1	Chronic disturbance of moist tropical forests favours deciduous over evergreen tree
2	communities across a climate gradient in the Western Ghats
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19	ABSTRACT
20	It is well established that climatic factors such as water stress and chronic anthropogenic
21	disturbances such as biomass extraction influence tropical forest tree community structure,
22	richness and composition. However, while the standalone effects of these two drivers on plant
23	communities are well-studied, their interactive effects are not. Moist tropical forests in India's

24 Western Ghats face a dual threat from increasingly erratic precipitation (and consequent water 25 stress), and an intensifying anthropogenic footprint. Here, we sampled 120 tree plots (0.05 ha 26 each) across forests with varying histories of biomass extraction and a gradient in climate water deficit (CWD, a proxy for water stress) within a 15,000 km² landscape in the northern Western 27 28 Ghats and examined whether and how disturbance history modulates relationships of tree 29 community structure and composition with climate. As expected, tree species richness increased 30 with decreasing water stress in less- and historically-disturbed forests but remained low in 31 repeatedly-disturbed forests. The increase in evergreen species richness with decreasing water 32 stress was far slower in repeatedly-disturbed forests than other categories, and the relative 33 abundance of evergreens in the repeatedly-disturbed forests (25%) was half that in less-disturbed 34 forests (50%) of comparable water stress in the driest parts. Overall, we show that disturbance 35 can amplify threats from climate change to wet forest-associated evergreen tree species, many of 36 which are threatened, while benefiting more widely distributed dry forest-associated deciduous 37 species. In the northern Western Ghats, where much of the remaining forest cover is disturbed 38 and dominated by deciduous tree species, the persistence of evergreen tree flora hinges on 39 protecting existing evergreen forest patches from future disturbances and restoring locally 40 appropriate evergreen species in secondary forests.

41 **INTRODUCTION**

42 Plant community assembly is the culmination of abiotic and biotic factors acting over different 43 spatial and temporal scales (Ramos et al., 2023; Zheng et al., 2022). Although climate is the 44 major factor shaping plant diversity and composition (Harrison et al., 2020), in the Anthropocene, human-mediated disturbances act in conjunction with such environmental factors 45 46 and contribute to altered tree richness, abundance and composition (Bentsi-Enchill et al., 2022), 47 with cascading impacts on ecosystem function (Hooper et al., 2012). However, the effects of anthropogenic disturbances and environmental factors on plant community assembly are often 48 49 studied in isolation. Few studies have examined their combined effects on plant diversity and 50 structure, and those that have, are primarily focused on seasonally dry tropical forests (Ramos et 51 al., 2023; Rito et al., 2017; Zorger et al., 2019). Despite wet tropical forests (forests receiving > 52 2000 mm of annual rainfall) harbouring more than half of the global terrestrial biodiversity 53 (Malhi et al., 2014) and being increasingly threatened by habitat degradation and biodiversity 54 loss (Saatchi et al., 2021), studies on the interactive effects of anthropogenic and environmental 55 factors on plant communities have been overlooked in these sensitive ecosystems. 56 For plants, annual precipitation is an important factor influencing plant richness and

composition across large environmental gradients (Harrison et al., 2020; Krishnadas et al., 2016).
Besides annual precipitation, water stress during dry months is critical in determining species
distributions (Esquivel-Muelbert et al., 2017), thereby influencing vegetation types (Hirota et al.,
2011). On almost half of the Earth's terrestrial surface, climate change is expected to lead to less
water availability (McLaughlin et al., 2017), thereby greater water stress. Reduced water
availability may limit species distributions (Saiter et al., 2016) and induce deciduousness in the
vegetation, particularly in wetter, evergreen forests (Saiter et al., 2016; Seiler et al., 2014),

64 thereby driving tropical biome transitions, like those between wet and dry forests (Dexter et al., 65 2018). These transition zones represent unstable states where the vegetation is most sensitive to climatic disturbances (Hirota et al., 2011). Like climate, anthropogenic disturbances can mediate 66 67 shifts in vegetation types. Although deforestation is a major global threat to biodiversity, tropical forests are increasingly imperilled by chronic anthropogenic disturbances like selective logging 68 69 and fuelwood extraction (Barlow et al., 2016). Despite research indicating that the interactive 70 effects of climatic and anthropogenic factors can have significant consequences on the resilience 71 of plant communities and affect vegetation types (Hirota et al., 2011; Shivaprakash et al., 2018), 72 studies explicitly testing for these effects are sparse. The need for assessing the interactions 73 between climate and anthropogenic factors has been highlighted (Krishnadas et al., 2021). While 74 some studies have investigated the interactive effects of climatic factors like precipitation and 75 chronic anthropogenic disturbances on plant diversity and structure (Rito et al. 2017; Ramos et 76 al. 2023), no studies have examined the effect of these interactive effects on vegetation 77 transitions. Although studies have suggested that fire mediates the transformation of forests to 78 drier vegetation types (savannization) (Sansevero et al., 2020; M. Wang et al., 2023), the 79 implications of chronic anthropogenic disturbances (hereafter, CAD) and environmental factors 80 on vegetation type transitions are less explored.

