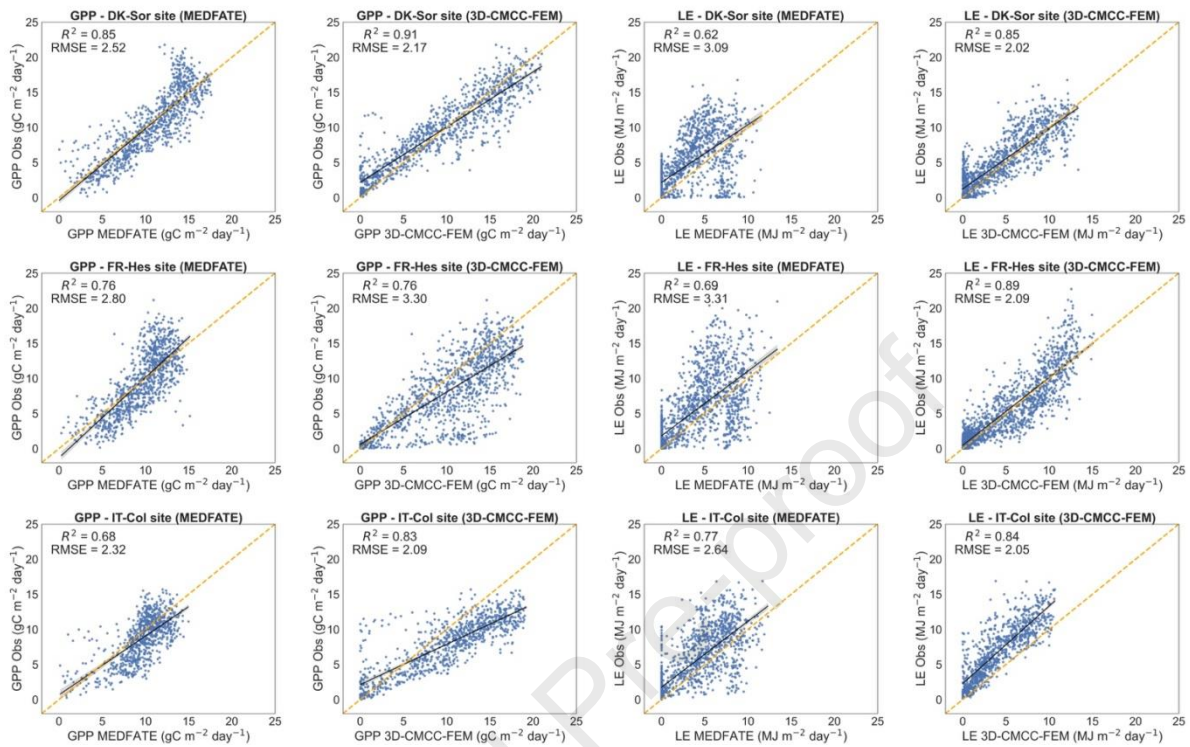
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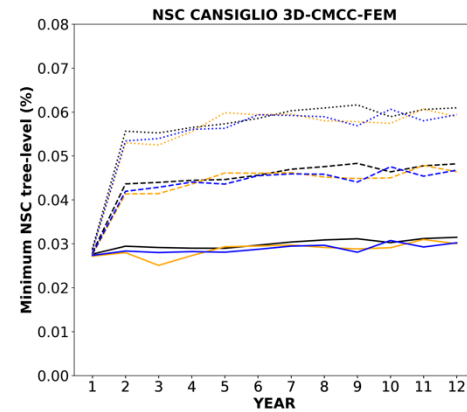
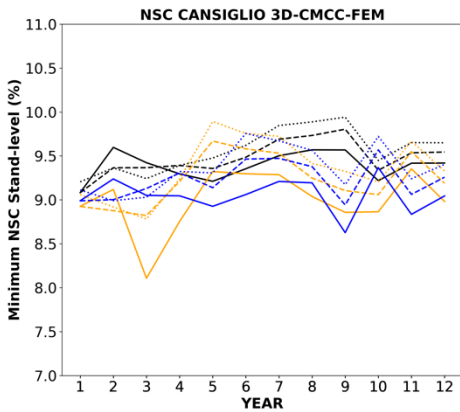
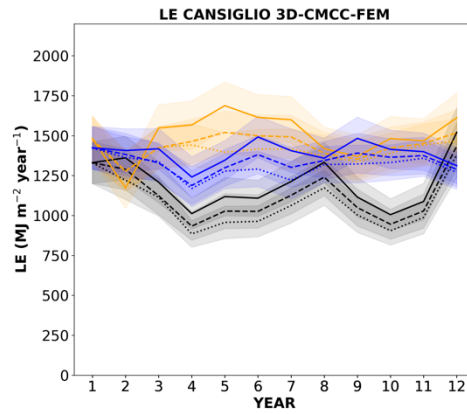
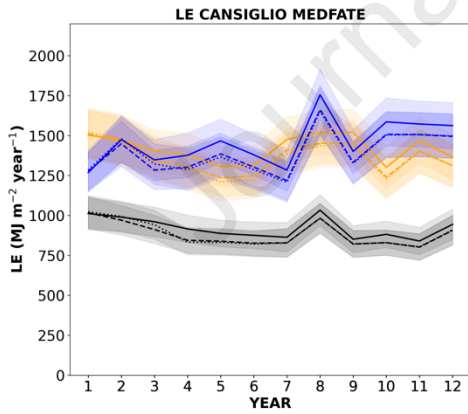
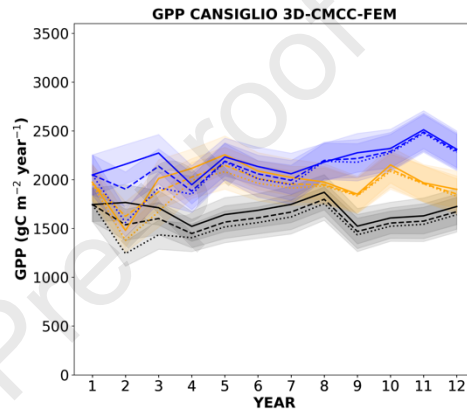
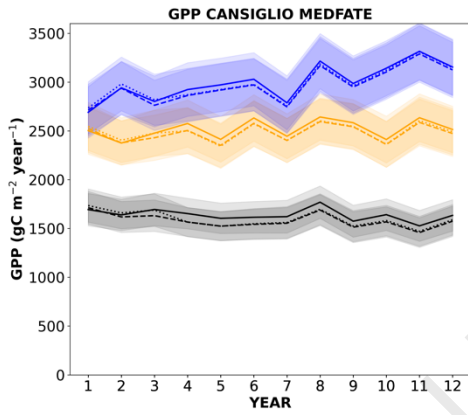
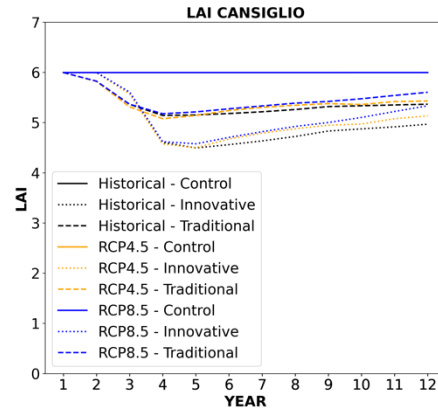
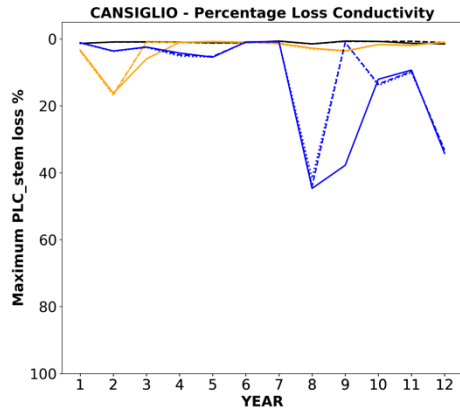
Fig. 2. Daily mean variations of GPP (gC·m⁻²·day⁻¹) and LE (MJ·m⁻²·day⁻¹) estimated from the direct micrometeorological eddy covariance measurements (GPP-Obs and LE-Obs) and models' simulation (GPP-3D-CMCC-FEM, LE-3D-CMCC-FEM and, GPP-MEDFATE, LE-MEDFATE) during the evaluation period at the DK-Sor, IT-Col and FR-Hes at the Beech forest in 2006–2010 and 2014–2018, respectively.

