

1 **Relative contribution of plant traits and soil properties to the functioning of a temperate**
2 **forest ecosystem in the Indian Himalayas**

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19 **Abstract.**

20 Plant-soil interactions are a major determinant of changes in forest ecosystem processes and
21 functioning. We conducted a trait-based study to quantify the contribution of plant traits and soil
22 properties to above- and below-ground ecosystem properties in temperate forest in the Indian
23 Himalayas. Nine plant traits (leaf area, specific leaf area, leaf water content, leaf dry matter
24 content, leaf carbon (C), nitrogen (N), phosphorus (P), leaf C/N, and leaf N/P) and eight soil
25 properties (pH, moisture, available N, P, potassium (K), total C, N, P) were selected for
26 determination of their contribution to major ecosystem processes (above-ground biomass C, soil
27 organic C, soil microbial biomass C, N, and P, and soil respiration) in temperate forest. Among
28 the plant traits studied, leaf C, N, P, and leaf N/P ratio proved to be the main contributors to
29 above-ground biomass, explaining 20-27% of variation. Leaf N, P, and leaf N/P were the main
30 contributors to below-ground soil organic C, soil microbial biomass C, N, and P, and soil
31 respiration (explaining 33% of variation). Together, the soil properties pH, available P, total N
32 and C explained 60% of variation in above-ground biomass, while pH and total C explained 56%
33 of variation in soil organic C. Other soil properties (available P, total C and N) also explained
34 much of the variation in soil microbial biomass C (52%) and N (67%), while soil pH explained
35 some of variation in soil microbial biomass N (14%). Available P, total N, and pH explained soil
36 microbial biomass P (81%), while soil respiration was only explained by soil total C (70%). Thus
37 leaf traits and soil characteristics make a significant contribution to explaining variations in
38 above- and below-ground ecosystem processes and functioning in temperate forest in the Indian
39 Himalayas. Consequently, tree species for afforestation, restoration, and commercial forestry
40 should be carefully selected, as they can influence the climate change mitigation potential of
41 forest in terms of C stocks in biomass and soils.

42 **Keywords:** Ecosystem services, biomass carbon, leaf nitrogen, specific leaf area, soil microbial
43 carbon, soil organic carbon

44

45 **1. Introduction**

46 Temperate forests provide various ecological services world-wide (Frelich, 2002; Tateno et al.,
47 2004). They contribute 17% of global net primary productivity, provide 315 Gigatons of carbon
48 (C) storage, are an important source of timber and wood products, and regulate hydrological,
49 nitrogen (N), and C cycles (de Gouvenain and Silander, 2017; Negi, 2018). Temperate forests are
50 characterized by a moderate climate, fertile soils, and high productivity (Singh et al., 2017; Saha
51 et al., 2018). However, they are sensitive to environmental changes such as global climate
52 change (Frelich, 2002; Alexander et al., 2018), and deforestation and degradation activities
53 (Walther et al., 2002; Malik et al., 2016; Negi, 2018). Natural (landslide, wind damage, grazing,
54 etc.) and anthropogenic (forest fire, grazing, logging, lopping, development activities, etc.)
55 disturbances in forest ecosystems can alter forest composition and structure (Frelich, 2002; Dar
56 and Sundarapandian, 2016). As a result, disturbances can lead to loss of biodiversity and altered
57 ecological functioning (Arunachalam et al., 1996; Negi, 2018).

58 World-wide, many studies have explored the relationship between plant functional traits,
59 soil properties, and ecological functions such as productivity (Eskelinen et al., 2012; Roscher et
60 al., 2012), decomposition (Bakker et al., 2010; Carrillo et al., 2017), nitrification and C storage
61 (Conti and Diaz, 2013), soil fertility (Berner and Law, 2016), above-ground biomass, litter, soil
62 organic matter, N mineralization, and leaching of inorganic N (Grigulis et al., 2013) in different
63 ecosystems. These studies have shown e.g., that 56.7% of variation in microbial properties in
64 semi-arid ecosystems is explained by plant traits, and 63.8% by soil properties (pH, total N, total
65 C, ammonium-N (NH₄-N), nitrate-N (NO₃-N), and phosphate-phosphorus (PO₄-P)) (Chai et al.,
66 2019). Plant traits such as shoot traits (shoot N and C, leaf dry matter content) and root traits
67 (root N, root C/N ratio, root dry matter content) can be used to predict soil properties and

68 ecosystem functions in temperate grasslands (Long et al., 2019). Soil properties and plant
69 functional traits are important for regulating the functioning of tropical cloud forests (Hernández-
70 Vargas et al., 2019), while soil heterogeneity and plant functional traits can predict variation in
71 species distribution in subtropical forests (Zhang et al., 2017).

72 Studies in Indian Himalayan forests are fewer and have focused solely on variations in
73 biodiversity (Sharma et al., 2010, 2011) and soil characteristics (Gairola et al., 2012; Joshi and
74 Negi, 2015; Dar and Sundarapandian, 2016; Saha et al., 2018). As a result, there is limited
75 understanding of the impact of plant traits and soil properties on ecosystem functioning in
76 temperate forests in the Indian Himalayan region. Under global change, the current
77 understanding is inadequate for determining the mechanisms of ecosystem processes and
78 functioning in temperate forests. The major causes of damage to Himalayan moist temperate
79 forests are forest fires (55.3%), damage to plants due to deforestation/logging activities (87.9%),
80 grazing (91.6%), and other human development activities (85.6%) (<http://fsi.nic.in>). Himalayan
81 temperate forests are sensitive to disturbance, but in undisturbed form they have great
82 biodiversity and contribute to global biogeochemical and nutrient cycling. There is thus a need to
83 quantify the contribution of plant traits and soil properties to ecosystem functioning in order to
84 manage temperate forests.

85 In this study, we analyzed the relative contributions of plant traits and soil properties to
86 above- and below-ground ecosystem processes. We hypothesized that above-ground plant traits
87 have stronger effects on above-ground ecosystem processes than on below-ground ecosystem
88 processes. We measured (i) plant functional traits, (ii) soil properties, and (iii) ecosystem
89 properties, including above-ground biomass, soil organic matter, soil microbial biomass C, N,

90 and P, and soil respiration. We then determined the relative effects of plant traits and soil
91 properties on ecosystem properties.

92

93 **2 Material and methods**

94 **2.1. Study site**

95 A forest stand within the temperate central Himalayan region of Uttarakhand, India
96 (30°28'02.6"N; 78°05'47.9"E) was selected as the study area. It is located at an altitude of 2200
97 m above sea level, in a region with mean annual temperature of 25°C and mean annual rainfall of
98 2150 mm. The vegetation in the area is Himalayan moist temperate forest. The soil is acidic and
99 comprises regosols, leptosols, and dolomite (Raina and Gupta, 2009).