81 The Western Ghats-Sri Lanka biodiversity hotspot harbours a high diversity of vascular 82 plants, many of which are endemic to the region (Gunawardene et al., 2007). It has lost a 83 substantial extent of its original vegetation cover (<30% remaining) due to human activities 84 (Mittermeier et al., 2011; Myers et al., 2000). The region exhibits high levels of topographic and 85 climatic heterogeneity, creating suitable habitats to support diverse vegetation types along a 86 gradient of deciduous to evergreen forests, each of which harbours unique fauna (Gunawardene

et al., 2007). Woody plant diversity in the Western Ghats exhibits a clear latitudinal gradient 87 88 with decreasing plant richness in northern portions primarily driven by the past geo-climatic history and niche conservatism (Gopal et al., 2023). Additionally, the topography of the 89 90 mountain ranges leads to a distinct gradient in rainfall from the low elevations to the mountain 91 crest, wherein precipitation increases towards the higher elevations (Venkatesh et al., 2021). 92 Distinct forest types are associated with different rainfall and elevation bands (Joseph et al., 93 2012). These varying rainfall patterns give rise to water availability gradients, which are known to be important drivers of tree diversity and structure (Terra et al., 2018) and are responsible for 94 95 a wide diversity of trees and vegetation types in the Western Ghats of India (Joseph et al., 2012; 96 Page & Shanker, 2018).

97 Despite the high biodiversity in this region, the forests in the Western Ghats are 98 threatened by a range of anthropogenic disturbances that have led to the loss of 35.3% of its 99 forest cover from the 1920s to 2013 (Reddy et al., 2016). The forests here are at an elevated risk 100 of degradation as they have the third highest human population density among all global hotspots 101 (Cunningham & Beazley, 2018). This is particularly true for the northern Western Ghats, where 102 the privately-owned forests are periodically clear-felled (every ten years) to harvest fuelwood, 103 resulting in CAD. Apart from the repeatedly disturbed privately-owned forests, the northern 104 Western Ghats also harbour government-owned Reserved Forests and Protected Areas, which 105 were historically disturbed. There are also sacred groves, many of which harbour among the least 106 disturbed forest habitats in the region. These disturbance and water availability gradients in the 107 northern Western Ghats make it a suitable system to study the interplay of climate and CAD on 108 plant vegetation types.

109	Given this background, we investigated the effects of CAD and climate on the diversity
110	and structure of plant communities. To do this, we sampled 120 vegetation plots across three
111	forest categories with varying degrees of protection and along a precipitation gradient. We
112	specifically asked the following questions: (1) How does tree composition differ across forest
113	categories? (2) How do climatic water deficit (CWD) and CAD impact the vegetation structure
114	and overall tree species richness? (3) How do CWD and CAD affect the proportion of evergreen
115	trees and richness of evergreen and deciduous trees? We expected that filtering mediated by
116	human disturbance and climate would result in differences in tree composition and diversity. We
117	expected a higher proportion and richness of evergreen trees to be found in the wetter (low
118	CWD) and cooler regions in the high-elevation forests, as compared to the drier (high CWD),
119	low-elevation forests. Moreover, we predicted that repeated chronic anthropogenic disturbance
120	would further lower evergreen tree richness and proportion, leading to a more deciduous
121	vegetation type.

122

123 METHODS

124 Study Area

125 We conducted the study in the south-western part of Maharashtra in India (15.72–17.74°N;

126 73.29–74.19°E) between October 2022 and March 2023. The study site forms the northern part

127 of the Western Ghats-Sri Lanka Biodiversity Hotspot, which is classified as one of the "hottest"

128 biodiversity hotspots globally, based on the high degree of endemism and levels of

129 anthropogenic threats (Myers et al. 2000). The climate here is tropical, with average annual

rainfall ranging from 2150–7450 mm, and the annual temperature varying between 16–35°C (Jog

131 2009).

We sampled sites experiencing varying levels of disturbance across an elevational 132 133 gradient, ranging from 8 to 1054 m a.s.l. (Fig. S1). The sites were classified as less-disturbed, 134 historically-disturbed and repeatedly-disturbed. For the less-disturbed sites, we sampled the 135 sacred groves. The landscape harbours several sacred groves, locally known as "Devrais", which 136 are protected by the local villages and home to relatively less disturbed patches of forests (Gadgil 137 & Vartak, 1976). They provide refuge to many endemic, evergreen and medicinal plants 138 (Blicharska et al., 2013; Kulkarni et al., 2018). Since some sacred groves experience 139 opportunistic fuelwood and timber collection, these sites were classified as less-disturbed. We 140 classified the government-managed forests (Protected Areas (PAs) and Reserved Forests (RFs)) 141 as historically disturbed sites. The PA sites were spread across the Sahyadri Tiger Reserve and 142 Radhanagari Wildlife Sanctuary, and the RF sites were spread across Chiplun and Sawantwadi 143 Forest Divisions. The PAs were designated in the late 1950s (Radhanagari) and mid-1980s 144 (Sahyadri), and these sites have a long human-use history for shifting cultivation and have likely 145 been clear-felled in the past (Chandran, 1997; Ghate et al., 1998), unlike the sacred groves. The 146 existing stunted evergreen forests here result from forest recovery after clear felling (Ghate et al. 147 1998). Since there are no PAs in lower elevations, all the PA sites were located in relatively 148 higher elevations (584–1012 m a.s.l). The elevation range of RF sites ranged between 26–386 m 149 a.s.l. While no form of resource utilisation by humans is permitted in PAs, some forms of 150 resource use, like cattle grazing and deadwood collection, are still allowed in RFs. We classified 151 the private forests as repeatedly-disturbed forests. The private forests sampled in the region 152 experience the highest level of chronic anthropogenic disturbance among the three forest 153 categories as people clear fell these forests every five to ten years to sell fuelwood (Kulkarni and 154 Mehta, 2013). The prevailing forest type here has been reported to be tropical moist deciduous in