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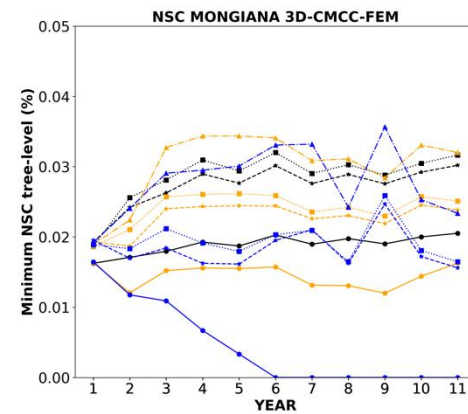
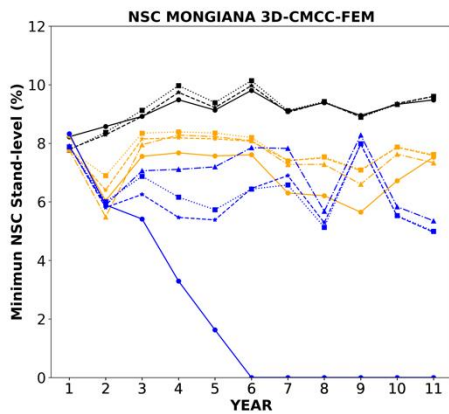
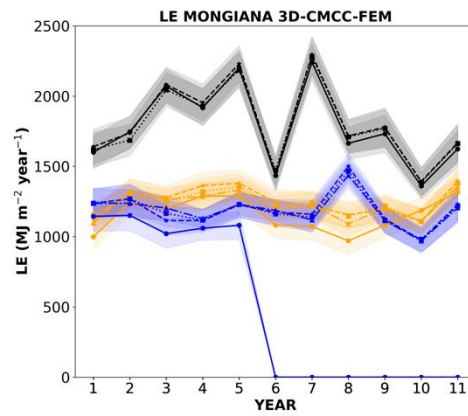
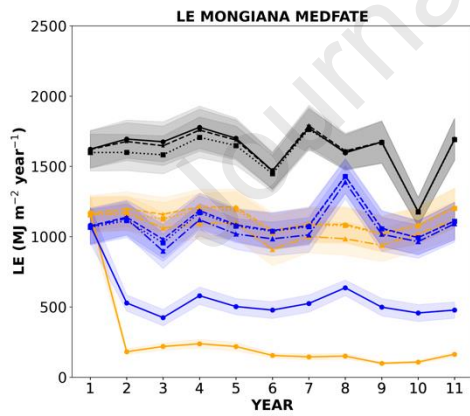
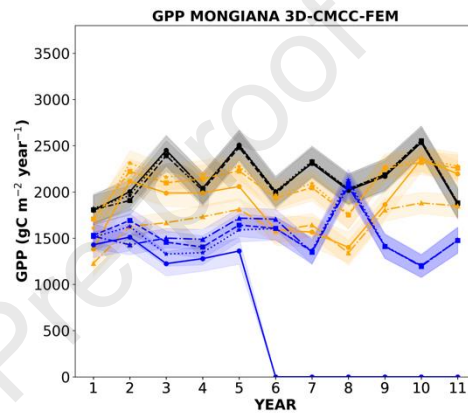
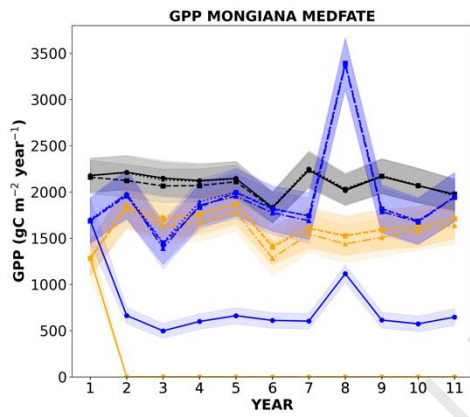
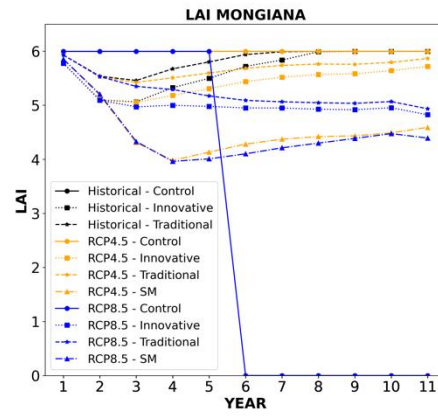
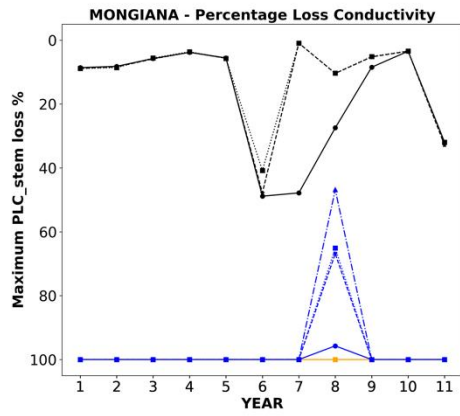