100 **2.2 Vegetation analysis**

101 A study plot was randomly selected through grid (cell) selection based on a random number table
102 applied to grids across the study area, and a vegetation survey was carried out in this selected
103 plot using a quadrat method. During the peak plant growing season (March-April), 20 plots (10
104 m x10 m) were staked out in the forest, using a gridded randomization method in which the plots
105 were selected on a non-replacement basis (Lavrakas, 2008). Three individual trees were selected
106 at random within each plot, for evaluation of plant traits. Therefore, a total of 60 individual trees
107 were studied. The dominance of plant species was determined using importance value index
108 (IVI) and species abundance in a field survey. Next, 10 dominant tree species (seven evergreens,
109 three deciduous) (Table 1) were selected within plots, to cover a variety of plant traits in the
110 forest area (Table 2). Sampling of plants and soils was performed in three different seasons:
111 winter (October-February), summer (May-June), and rainy season (July-September).

112 **2.3 Plant traits**

113 All measurements were taken following an existing protocol (Pérez-Harguindeguy et al., 2013).
114 In each of the 20 plots, three healthy adult specimens of the most dominant woody plant species
115 were selected for leaf trait measurements (Table 2), making a total of 60 individuals. For each
116 selected individual, leaves that were young, fully expanded, and under the highest sunlight
117 exposure were collected. The leaves were placed in plastic bags and transported in a chilled, dark
118 container to the laboratory, where they were stored at 10-15°C. Leaf area was measured on five
119 fully expanded fresh young leaves per individual tree, by scanning using an LI-3100C instrument
120 (LI-COR, LI-3100C). Specific leaf area was calculated as the area of one side of a leaf divided
121 by its oven-dried weight (at 70°C for 48 h). Leaf dry matter content was calculated as the ratio
122 between dry and fresh (saturated) leaf weight. Additional leaves (10-20 leaves) were collected
123 from the outer canopy of each tree for chemical analysis. These leaves were separated, oven-
124 dried (at 70°C for 48 h) and milled using a stainless steel mill (Wiley, Thomas Scientific). Leaf C
125 and N content were determined using an elemental analyzer (CHNS, Euro, EA-3000). Leaf P
126 content was analyzed by the acid digestion method (Pérez-Harguindeguy et al., 2013).

127 **2.4 Soil properties**

128 Each of the 20 selected plots was characterized by analyzing soil samples (three per plot).
129 Standard procedures were followed for soil analysis (Anderson and Ingram, 1994). Soil samples
130 were collected from 0-30 cm depth at randomly selected points, to assess the physicochemical
131 properties of soil under the canopy of the selected dominant tree species. A total of 60 soil
132 samples were taken from each plot in three seasons. The composite soil sample from each
133 sampling point was divided into two equal parts. One part was immediately (within 24 h) sieved
134 (2-mm mesh) and analyzed for pH (digital pH meter), moisture content (gravimetric method),
135 ammonium-N (Kjeldahl method), and available P (molybdenum blue method). The other part

136 was sieved through a 2-mm mesh, air-dried under laboratory conditions, and analyzed for total C
137 and total N, using a CHNS analyzer (CHNS, Euro, EA-3000). Total P was determined by
138 spectrophotometer using the Bray-Kurtz method and soil potassium (K) content was determined
139 using a flame photometer (Systonics, 128).

140 **2.5 Ecosystem properties**

141 The primary ecosystem properties on which the influence of plant traits and soil properties was
142 studied were: above-ground biomass, soil organic C, soil microbial biomass C, N, and P, and soil
143 respiration. Three samples per plot of tree biomass and three samples per plot of soil were taken
144 for the analyses. The height and diameter at breast height (DBH) of all trees with DBH ≥ 10 cm
145 within each sample plot were measured. The biomass of dominant tree species was taken to
146 represent above-ground biomass (Mg C ha^{-1}). The growing stock volume density (GSVD) of
147 each tree species was estimated using species-specific volume equations developed using
148 multiple regression methods by the Forest Survey of India (FSI, 1996) (Table S1). The estimated
149 GSVD values ($\text{m}^3 \text{ha}^{-1}$) were then converted into above-ground biomass by multiplying by an
150 appropriate biomass expansion factor (BEF, Mg m^{-3}):

$$151 \quad \text{Biomass} = \text{GSVD} \times \text{BEF}$$

152 where BEF was defined as the ratio of above-ground biomass density of all living trees with
153 DBH ≥ 2.54 cm to GSVD for all trees with DBH ≥ 12.7 cm.

154 The equations used for calculation of BEF for hardwood, spruce-fir, and pine tree species were
155 (FSI, 1996):

156 Hardwood:

$$157 \quad \text{BEF} = \exp\{1.91 - 0.34 \times \ln(\text{GSVD})\} \text{ (for GSVD} < 200 \text{ m}^3 \text{ha}^{-1}\text{)}$$

$$158 \quad \text{BEF} = 1.0 \text{ (for GSVD} > 200 \text{ m}^3 \text{ha}^{-1}\text{)}$$

159 Spruce-fir:

$$160 \quad \text{BEF} = \exp\{1.77 - 0.34 \times \ln(\text{GSVD})\} \text{ (for GSVD} < 160 \text{ m}^3 \text{ ha}^{-1}\text{)}$$

$$161 \quad \text{BEF} = 1.0 \text{ (for GSVD} > 160 \text{ m}^3 \text{ ha}^{-1}\text{)}$$

162 Pine:

$$163 \quad \text{BEF} = 1.68 \text{ (for GSVD} < 10 \text{ m}^3 \text{ ha}^{-1}\text{)}$$

$$164 \quad \text{BEF} = 0.95 \text{ (for GSVD} = 10 \text{--} 100 \text{ m}^3 \text{ ha}^{-1}\text{)}$$

$$165 \quad \text{BEF} = 0.81 \text{ (for GSVD} > 100 \text{ m}^3 \text{ ha}^{-1}\text{)}$$

166 Above-ground biomass C was estimated using the IPCC conversion rate of 0.47 for biomass to
167 carbon:

$$168 \quad \text{C} = 0.47 \text{ Biomass}$$

169 where C and biomass are in Mg ha^{-1} .

170 For determination of soil organic C, three soil samples were taken from the top 30 cm of each
171 plot using a soil corer (10 cm diameter) and mixed thoroughly to obtain a composite soil sample.
172 These samples were air-dried, sieved through a 2-mm mesh, and analyzed for organic C using an
173 established procedure (Walkley and Black, 1934).