the lower elevations, and semi-evergreen to evergreen in the higher elevations, with stunted
evergreen trees being present after recovery from clear-felling (Champion & Seth, 1968; Ghate
et al. 1998). The dominant tree families in the high-elevation PAs are Melastomataceae and
Myrtaceae (Joglekar et al., 2015; Kanade et al., 2008). While there is a considerable amount of
information from the higher-elevation protected and non-protected forests (Joglekar et al. 2015;
Kulkarni et al. 2018), there is very little information on vegetation type and composition in the
low-elevation forests of the northern Western Ghats.

162 Vegetation Sampling

163 We conducted the field study in an approximately 15,000 km² landscape in the northern Western

164 Ghats (Fig. S1). To capture the gradients of CAD and climate, we laid $12050 \times 10m^2$ plots that

165 were evenly distributed across less-disturbed sacred groves (n=40), historically-disturbed

166 government-owned forests (Reserved Forests and Protected Areas) (n=40), and repeatedly-

167 disturbed privately-owned forests (n=40), across low and high elevations. Henceforth, we will

168 refer to these three categories of forests as less-disturbed, historically-disturbed and repeatedly-

169 disturbed forests. In each plot, we recorded the identity, girth and height of all woody stems ≥ 10

170 cm Girth at Breast Height (GBH). We also recorded the number of cut stems in each plot as an

171 indicator of human disturbance. All trees were classified as evergreen or deciduous using

172 regional floras and expertise within the team (NP). We excluded climbers from further analyses.

173 Analysis

174 All the analysis was carried out in R ver. 4.3.1(R Core Team, 2023)

175 Tree composition across land-use types

176 To find out the difference in the composition of trees across the six forest categories in low- and

177 high-elevations, we used 3-D non-metric multidimensional scaling (NMDS) with Bray-Curtis

178 dissimilarity metric using the R package 'vegan' (Oksanen et al., 2022) and 'vegan3d' (Oksanen 179 et al., 2018). We used 3-D NMDS since 2-D NMDS had stress values greater than 0.2. The 180 difference in species composition among the categories was tested using the 'ANOSIM' or 181 analysis of similarities function in the R package 'vegan', along with a permutation test 182 (permutations = 999).183 Influence of CAD and CWD on species richness 184 We examined the influence of climate and CAD on the species richness per plot. We used the 185 predictor, CWD, to account for the climatic gradient across our study site. We obtained the CWD 186 values for each vegetation plot from a global gridded layer available at http://chave.ups-187 tlse.fr/pantropical_allometry.htm#CWD (Chave et al., 2014). CWD is a measure of the water 188 stress experienced by plants in the dry months. It is measured as the difference between 189 precipitation and evapotranspiration in dry months when evapotranspiration exceeds 190 precipitation (Krishnadas et al. 2021). The values are always negative, and a less negative CWD 191 indicates a higher water availability for plants. It has been known to play a key role in assessing 192 the sensitivity of vegetation to drought (Vicente-Serrano et al., 2013) and transitions of 193 evergreen forests to dry deciduous forests in Bolivia (Seiler et al., 2014). In the Western Ghats 194 too, studies have shown CWD to influence plant diversity (Krishnadas et al., 2021, Gopal et al., 195 2023); however, the interactive effects between CWD and CAD are not known. In our area of interest, elevation and CWD are strongly positively correlated (Spearman's $\mathbb{Z} = 0.79$; p < 0.001), 196 197 indicating that high-elevation sites had lower water deficit than low-elevation sites. We used 198 GLMs to evaluate the combined (interactive) effect of CWD and forest category on the observed species richness per plot with a Poisson error structure (since the response variable was not over-199 200 dispersed).

201 Influence of CAD and CWD on vegetation structure

202	We estimated the basal area of trees $(m^2 ha^{-1})$. We square root transformed the basal area since it
203	was not normally distributed (Shapiro-Wilk's Test; $p < 0.05$). We used the General Linear Model
204	(Gaussian error structure) to test the relationship between basal area, disturbance categories, and
205	CWD.

Influence of CAD and CWD on the proportion of evergreen trees and richness of evergreen and deciduous trees

- 208 We examined the influence of CWD and CAD on the proportion of evergreen trees and richness
- 209 of evergreen and deciduous trees per plot. We used GLMs to evaluate the combined (interactive)
- effect of CWD and forest category on the proportion of evergreen trees with a binomial error
- 211 structure and the richness of evergreen and deciduous trees with a Poisson error structure.