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Fig. 3. Scatter plots and linear regressions of GPP ($\text{gC} \cdot \text{m}^{-2} \cdot \text{day}^{-1}$) and LE ($\text{MJ} \cdot \text{m}^{-2} \cdot \text{day}^{-1}$) of the models versus the direct micrometeorological eddy covariance measurements (Obs) at the Sorø (DK-Sor; 2006–2010 period), Collelongo (IT-Col; 2006–2010 period) and Hesse (FR-Hes; 2014–2018 period).



1311 **Fig. 4.** Comparative analysis between models output at the Cansiglio site. The top-left panel displays the
1312 PLC_{stem} as modelled by MEDFATE, while the top-right panel shows the modelled LAI for 3D-CMCC-FEM
1313 (and used by MEDFATE). The middle-up panels (left and right) present annual GPP ($gC \cdot m^{-2} \cdot year^{-1}$) as
1314 modelled by the MEDFATE and 3D-CMCC-FEM, respectively. The middle-down panels (left and right)
1315 depict annual LE ($MJ \cdot m^{-2} \cdot year^{-1}$) modelled by the MEDFATE and 3D-CMCC-FEM, respectively. The bottom
1316 panels (left and right) depict the annual minimum of NSC concentration (%) at the stand and tree level,
1317 respectively, as modelled by the 3D-CMCC-FEM. Different plot management strategies are represented
1318 by distinct line styles: solid lines for 'Control' plots ('no management'), dotted lines for 'Innovative' plots,
1319 and dashed lines for 'Traditional' plots (Shelterwood). Climate scenarios are indicated by line colours: black
1320 for 'Hist' climate data (2010–2022), orange and blue for RCP4.5 and RCP8.5 climate (2059–2070),
1321 respectively.
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1324 **Fig. 5.** Comparative analysis between models output at the Mongiana site. The top-left panel displays the
 1325 percent loss of PLC_{stem} as modelled by MEDFATE while the top-right panel shows the modelled LAI for
 1326 3D-CMCC-FEM (and used by MEDFATE). The middle-up panels (left and right) present annual GPP
 1327 ($gC \cdot m^{-2} \cdot year^{-1}$) as modelled by the MEDFATE and 3D-CMCC-FEM, respectively. The middle-down panels
 1328 (left and right) depict annual LE ($MJ \cdot m^{-2} \cdot year^{-1}$) modelled by the MEDFATE and 3D-CMCC-FEM,
 1329 respectively. The bottom panels (left and right) depict the annual minimum of NSC concentration (%) at
 1330 the stand and tree level, respectively, as modelled by the 3D-CMCC-FEM. Different plot management
 1331 strategies are represented by distinct line styles: solid lines with circles for 'Control' plots ('no
 1332 management'), dotted lines with squares for 'Innovative' plots, dashed lines with stars for 'Traditional' plots
 1333 (Shelterwood) and dash-dotted lines with triangles for 'SM' management. Climate scenarios are indicated
 1334 by line colours: black for 'Hist' climate data (2010–2022), orange and blue for RCP4.5 and RCP8.5 climate
 1335 (2060–2070), respectively.