174 Soil microbial biomass C and N were estimated by the chloroform fumigation-extraction method
175 (Brookes et al., 1984; Vance et al., 1987). Soil samples were subjected to one of two treatments
176 (chloroform-fumigated and unfumigated), extracted in 0.5 N K_2SO_4 and simultaneously digested
177 and titrated against ferrous ammonium sulfate using 1,10-phenanthroline monohydrate as the
178 indicator (chloroform-fumigated samples) or against N/140 HCl using boric acid as the indicator
179 (unfumigated samples). Soil microbial biomass P was estimated by chloroform fumigation-
180 extraction using 0.5 N NaHCO_3 (Brookes et al., 1984). In all cases, the values obtained for

181 unfumigated samples were subtracted from those obtained for chloroform-fumigated samples, to
182 obtain microbial C, N, and P content.

183 Soil respiration was measured as carbon dioxide (CO₂) evolution, using a single-chamber device
184 (LI-8100, LI-COR Biosciences, Lincoln, NE). The closed-chamber method was used to estimate
185 the flux of CO₂ at the soil surface. In this method, a small portion of air is circulated from a
186 chamber to an infrared gas analyzer, and then sent back to the chamber (Madsen et al., 2009). All
187 measurements were programmed using a LICOR 8100 palm-held wireless controller linked with
188 the LI-8100 device. PVC soil collars measuring 20 cm in diameter were inserted to a depth of 3-
189 5 cm and extended approximately 6-10 cm above the soil surface. In total, measurements were
190 made at 60 points.

191 **2.6 Data analysis**

192 Principal component analysis (PCA) and linear mixed modeling with residual maximum
193 likelihood (REML) estimations were carried out with lme4 package R studio (R-3.5.1 version) to
194 quantify the respective contributions of plant traits and soil properties to the variation in selected
195 ecosystem processes. PCA was performed to identify significant contributors among the
196 multivariable components on the basis of their eigenvalue and percentage variance (Orwin et al.,
197 2010). REML is defined as a form of maximum likelihood estimation that uses a likelihood
198 function calculated from a transformed set of data, so that nuisance parameters have no effect.
199 REML provides estimates of parameter effects and variance components for both the fixed and
200 random effects in the model (Grigulis et al., 2013). In this analysis, the response variables were
201 ecosystem processes (above-ground biomass, soil organic C, soil microbial biomass C, N and P,
202 and soil respiration). Fixed effects were plant traits (leaf area, specific leaf area, leaf water
203 content, leaf dry matter content, leaf C content, leaf N content, leaf P content, leaf C/N, and leaf

204 N/P) and soil properties (pH, soil moisture content, total C, total N, available N, available P, total
205 P, K), with plant species and plot as random effects. All variables were tested for normality and
206 log-transformed before analysis.

207

208 **3 Results**

209 The average values obtained for plant traits, soil characteristics, and ecosystem properties are
210 presented in Figure S1 in Supplementary Information. Total percentage of variation explained by
211 plant traits was 84% (Figure 1a) and by soil parameters 64% (Figure 1b). Among all plant traits
212 studied, specific leaf area, leaf water content, leaf C, N, and P, and N/P ratio were significant
213 contributors to the variation in ecosystem properties (Table 3a). Among the soil characteristics
214 studied, pH, available P, total C, and total N were significant contributors (Table 3a). These
215 important contributors were considered in model building, to identify their contribution.

216

217 **3.1 Control of above-ground biomass and leaf traits**

218 Variation in above-ground biomass was best explained (to 20-27%) by functional properties of
219 the leaf, i.e., leaf C, N, and P, and leaf N/P ratio. Variation in soil organic C was best explained
220 (33%) by leaf N and P, and leaf N/P ratio. Above-ground biomass and soil organic C showed a
221 positive correlation with leaf N (Table 3a). Both variables were negatively correlated with leaf P
222 and leaf N/P ratio, while above-ground biomass also showed a negative correlation with leaf C.
223 The effects of random variables (i.e., soil properties, species, and plot) on above-ground biomass
224 and soil organic C were as follows: soil properties (pH, available P, total C and N) explained
225 60% of variation in above-ground biomass and 56% of variation in soil organic C; species
226 explained 30% of variation in above-ground biomass and 10% of variation in soil organic C; and

227 plot explained 10% of variation in above-ground biomass and 34% of variation in soil organic C
228 (Table 3a).

229

230 **3.2 Control of below-ground microbial biomass and respiration**

231 Common leaf traits, i.e., leaf N, leaf P, and leaf N/P, explained 33% of the variation in soil
232 microbial biomass C, N, and P, and in soil respiration. These ecosystem processes were
233 positively correlated with leaf P and leaf N/P, and negatively with leaf N (Table 3a). Soil
234 microbial biomass C (52% explained) and N (67% explained) were affected by common soil
235 parameters (available P, total C, and total N) (Table 3a). Soil pH affected soil microbial biomass
236 N (14.5%), but not microbial biomass C (Table 3a). Above-ground species had only a minor
237 effect on soil microbial biomass C (2%) and no effect on soil microbial biomass N (Table 3a).
238 Plot variance made a significant contribution to explaining soil microbial biomass C (45%) and
239 N (33%) (Table 3a). Soil microbial biomass P was highly influenced (81%) by soil properties
240 (available P, total N, and pH) and plant species (19%), while plot had no effect (Table 3a). Soil
241 total C (70%) and plot (30%) explained all of the variation in soil respiration (Table 3a).

242 The full and null models were estimated and validated based on Akaike Information
243 Criterion (AIC), deviance, and chi-square (Winter, 2013). There model fit was better for some
244 parameters, as indicated e.g., by lower AIC values (Table 3b).

245

246 **4. Discussion**

247 This study identified the relative contribution of plant traits and soil properties in explaining
248 major ecosystem properties in a temperate forest ecosystem in the Indian Himalayas. In
249 combination, the plant traits and soil properties tested explained 33-70% of the variation in

250 ecosystem properties. In previous studies, the contribution of plant traits and soil properties to
251 explaining variation in above- and below-ground productivity has been within the range 16-38%
252 (Orwin et al., 2010), 60-90% (Grigulis et al., 2013), 36-100% (Legay et al., 2014), and 3-100%
253 (Legay et al., 2016).