212 **RESULTS**

- 213 We identified 97% of the 7001 (\geq 10 cm GBH) individual trees across our 120 plots. Two
- 214 individuals were identified till the genus level. Six individuals that could not be identified were
- excluded from further analyses. We recorded 192 plant species (166 trees) from 52 families
- 216 (Table S1). The most speciose families were Fabaceae and Moraceae, accounting for 14.6% of
- all species. We recorded a total of 10,071 stems. The mean (\pm SE) stem density was 1678.5
- 218 (\pm 79.8) stems/ha, and the basal area was 32 (\pm 1.7) m²/ha. A summary of tree richness is
- 219 presented in Figure S3.

220 Woody plant composition across land-use types

- 221 The 'ANOSIM' analysis revealed significant differences in tree composition among the six
- different forest categories ($R_{anosim} = 0.44$, p = 0.001, stress = 0.17) (Fig. S2). The NMDS plots
- suggested that the tree composition in high-elevation forests tended to be different than low-

- elevation ones (Fig. S2). Even at a particular elevation-level, tree composition differed across
- disturbance categories.
- 226 Influence of CAD and CWD on species richness
- 227 The GLM result showed that the interaction between CWD and disturbance category had a
- significant effect on the species richness of trees (Fig. 1; Table S2). The observed species
- richness per plot increased with decreasing CWD for less-disturbed sites but was lower and
- tended to decrease with decreasing CWD for repeatedly-disturbed sites (Fig. 1; Table S2).

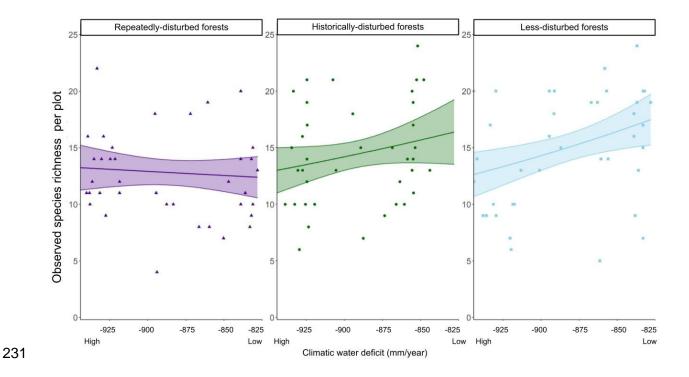


Figure 1. Plots showing the relationship between climatic water deficit and observed species
richness per plot across the three disturbance categories. The solid line with a band indicates the
best-fitted line for the Generalized Linear Model (Poisson error structure) with associated 95%
CI. Higher values of CWD (less negative) imply lower water stress experienced by plants.

237 Role of disturbance in influencing forest structure

238 The interaction between disturbance categories and CWD was significant for the basal area (R^2 =

239 0.51; Fig. 2; Table S3). Basal area increased with CWD for historically-disturbed and less-

240 disturbed sites but decreased with CWD for the repeatedly-disturbed site (Fig. 2). This indicates

that the trees are larger in wetter regions when chronic anthropogenic disturbance is lower.

However, in chronically-disturbed sites trees are smaller in wetter regions. A summary of basal

area, tree density, and canopy cover is provided in Table S4.

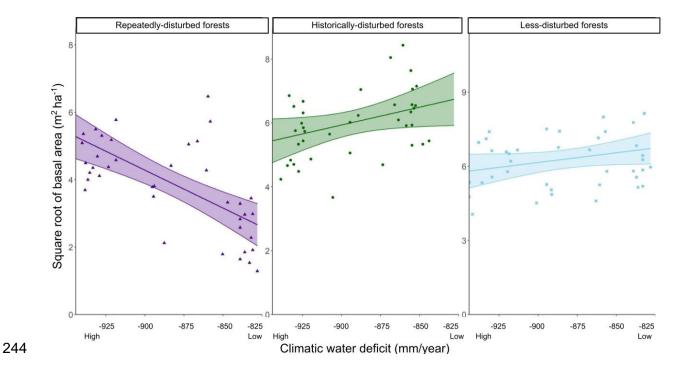


Figure 2. Plots showing the relationship between climatic water deficit and square root of basal
area per hectare, across the three disturbance categories. The solid line with a band indicates the
best-fitted line for the Generalized Linear Model (Gaussian error structure) with associated 95%
CI. Higher values of CWD (less negative) imply lower water stress experienced by plants.

249

Influence of CAD and CWD on the proportion of evergreen trees and richness of evergreen and deciduous trees

252 The GLM results showed that the interaction terms between CWD and disturbance category 253 significantly affected the proportion and richness of evergreen trees. The proportion of evergreen 254 trees per plot increased with the increase in CWD values (which implies less water deficit) and 255 this increase was the fastest in less-disturbed forests and slowest in repeatedly-disturbed forests (Pseudo $R^2 = 0.38$, Fig. 3; Table S5). In regions with low CWD values (indicating higher water 256 257 deficit i.e., drier areas), the predicted proportion of evergreen trees is more than 0.5 for less-258 disturbed forests but only 0.25 for repeatedly disturbed forests, indicating dominance of 259 deciduous trees in repeatedly-disturbed sites (Fig. 3). In regions with high CWD values 260 (indicating lower water deficit i.e., wetter areas) the predicted proportion of evergreen trees for 261 less-disturbed forests was close to 1 but for repeatedly-disturbed forests it was around 0.75. 262 Similarly, the increase in the number of evergreen tree species per plot with decreasing CWD 263 was very fast for less- and historically-disturbed sites but gradual for the repeatedly-disturbed sites, indicating that disturbance negatively impacted evergreen tree species richness (Pseudo R^2 264 265 = 0.33, Fig. 4A; Table S6). Consequently, the deciduous species richness per plot rapidly 266 decreased with decreasing CWD for less-disturbed and historically-disturbed sites but this decrease was significantly more gradual in repeatedly disturbed sites (Pseudo $R^2 = 0.36$, Fig. 4B; 267 268 Table S7).