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Table 1. Characteristics of the study sites. The age of the stands refers to 2010. The mean annual
 temperature (MAT) and mean annual precipitation (MAP) for DK-Sor, FR-Hes and IT-Col refer to the period
 evaluated (i.e., 2006–2010 for the Sorø and Collelongo site and 2014–2018 for the Hesse site) while for
 Cansiglio and Mongiana from 2010 to 2022. The sum of precipitation in summer refers to June (J), July (J)
 and August (A) for the same period.

Variables	Site description				
	Evaluation sites			Test sites	
	DK-Sor	FR-Hes	IT-Col	Cansiglio	Mongiana
Coordinates (WGS84)	55°49' N, 11°64' E	48°66' N, 7°08' E	41°85' N, 13°59' E	46°02' N, 12°22' E	38°29' N, 16°14' E
Country	Denmark	France	Italy	Italy	Italy
Altitude (m a.s.l.)	40	305	1500	1300	1300
Area (ha)	1	1	1	27	27
MAT (°C)	8.52	10.27	6.95	6.44	11.01
MAP (mm)	818	853	1075	2219	1701
Slope (%)	-	5	35	12	10
Aspect (°)	0	0	252	135	135
Stand age (year)	90	45	118	120	90
Summer prec (J-J-A) (mm)	292	205	120	493	141

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1347 **Table 2.** Parameters and variables set for both models during the simulations.

Parameters and variables		
Names	Values	Units
g_{s_max}	0.006, Pietsch et al. (2005)	$m \cdot s^{-1}$
J_{max}	-160, De Cáceres et al. (2023)	$\mu mol \text{ photons} \cdot m^{-2} \cdot s^{-1}$
V_{cmax}	-95, De Cáceres et al. (2023)	$\mu mol \text{ CO}_2 \cdot m^{-2} \cdot s^{-1}$
LAI	from 3D-CMCC-FEM to MEDFATE	$m^2 \cdot m^{-2}$
ASW	from 3D-CMCC-FEM to MEDFATE	mm

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1375 **Table 3.** The correlation coefficient (R^2), the Root Mean Square Error (RMSE) and the Mean Absolute Error (MAE) for, the GPP ($\text{gC}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$) and LE ($\text{MJ}\cdot\text{m}^{-2}\cdot\text{day}^{-1}$)
 1376 of the daily simulations at DK-Sor, IT-Col, and FR-Hes sites performed from both models 3D-CMCC-FEM and MEDFATE in the beech forest stands.

Sites	3D-CMCC-FEM						MEDFATE					
	GPP			LE			GPP			LE		
	R^2	RMSE	MAE	R^2	RMSE	MAE	R^2	RMSE	MAE	R^2	RMSE	MAE
DK-Sor	0.91	2.17	1.65	0.85	2.02	1.58	0.85	2.52	1.97	0.62	3.09	2.39
FR-Hes	0.76	3.30	2.46	0.89	2.09	1.47	0.76	2.80	2.21	0.69	3.31	2.37
IT-Col	0.83	2.09	1.56	0.84	2.05	1.57	0.68	2.32	1.87	0.77	2.64	1.90

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Highlights

- Modelled carbon and water fluxes under different climates and management regimes
- Different climates increases fluxes in the north and decreases them in the south
- 3D-CMCC-FEM and MEDFATE satisfactorily predicted productivity and latent heat
- 3D-CMCC-FEM predicts carbon starvation, MEDFATE predicts stem embolism in the south
- High thinning intensity of the stand in the south negatively affected carbon fluxes

Declaration of Competing Interest

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

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