254

255 **4.1 Plant functional traits and ecosystem properties**

256 We found that above-ground biomass and soil organic C were both positively associated with
257 leaf N. The positive association of leaf N with above-ground biomass reflects the exploitative
258 strategy employed by plant species to gain a faster return from nutrient exchange and investment
259 in the ecosystem (Wright et al., 2004; Díaz et al., 2016; Rawat et al., 2019). These results are
260 consistent with previous findings that plant traits explain 60% of variation in above-ground
261 biomass (Becknell and Powers, 2014). Other studies have also found a positive association
262 between leaf traits and biomass increments in forest ecosystems (Finegan et al., 2015 and Bu et
263 al., 2019). Similarly to previous studies (Angst et al., 2019; Bardgett, 2017; Chen et al., 2018;
264 Qiu et al., 2018), we found a significant positive association of leaf traits with soil C. However,
265 the impact of leaf traits on soil C is not straight-forward, as different studies have found
266 contrasting results. For instance, soil organic C has been found to have a positive link with leaf
267 N, as N-rich vegetation contains high soil C stocks (Manning et al., 2015). Other have reported a
268 negative relationship between leaf traits and community-weighted mean leaf N concentration
269 (Ali et al., 2017), or no relationship between leaf traits and C storage (Conti and Diaz, 2013).

270 Previous studies have indicated a negative relationship between leaf N/P ratio and
271 biomass production (Güsewell, 2004; Wang et al., 2015). Similarly, we found that above-ground
272 biomass and soil organic C were negatively correlated to leaf N/P ratio, and also to leaf P. In

273 contrast, a study across terrestrial biomes in China found an overall positive relationship between
274 leaf P and biomass production (Tang et al., 2018). However, that study also found major
275 fluctuations in the relationship between leaf P and biomass production in different biomes, with
276 no significant relationship in deciduous broadleaf forests and mixed forests, but a negative
277 relationship in evergreen broadleaf forests (Tang et al., 2018). Deciduous trees have a higher
278 leaf P content than evergreen trees (Tang et al., 2018). In addition, plant C and plant P have been
279 shown to decrease with increasing temperature (Reich and Oleksyn, 2004; Tang et al., 2018) and
280 precipitation (Tang et al., 2018). Considering these large variations between biomes and forest
281 types, there is a need for more studies on the relationships between leaf N and P and biomass
282 production in different ecosystems, to get a better understanding of the potential biomass
283 response to the high N and P deposition that is occurring world-wide (Schreeg et al., 2014; Tang
284 et al., 2018).

285 In the present study, soil microbial biomass C, N, and P, and soil respiration were
286 positively associated with leaf P and leaf N/P ratio, but negatively with leaf N. The negative
287 association of microbial biomass C, N, and P with leaf N might be due to plant-microbe
288 competition for N in temperate forest, as suggested in other studies reporting a negative
289 association of leaf N with bacterial abundance (Pei et al., 2016). However, positive relationships
290 between microbial biomass C or N and leaf N have also been reported (Xue et al., 2014). These
291 contrasting results might be because plants with high leaf N need a rapid nutrient trade-off to
292 fulfil their physiological metabolism. Therefore, leaf N is affected by soil micro-organisms with
293 the ability to compete for soil C and N.

294 We found a negative association between soil respiration and leaf N, which might be due
295 to differences in litter quality, as suggested by other studies (Jiang et al., 2017; Han et al., 2015).

296 In contrast, several studies have found a positive relationship between plant diversity, microbial
297 biomass, and respiration (Bardgett et al., 1999; Liu et al., 2017; Chen et al., 2019; Chen and
298 Chen, 2019). Overall, our results indicate that leaf traits are significant drivers of forest above-
299 and below-ground dynamics, as they are the major contributor to acquisition and conservation
300 trade-offs.

301

302 **4.2 Soil characteristics and ecosystem properties**

303 Soil properties explained in total 60% of the variation in above-ground biomass (pH, available P,
304 total N, total C) and 56% of the variation in soil organic C (pH, total C). This is consistent with
305 previous findings for spruce and pine forests, with global forest data (Paré and Cleve, 1993;
306 Zhou and Dean, 2004; Ordoñez et al., 2009). In other studies, soil, climate, landscape, and space
307 together explained 61% of the variation in above-ground biomass in rain forest and 69.2% in
308 spruce forest (Cao et al., 2019; Santiago-García et al., 2019). Our results also indicated that
309 above-ground species of tree is important for above-ground and soil organic C stocks, explaining
310 30% and 10% of variation, respectively. This is consistent with previous findings for above-
311 ground biomass in a *Pinus kesiya* forest and soil organic C in pine, oak, conifer, and deciduous
312 forests (Cha et al., 2019; Li et al., 2018). The importance of plant species for soil properties has
313 been shown in several studies. At natural and afforested sites in China, the C, N, and P content in
314 both plants and soils are strongly influenced by plant species (Bai et al., 2019), while in northern
315 Iran, soil C and N vary between forests dominated by different tree species (Kooch et al., 2019).
316 In addition, in the present study, plot explained 10% of variation in above-ground biomass and
317 34% of variation in soil organic C. Thus, tree species used for afforestation, restoration, and

318 commercial forestry need to be selected with care, as they can influence the climate change
319 mitigation potential of forests in terms of C stocks in biomass and soils.

320 This study revealed that soil microbial biomass C and N are influenced by available P
321 and soil total C and N, which is consistent with previous findings for Indian *Pinus kesiya* forest
322 and tropical wet evergreen forests (Arunachalam et al., 1996; Barbhuiya et al., 2004). Many
323 previous studies report that microbial biomass C and N show significant correlations with soil
324 total N and pH (Kara and Bolat, 2008; Yang et al., 2010; Ravindran and Yang, 2015; Legay et
325 al., 2016, 2014; Liu et al., 2019). In our case, only soil microbial N was affected by soil pH,
326 while soil microbial biomass C was mainly affected by above-ground tree species. This is
327 consistent with findings for central Himalayan forests (Bargali et al., 2018).

328 In the present study, soil microbial biomass P was influenced by soil properties (pH,
329 available P, total N). A significant contribution of available soil P to microbial biomass P has
330 also been reported in other studies (Wang et al., 2004; Xue et al., 2014). Soil microbial biomass
331 P was influenced by above-ground tree species in our study. As in previous studies (Chen et al.,
332 2019; Nguyen and Marschner, 2017), the variation in soil respiration was mainly explained by
333 total C.

334

335 **5. Conclusions**

336 In this study, leaf traits and soil characteristics made a significant contribution to explaining the
337 variation in ecosystem processes and functioning in temperate forest in the Indian Himalayas.
338 While there is considerable variation between biomes globally, the results confirm findings in
339 previous large-scale studies of a general relationship between leaf P and N and biomass
340 production in different forest types. Therefore, use of existing broad-scale patterns of leaf traits

341 for modeling biomass and soil properties in relatively understudied areas, such as the Himalayas,
342 is justified. However, leaf traits had large impacts on biomass production and soil C stocks, so
343 more studies are needed to gain a better understanding of functional trait variations among
344 plants/forest types and the links to soil properties. Choice of tree species used for afforestation,
345 restoration, and commercial forestry also needs to be considered, as it can influence the climate
346 change mitigation potential of forests in terms of C stocks in biomass and soils.