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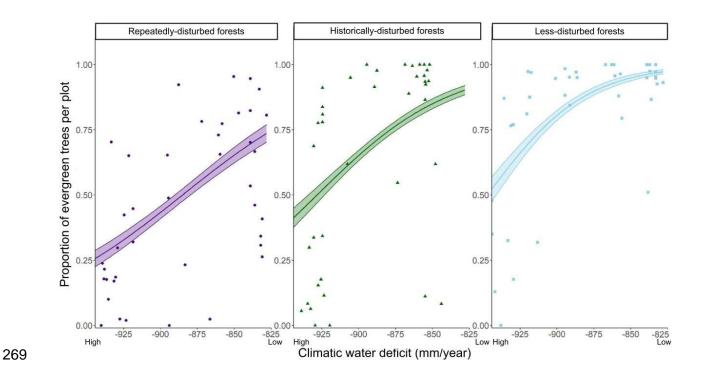


Figure 3. Plots showing the relationship between climatic water deficit and proportion of
evergreen tree individuals per plot across the three disturbance categories. The solid line with a
band indicates the best fitted line for the Generalized Linear Model (binomial error structure)
with associated 95% CI. Higher values of CWD (less negative) imply lower water stress
experienced by plants.

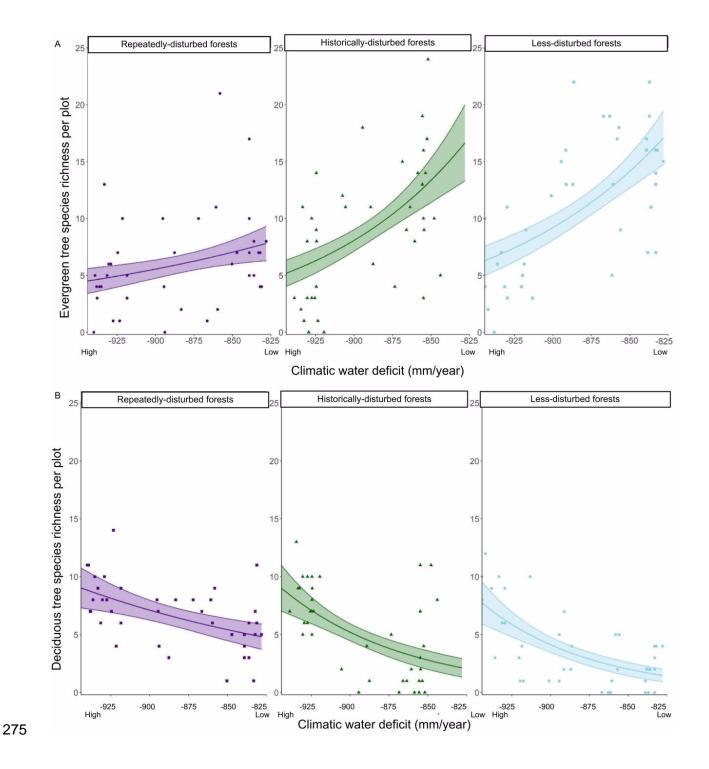


Figure 4. Plots showing the relationship between climatic water deficit and number of evergreen
tree species per plot (A) and number of deciduous tree species per plot (B) across the three
disturbance categories. The solid line with a band indicates the best-fitted line for the

279 Generalized Linear Model (Poisson error structure) with associated 95% CI. Higher values of
280 CWD (less negative) imply lower water stress experienced by plants.

281

282 DISCUSSION

283 While previous studies have mostly examined the effects of CAD and climate on vegetation in 284 isolation, we examined the interactive effects of the two drivers on diversity, structure and 285 composition of trees in a biodiversity hotspot. Our results demonstrate that CAD and CWD 286 significantly influence woody plant composition, overall tree species richness and structure. We 287 found that CAD disrupts the relationship between CWD and vegetation type, resulting in an 288 increased representation of deciduous trees in chronically disturbed forests, an aspect rarely 289 documented in previous studies. This study provides a scientific basis for planting evergreen 290 trees in ecological restoration efforts in low-elevation, chronically disturbed forests, which 291 predominantly harbour deciduous trees. Given the absence of Protected Areas in low elevations 292 and significant differences in vegetation composition in high- and low-elevation forests, the 293 existing Reserved (historically-disturbed forests) and Community-owned sacred groves (less-294 disturbed forests) must be protected from further degradation and conversion in partnership with 295 local stakeholders.

296 CAD-induced shift from evergreen to deciduous forests

Most studies have looked at the impacts of CAD and climate in isolation. Given our poor understanding of the interactive effects between CAD and climate, this study helps fill that knowledge gap. Higher water deficit induces deciduousness in the vegetation and contributes to the transition from evergreen to deciduous vegetation type (Saiter et al. 2015). While CAD is thought to impact plant diversity and structure (Rito et al. 2017), very few studies have

302 documented CAD causing shifts in vegetation types (Bradshaw & Hannon, 1992) as documented 303 in this study. We demonstrate that CAD disrupts this relationship between CWD and 304 evergreenness. We find that for given levels of CWD, CAD results in a significantly lower 305 representation of every even trees and species in the community. The greater representation of 306 deciduous trees with CAD could be because of several reasons: source, dispersal and 307 establishment limitation, poor competitive ability of evergreen tree species in more open 308 conditions and greater propensity of deciduous trees to coppice (Chandran 1997). The species 309 composition of neighbouring forests shapes plant composition in a forest patch through seed 310 dispersal (Butaye et al., 2002). In the low-elevation areas of our study site, there are very few 311 remaining patches of intact, evergreen forests. Additionally, our previous study shows that CAD 312 negatively impacts the prevalence of avian frugivores (Biswas et al., 2023), which play an 313 important role in the seed dispersal of fleshy-fruited evergreen plants (Naniwadekar et al., 2019). 314 Thus, the absence of evergreen tree species and reduced habitat use by frugivores in repeatedly-315 disturbed forests could result in source limitation. Even if the seeds are dispersed in these open 316 degraded patches, evergreen seedlings, which tend to have a higher shade tolerance (Baldocchi et 317 al., 2010; Kitajima et al., 2013), may not be able to establish in low canopy cover conditions 318 (Swinfield et al., 2016). Human disturbances such as clear felling of forests initiate secondary 319 succession, which often begins with a resource-rich condition associated with increased light 320 availability due to reduced canopy cover (Dalling, 2008). Often, the pioneer or early successional 321 species in degraded forests are shade-intolerant deciduous trees (Jin et al., 2017). Deciduous tree 322 species usually adopt a resource-acquisitive strategy and have higher growth rates, unlike 323 evergreen tree species that adopt a resource-conservative strategy (L. Wang et al., 2023). Thus, 324 deciduous tree species can be expected to be competitively superior in more open conditions,

resulting in the filtering of evergreen tree species recruits in degraded habitats. Furthermore, deciduous trees are known to coppice well (Chandran 1997) and are associated with degraded habitats, as seen in our study. Future studies need to determine the relative influence of different processes in causing the shift from evergreeness to deciduousness. Our findings offer evidence that there is a need for active restoration efforts with focus on evergreen plants in degraded habitats of the northern Western Ghats since natural regeneration may not be able to restore evergreen plants, which were likely present in the past.

332 The distribution of humid tropical forests (HTFs) is best characterised by high rainfall 333 regimes with low water stress environments (Zelazowski et al., 2011). Climate change 334 projections predict that rainfall will decrease in parts of the Western Ghats (Katzenberger et al., 335 2021; Rajendran et al., 2012). Thus, chronic anthropogenic disturbance and lowered precipitation 336 can trigger shifts in vegetation type to drier vegetation in these humid tropical forests. Moreover, 337 a greater proportion of evergreen tree species are under the threatened category than deciduous 338 tree species (Fig. S4). Thus, the transition from evergreen to deciduous forests also has 339 consequences for threatened plant species in the region.

340

341 Sacred groves as reservoirs of biodiversity

We found that the community-owned, less-disturbed forests, i.e., sacred groves, had comparable richness and structure as the state-owned, historically-disturbed forests. This contrasts with the central Western Ghats, where state-owned forests performed better on diversity and structure metrics than community-owned sacred groves (Osuri et al., 2014). The drivers of higher plant diversity in sacred groves could be multifold, as they could be remnants of historically contiguous forests or could have been actively restored in the historical past. The existing

348 evergreen tree cover in two sacred groves in central Western Ghats was attributed to cultural 349 practice-driven forest recovery from around 1000 years before the present (Bhagwat et al., 2013). 350 Therefore, there is a need to understand the socio-ecological history of these sacred groves that 351 can throw more light on the observed diversity of woody plants in these groves. 352 Sixty percent of sacred groves in the central Western Ghats originally present in official 353 records were lost and there was a decrease in above-ground biomass and proportion of evergreen 354 species in existing ones (Osuri et al., 2014; Bhagwat et al., 2005). Similarly, there is documented 355 evidence of sacred groves in the northern Western Ghats being cleared and lost for almost 50 356 years (Gadgil and Vartak, 1976). The drivers of loss are many, including logging, 357 encroachments, and habitat conversion. With monoculture plantations replacing private forests 358 and increased development, the pressures on the sacred groves as a source of timber and 359 fuelwood for local communities will increase. Partnerships between conservation practitioners, 360 government departments and local communities are critical to safeguard these groves in a 361 socially just manner. 362