347

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352

353 **Additional information**

354 Supplementary material accompanies this paper at (link).

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Table 1. Characteristics of dominant tree species selected for study in temperate forest in the central Himalayan region of Uttarakhand, India

Species	Family	Leaf habit	Leaf type	Tree density (tree ha ⁻¹)*	DBH (cm)*	IVI
Deciduous						
<i>Aesculus indica</i> , Colebr	Hippocastanaceae	D	B	240	38.99 ± 2.22	27.50
<i>Pyrus pashia</i> , Buch.Hemex D.Don	Rosaceae	D	B	160	25.66 ± 3.33	28.94
<i>Toona ciliata</i> , R.	Meliaceae	D	B	157	33.45 ± 2.98	14.44
Evergreen						
<i>Abies pindrow</i> , Spach Ham	Pinaceae	E	N	220	35.89 ± 3.23	50.82
<i>Cedrus deodara</i> , Loud	Pinaceae	E	N	250	35.78 ± 1.23	41.98
<i>Cupressus torulosa</i> , D. Don	Cupressaceae	E	S	230	29.76 ± 0.12	39.37
<i>Euonymus pendulous</i> , Wall	Celastraceae	E	B	123	23.44 ± 1.11	27.84
<i>Quercus leucotrichophora</i> , A.Comm	Fagaceae	E	B	250	31.12 ± 5.54	31.15
<i>Rhododendron arboreum</i> , Smith	Ericaceae	E	B	260	26.93 ± 6.69	20.62
<i>Pinus wallichiana</i> , Jackson	Pinaceae	E	N	210	34.67 ± 1.45	17.33

E - Evergreen; D - Deciduous; N - Needle; B - Broadleaf; S - Scaly; DBH - Diameter at breast height *Own survey, IVI -Importance value index.

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Table 2. Description of plant functional traits and related functions, based on the literature*

Trait	Abbreviation	Description	Function
Leaf area	LA (cm ²)	One-sided projected surface area of the leaf	Evapotranspiration, canopy light interception, photosynthetic efficiency, relative growth rate
Specific leaf area	SLA (cm ² g ⁻¹)	One-sided area of the fresh leaf divided by its oven-dry mass	Photosynthesis, leaf longevity, relative growth rate
Leaf water content	LWC (%)	Proportion of leaf water content related to the maximum water content that can potentially be achieved by the leaf	Plant water status, flammability, water use efficiency
Leaf dry matter content	LDMC (%)	Oven-dry mass of a leaf divided by its water-saturated fresh mass	Leaf physical resistance, leaf lifespan, stress tolerance
Leaf carbon content	LCC (%)	Total amount of carbon per unit of dry leaf mass	Gas exchange, water use efficiency
Leaf nitrogen content	LNC (%)	Total amount of nitrogen per unit of dry leaf mass	Light capture, photosynthetic rate
Leaf phosphorus content	LPC (%)	Total amount of phosphorus per unit of dry leaf mass	Photosynthesis, nutritional quality
Leaf C/N	LCC/LNC	Ratio between leaf carbon and nitrogen	Life history, availability of carbon and nitrogen
Leaf N/P	LNC/LPC	Ratio between leaf nitrogen and phosphorus	Availability of nitrogen and phosphorus, life history

*Wright et al., 2004; Pérez-Harguindeguy et al., 2013; Díaz et al., 2016.

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633 **Table 3 (a).** Residual Maximum Likelihood (REML) for each of the ecosystem properties
634 studied in temperate forest in the central Himalayan region of Uttarakhand, India. For trait
635 abbreviations, see Table 2
636 **Linear mixed model**

Response variable	Fixed effect (Plant trait)	Estimate	Variation explained (in %)	Random effect (Soil property, species and plot)	Std. dev	% of variation explained
Above-ground biomass	SLA	79.75 (57.47)	0.01	Available P	7.09	16.25
	LWC	52.99 (211.44)	0.12	Total N	1.52	7.50
	LCC	-1632.18 (2743.65)	20.37	Total C	3.67	4.36
	LNC	2113.39 (3119.54)	26.34	Plot	1.74	9.79
	LPC	-2094.89 (3125.86)	26.44	pH	9.97	32.15
	LNC/LPC	-2094.39 (3140.29)	26.69	Species	9.62	29.95
Soil organic carbon	SLA	0.04 (0.11)	0.00	Available P	0.00	0
	LWC	0.11 (0.18)	0.01	Total N	0.00	0
	LCC	-0.25 (0.26)	0.02	Total C	2.70	36.44
	LNC	6.86 (10.42)	33.21	Plot	8.30	34.21
	LPC	-6.76 (10.43)	33.27	pH	6.37	20.18
	LNC/LPC	-6.70 (10.46)	33.47	Species	1.35	9.14
Soil microbial biomass carbon	SLA	262.18 (234.78)	0.00	Available P	55	1.02
	LWC	43.11 (406.18)	0.01	Total N	249.3	20.92
	LCC	-446.36 (571.71)	0.02	Total C	302.77	30.86
	LNC	-3264.85 (22531.64)	33.20	Plot	364.96	44.84
	LPC	3112.5 (22560.71)	33.29	pH	0	0.00
	LNC/LPC	3372.78 (22615.99)	33.45	Species	83.69	2.36
Soil microbial biomass nitrogen	SLA	-2.21 (7.42)	0.00	Available P	7.15	19.17
	LWC	-11.61 (13.45)	0.01	Total N	1.75	11.46
	LCC	28.67 (19.1)	0.02	Total C	7.64	21.86
	LNC	-67.81 (726.21)	33.22	Plot	2.86	32.98
	LPC	51.59 (726.76)	33.27	pH	6.23	14.53
	LNC/LPC	55.5 (728.61)	33.44	Species	0	0.00
Soil microbial biomass phosphorus	SLA	-5.04 (1.98)	0.00	Available P	3.06	49.37
	LWC	2.58 (4.08)	0.01	Total N	1.34	9.47
	LCC	1.7 (5.48)	0.02	Total C	0	0.00
	LNC	-135.79 (222.15)	33.21	Plot	0	0.00
	LPC	136.08 (222.38)	33.28	pH	2.07	22.68
	LNC/LPC	136.05 (222.97)	33.46	Species	1.87	18.47
Soil respiration	SLA	-0.45 (0.93)	0.00	Available P	0	0.00
	LWC	-0.2 (1.69)	0.01	Total N	0	0.00
	LCC	0.28 (2.35)	0.02	Total C	1.65	69.72
	LNC	-39.27 (95.78)	33.19	Plot	1.09	30.28
	LPC	40.86 (95.92)	33.29	pH	0	0.00
	LNC/LPC	39.08 (96.17)	33.47	Species	0	0.00