363 Value of low-elevation forests

We recorded a higher taxonomic diversity of woody plants in our low-elevation sampling sites than the high-elevation ones, especially for the repeatedly-disturbed and less-disturbed forests (Figure S3). Previous studies documenting plant diversity across elevational gradients have also reported similar patterns, with species richness decreasing with elevation (Musciano et al., 2021; Malizia et al., 2020). In the Western Ghats, forests at or below 500 m have the least representation in the protected area network (Bawa et al., 2007). Moreover, only around 1% of land in the northern Western Ghats is legally protected (Blicharska et al., 2013). All the PAs in 371 our study region are in high elevations, leaving the highly diverse, lowland forests unprotected. 372 PAs, which are the high-elevation historically disturbed sites, had the least abundance of cut 373 stems, indicating their efficacy in reducing extractive pressures. The privately or community-374 owned forests here are prone to habitat degradation (Kulkarni and Mehta, 2013) and conversion, 375 mostly to cash crop plantations (Kale et al., 2016). More than one-third of the geographic area of 376 two tehsils in Sindhudurg District is under cashew plantation (area: 533.5 km²) (Rege et al., 377 2022). This study did not estimate areas under rubber and mango plantations, which are also 378 prevalent in the region. Most of these plantations were erstwhile private or community-owned 379 forests. Therefore, these remnant patches of RFs and sacred groves in the low elevations are 380 important reservoirs of diverse assemblages of woody plants and other biodiversity. The RFs are 381 relatively less protected than PAs and are more vulnerable to getting denotified and converted to 382 other land uses (Patil 2023). Protection and active restoration, wherever needed, must be 383 prioritised in these forests so that they can continue to sustain high levels of biodiversity. 384 The higher number of cut stems and low basal area of trees in the repeatedly disturbed 385 private forests (Table S4) of high elevations suggests that these forests experience greater CAD 386 than those in the low elevations. As CWD is strongly correlated with elevation, forests with low CWD are high-elevation forests. Generally, an opposite trend is observed globally, where low-387 388 elevation forests are more vulnerable to logging, deforestation and habitat conversion, even in 389 the world's biodiversity hotspots (Hamunyela et al., 2020; Tapia-Armijos et al., 2015). This may 390 be attributed to the easy access to large trees by logging companies in the foothills compared to 391 the stunted but steep forests of the highlands (Danielsen et al., 2010). However, in our study 392 area, the high-elevation forests, which receive a high amount of rainfall (> 5000 mm), are 393 relatively easily accessible as they are on the Deccan plateau. These high-elevation, privately-

394 owned forests are also a source of fuelwood for the locals, who may rely on it to provide warmth 395 in their homes during the cooler and wetter seasons. A study across an altitudinal gradient in the 396 western Himalayas revealed a similar occurrence. Fuelwood consumption was found to be 2.6 397 times higher at high elevations (above 2000 m) than its use at low elevations (up to 500 m), 398 owing to people's need to heat spaces and water in harsh weather (Bhatt and Sachan, 2004). The 399 fuelwood is also utilised by the sugar factories in the region, most of which are in the high-400 elevations. Fuelwood collection can lead to forest degradation (Sassen et al., 2015) and 401 negatively affect regeneration, thereby changing the vegetation type (as documented in this 402 study) and leading to biodiversity decline. Evergreen forests in the Western Ghats harbour a higher diversity of endemic and threatened trees than the deciduous forests (Chandran 1997; 403 404 Ghate et al. 1998). Thus, the relative needs of industry and local communities for fuelwood must 405 be determined to find suitable alternatives to fuelwood and restore the degraded private forests in 406 these regions.

407 **Conservation implications**

408 Tropical wet forests are vulnerable to climate fluctuations and anthropogenic disturbance. We 409 can expect CAD to shift the wetter forests to drier vegetation types. This effect will be 410 exacerbated by reduced water availability due to climate change. Thus, there is a need to find 411 suitable alternatives to reduce CAD in tropical forests since conversion to deciduous forests will 412 be associated with a significant loss of threatened and endemic plant diversity, which is generally higher in wet tropical forests. Due to the degraded nature of privately owned forests in the 413 414 northern Western Ghats, primarily driven by periodic clear felling of forests that cater to the 415 fuelwood needs of nearby factories, there is a need to find suitable alternatives to fuelwood in 416 these factories. Given the agroforestry-friendly government policies and the increased advent of

417 technology that allows easy conversion of forests to agroforestry plantations, many existing 418 privately-owned forests will likely be converted to agroforestry plantations. Given compositional 419 and diversity differences in low- and high-elevation forests, ecological restoration efforts of 420 degraded, low-elevation private forests must be prioritised in partnership with local landowners. 421 This is especially important for the lower elevations, which do not harbour any Protected Areas 422 but continue to harbour significant threatened and endemic biodiversity. The existence of 423 restored habitat patches in a predominantly human-modified landscape will enable the threatened 424 biodiversity to persist in the long term. Given that the existing degraded forest patches harbour 425 predominantly deciduous forests due to CAD, ecological restoration efforts should plant 426 evergreen tree species, particularly in the lower-elevation forests.

427

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712 SUPPLEMENTARY MATERIAL

- **Table S1.** Checklist of woody plant species detected in $120 (50 \times 10 \text{ m}^2)$ plots. The table also
- 714 provides information on the number of plots in which the species was detected and the total
- 715 number of individuals detected.