637 Values within brackets are the standard error.

638 **Table 3 (b).** Model statistics for the null and full models

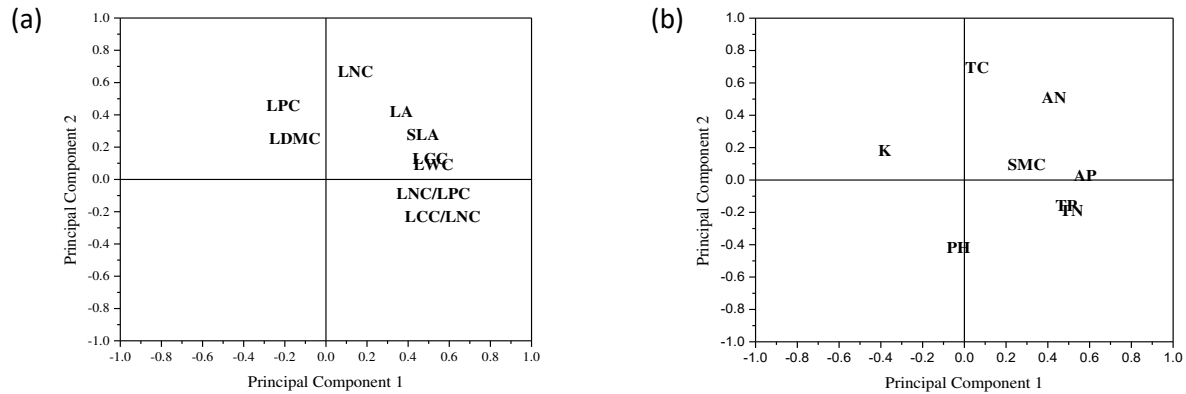
Model Parameter					
Null Model					
Response variables	df	AIC	log Lik	Deviance	Chi-square (P-value)
Above-ground biomass	8	741.20	-362.60	725.20	
Soil organic carbon	8	291.57	-137.78	275.57	
Soil microbial biomass carbon	8	3079.9	-1532.0	3063.9	
Soil microbial biomass nitrogen	8	1848.1	-916.04	1832.1	
Soil microbial biomass phosphorus	8	1424.8	-704.40	1408.8	
Soil respiration	8	1141.8	-562.91	1125.8	
Full Model					
Above-ground biomass	15	752.43	-361.22	722.43	2.76 (0.90)
Soil organic carbon	17	295.62	-130.81	261.62	13.94 (0.12)
Soil microbial biomass carbon	17	3065.4	-1515.7	3031.4	32.50 (0.00 ^{***})
Soil microbial biomass nitrogen	17	1827.2	-896.61	1793.2	38.86 (1.21 ^{***})
Soil microbial biomass phosphorus	17	1391.1	-678.57	1357.1	51.66 (5.23 ^{***})
Soil respiration	17	1095.6	-530.78	1061.6	64.25 (2.01 ^{***})

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645 **Figure 1.** Principal component analysis plot of (a) nine plant traits and (b) eight soil properties
 646 studied in temperate forest in the central Himalayan region of Uttarakhand, India.

647 *Plant traits:* leaf area (LA), cm²; specific leaf area (SLA), cm² g⁻¹; leaf dry matter content
 648 (LDMC), %; leaf water content (LWC), %; leaf carbon content (LCC), %; leaf nitrogen content
 649 (LNC), %; leaf phosphorus content (LPC), %; LCC/LNC ratio; LNC/LPC ratio.

650 *Soil properties:* pH; soil moisture content (SMC), %; total carbon (TC), %; available nitrogen
 651 (AN), μg/g; total nitrogen (TN), %; available phosphorus (AP), μg/g; total phosphorus (TP), %;
 652 potassium (K), μg/g.

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Supplementary materials: Relative contribution of plant traits and soil properties to the functioning of a temperate forest ecosystem in the Indian Himalayas

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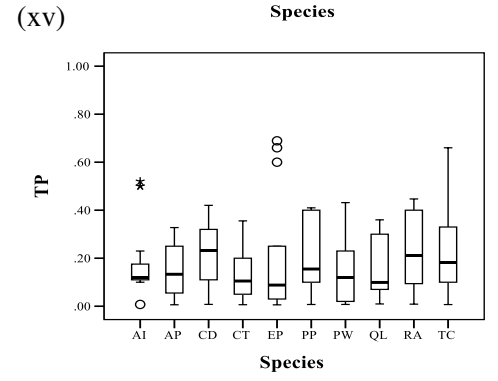
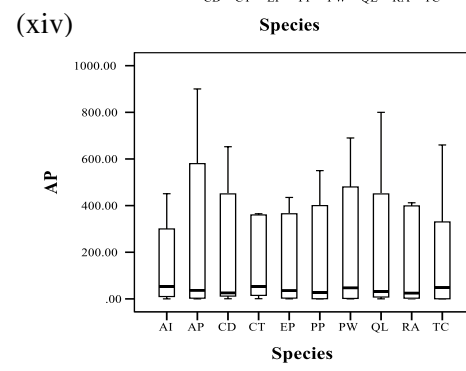
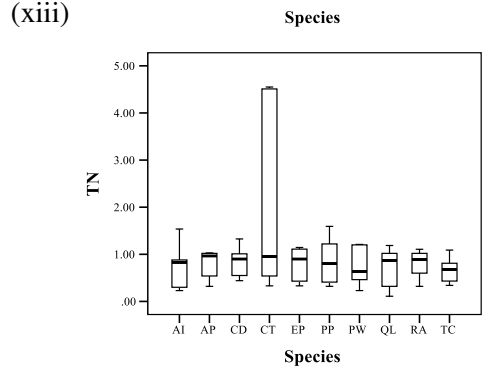
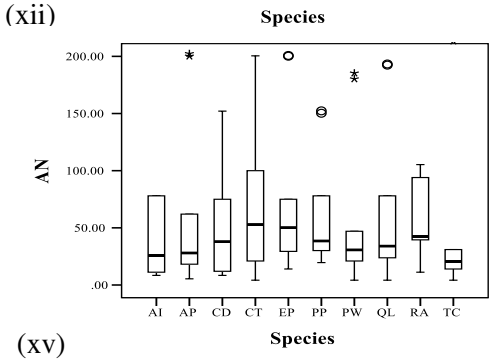
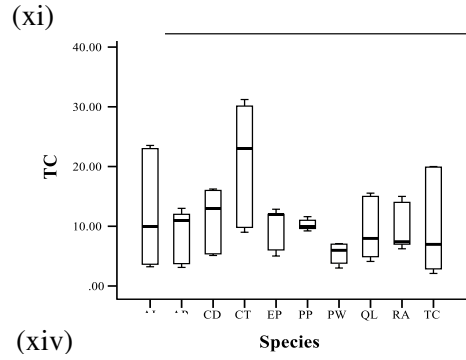
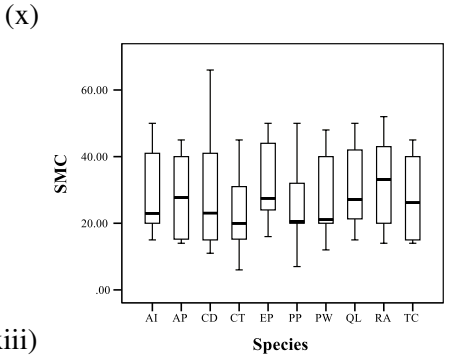
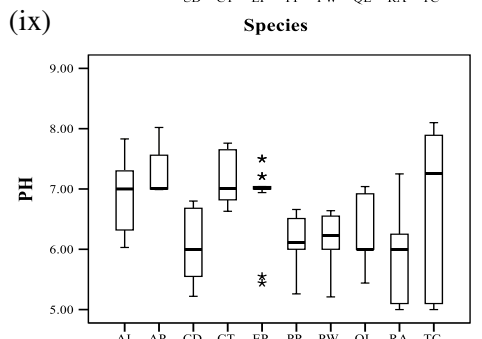
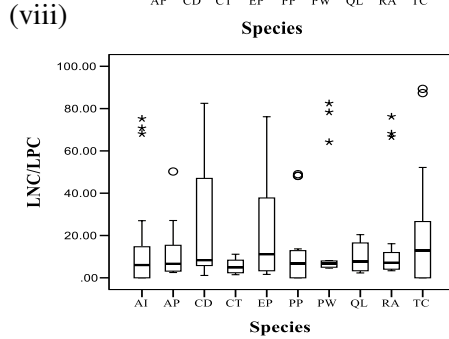
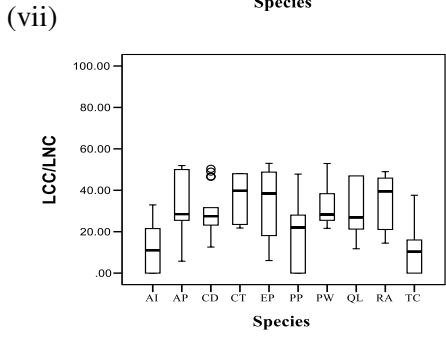
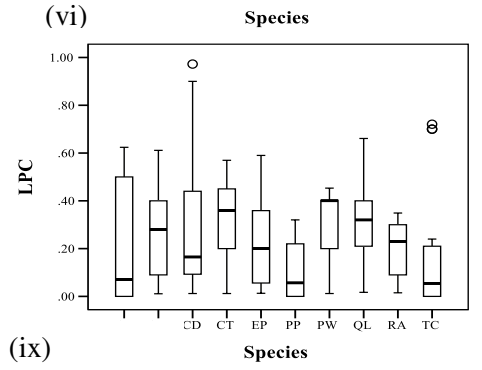
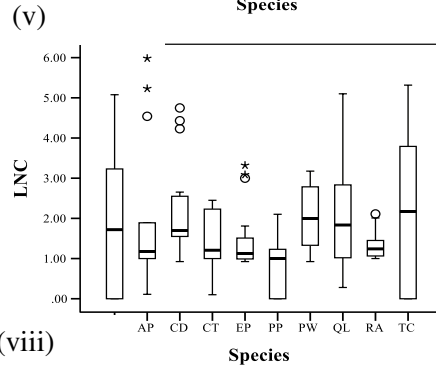
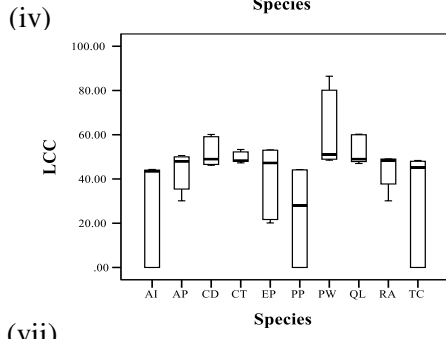
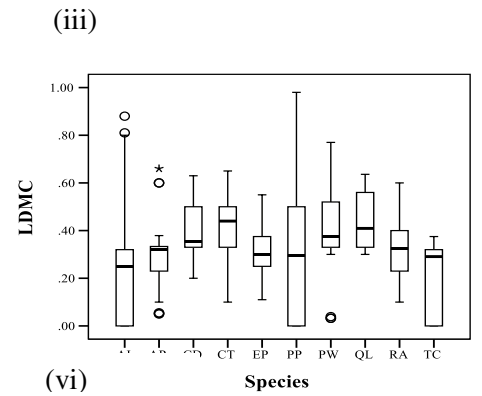
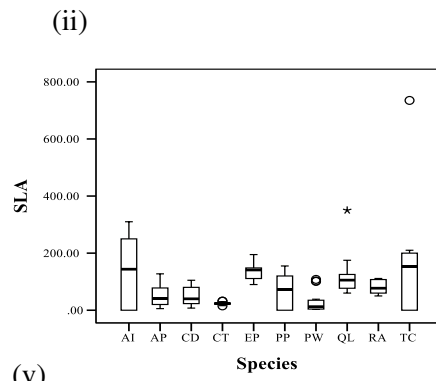
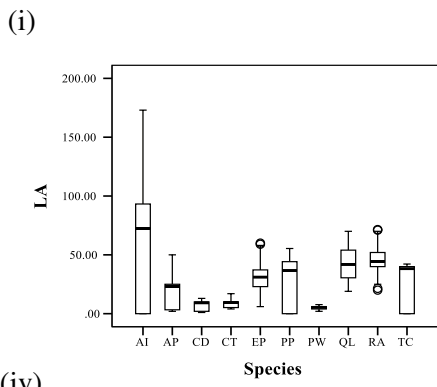
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Table S1: Volume equations used for aboveground biomass calculation for different tree species in the temperate forest, central Himalayan region of Uttarakhand, India

S. no	Dominant tree species	Volume equation
	Deciduous	$\sqrt{V} = 0.220191 + 3.923711 D - 1.117475 \sqrt{D}$ (L)
1	<i>Aesculus indica</i> , Colebr	
2	<i>Pyrus pashia</i> , Buch.Hemex D.Don	$V = 0.01284 + 0.2138 D^2 H$ (G)
3	<i>Toona ciliata</i> , R.	$V = 0.01284 + 0.2138 D^2 H$ (G)
	Evergreen	
4	<i>Abies pindrow</i> , Spach Ham	$V = 0.17507 + 0.22606 D^2 H$ (G))
5	<i>Cedrus deodara</i> , Loud	$V = -0.0789 + 0.2836 D^2 H$ (G)
6	<i>Cupressus torulosa</i> , D. Don	$V = 0.01284 + 0.2138 D^2 H$ (G)
7	<i>Euonymus pendulous</i> , Wall	$V = 0.01284 + 0.2138 D^2 H$ (G)
8	<i>Quercus leucotrichophora</i> , A.Comm	$V = 0.014796 + 0.319061 D^2 H$ (G)
9	<i>Rhododendron arboreum</i> , Smith	$V = 0.06007 - 0.21874V D + 3.63428 D^2$ (L)
10	<i>Pinus wallichiana</i> , Jackson	$V = 0.223139 + 2.35096 D + 11.90669 D^2$ (L)

Note: L = local volume equations (they have limited application for a forest or small locality and are based only on diameter at breast height); G= general volume equations (they are more broadly based and cover the full distribution of the species); \sqrt{V} or V = volume of tree (Mg ha⁻¹); \sqrt{D} or D = diameter of tree at breast height; H = height of tree. References: FSI (1996)



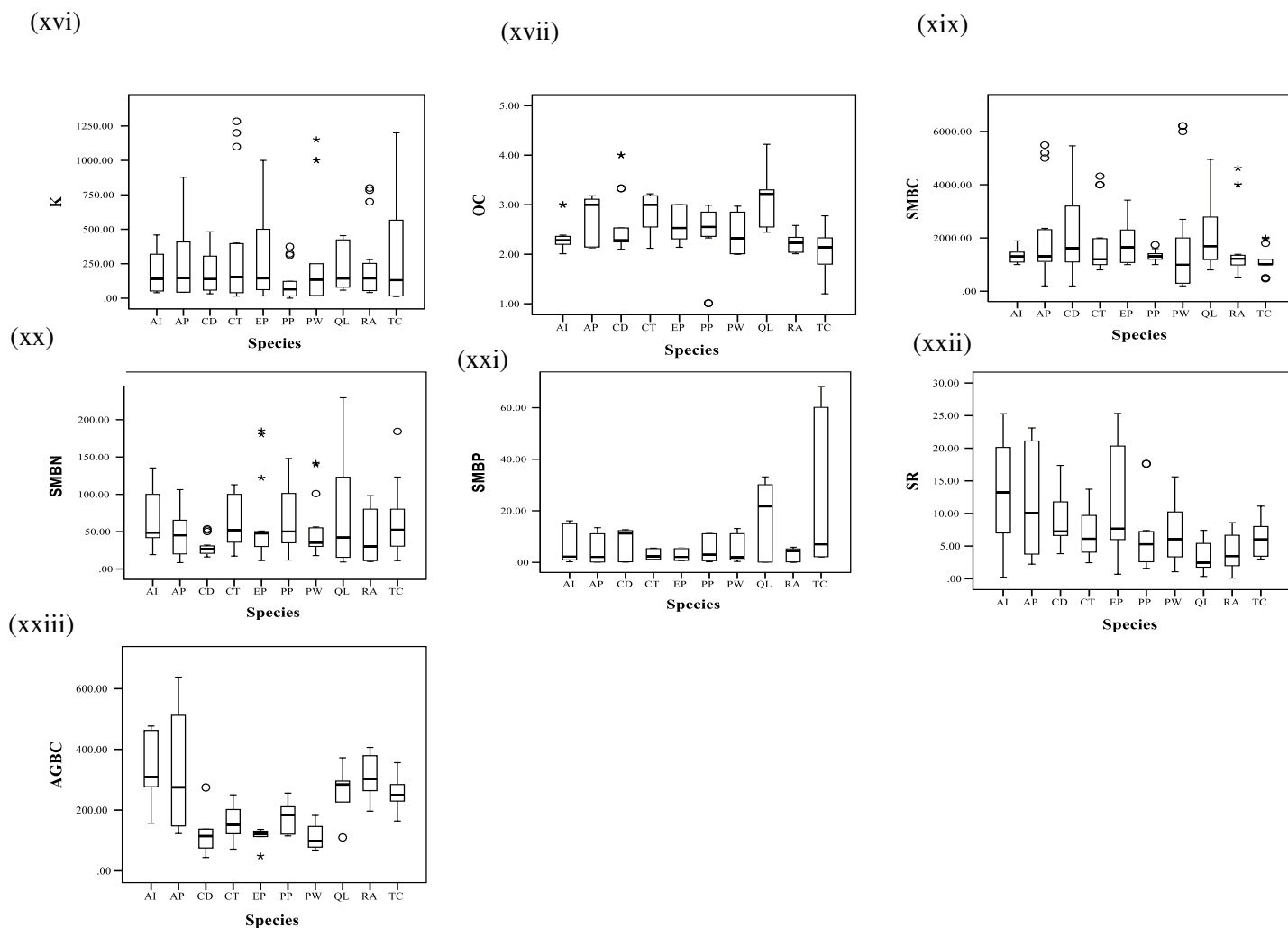


Figure S1. Distribution of values of plant traits, soil properties and ecosystem services observed for temperate forest tree species in the temperate forest, central Himalayan region of Uttarakhand, India: *Aesculus indica* (AI), *Abies pindrow* (AP), *Cedrus deodara* (CD), *Cupressus torulosa* (CT), *Euonymys pendulous* (EP), *Purys pashia* (PP), *Pinus wallichiana* (PW), *Quercus leucotrichophora* (QL), *Rhododendron arboretum* (RA), and *Toona ciliata* (TC).

Plant traits: (i) Leaf area (LA), cm²; (ii) specific leaf area (SLA), cm² g⁻¹; (iii) leaf dry matter content (LDMC), %; (iv) leaf water content (LWC), %; (v) leaf carbon content (LCC), %; (vi) leaf nitrogen content (LNC), %; (vii) leaf phosphorus content (LPC), %; (viii) LCC/LNC ratio; (xi) LNC/LPC ratio.

Soil properties: (x) pH; (xi) soil moisture content (SMC), %; (xii) total carbon (TC), %; (xiii) available nitrogen (AN), μg/g; (xiv) total nitrogen (TN), %; (xv) available phosphorus (AP), μg/g; (xvi) total phosphorus (TP), %; (xvii) potassium content (K), μg/g.

Ecosystem services: (xviii) soil organic carbon (OC), %; (xix) soil microbial biomass carbon (SMBC), μg/g; (xx) soil microbial biomass nitrogen (SMBN), μg/g; (xxi) soil microbial biomass phosphorus (SMBP), μg/g; (xxii) soil CO₂ respiration (SR), μmol mol⁻¹ s⁻¹ and (xxiii) aboveground biomass carbon (AGBC), Mg C ha⁻¹. The centre line in each plot indicates the median and the upper and lower box height indicates the inter-quartile range.