716

Sr. No	Species	Family	Number of individuals	Number of plots where the species was found
1	Acacia auriculiformis	Fabaceae	3	2
2	Actephila excelsa	Phyllanthaceae	1	1
3	Actinodaphne lanceolata	Lauraceae	21	10
4	Adina cordifolia	Rubiaceae	3	1
5	Aglaia elaeagnoidea	Meliaceae	46	12
6	Aglaia lawii	Meliaceae	51	13
7	Agrostistachys indica	Euphorbiaceae	4	1
8	Albizia lebbeck	Fabaceae	20	7

		1		
9	Albizia procera	Fabaceae	1	1
10	Alstonia scholaris	Apocynaceae	3	3
11	Anacardium occidentale	Anacardiaceae	8	3
12	Antiaris toxicaria	Moraceae	11	6
13	Aphanamixis polystachya	Meliaceae	1	1
14	Aporosa cardiosperma	Phyllanthaceae	113	19
15	Ardisia solanacea	Primulaceae	163	7
16	Artocarpus gomezianus	Moraceae	1	1
17	Artocarpus heterophyllus	Moraceae	12	6
18	Atalantia racemosa	Rutaceae	43	11
19	Bauhinia racemosa	Fabaceae	5	3
20	Beilschmiedia dalzellii	Lauraceae	42	13
21	Bergera koenigii	Rutaceae	12	8

			I	
22	Bischofia javanica	Phyllanthaceae	21	1
23	Blachia denudata	Euphorbiaceae	21	3
24	Bombax ceiba	Malvaceae	19	15
25	Bridelia retusa	Phyllanthaceae	74	32
26	Buchanania lanzan	Anacardiaceae	21	6
27	Butea monosperma	Fabaceae	2	2
28	Callicarpa tomentosa	Lamiaceae	22	12
29	Canarium strictum	Burseraceae	1	1
30	Capparis rotundifolia	Capparaceae	2	1
31	Carallia brachiata	Rhizophoraceae	12	5
32	Careya arborea	Lecythidaceae	114	41
33	Caryota urens	Arecaceae	48	23
34	Casearia graveolens	Salicaceae	52	25

35	Cassia fistula	Fabaceae	2	2
36	Catunaregam spinosa	Rubiaceae	149	30
37	Celtis philippensis	Cannabaceae	7	5
38	Celtis timorensis	Cannabaceae	1	1
39	Chionanthus mala-elengi	Oleaceae	12	4
40	Chukrasia tabularis	Meliaceae	1	1
41	Cinnamomum verum	Lauraceae	14	7
42	Clausena anisata	Rutaceae	1	1
43	Cleidion spiciflorum	Euphorbiaceae	18	2
44	Clerodendrum infortunatum	Lamiaceae	2	2
45	Croton zeylanicus	Euphorbiaceae	1	1
46	Cryptocarya wightiana	Lauraceae	9	3
47	Dalbergia sissoo	Fabaceae	16	5

48	Delonix regia	Fabaceae	1	1
49	Dichapetalum gelonioides	Dichapetalaceae	35	9
50	Dillenia pentagyna	Dilleniaceae	13	9
51	Dimocarpus longan	Sapindaceae	193	23
52	Diospyros candolleana	Ebenaceae	147	27
53	Diospyros montana	Ebenaceae	13	10
54	Diospyros nigrescens	Ebenaceae	123	18
55	Diospyros oocarpa	Ebenaceae	70	10
56	Diospyros sylvatica	Ebenaceae	9	5
57	Donella lanceolata	Sapotaceae	3	1
58	Drypetes venusta	Putranjivaceae	23	6
59	Dysoxylum gotadhora	Meliaceae	46	9
60	Ehretia aspera	Boraginaceae	6	5

	1			1
61	Elaeocarpus variabilis	Elaeocarpaceae	5	2
62	Elaeodendron paniculatum	Celastraceae	5	4
63	Erinocarpus nimmonii	Malvaceae	1	1
64	Erythrina stricta	Fabaceae	6	4
65	Eugenia kalamii	Myrtaceae	5	2
66	Euonymus indicus	Celastraceae	7	1
67	Falconeria insignis	Euphorbiaceae	12	6
68	Ficus amplissima	Moraceae	1	1
69	Ficus arnottiana	Moraceae	1	1
70	Ficus callosa	Moraceae	1	1
71	Ficus exasperata	Moraceae	5	3
72	Ficus hispida	Moraceae	41	12
73	Ficus microcarpa	Moraceae	1	1

			-	
74	Ficus nervosa	Moraceae	7	7
75	Ficus racemosa	Moraceae	32	16
76	Ficus talbotii	Moraceae	1	1
77	Ficus tsjakela	Moraceae	2	2
78	Ficus virens	Moraceae	1	1
79	Firmiana colorata	Malvaceae	1	1
80	Flacourtia montana	Salicaceae	26	13
81	Garcinia gummi-gutta	Clusiaceae	8	4
82	Garcinia indica	Clusiaceae	39	8
83	Garcinia talbotii	Clusiaceae	64	16
84	Glochidion heyneanum	Phyllanthaceae	16	6
85	Glochidion hohenackeri	Phyllanthaceae	9	5
86	Glochidion zeylanicum	Phyllanthaceae	3	3

87	Glycosmis pentaphylla	Rutaceae	15	9
88	Gmelina arborea	Lamiaceae	15	9
89	Gomphandra tetrandra	Stemonuraceae	4	3
90	Grewia serrulata	Malvaceae	14	5
91	Grewia tiliifolia	Malvaceae	66	22
92	Gymnosporia rothiana	Celastraceae	1	1
93	Helicteres isora	Malvaceae	13	8
94	Heterophragma quadriloculare	Bignoniaceae	4	4
95	Holigarna arnottiana	Anacardiaceae	130	20
96	Holigarna grahamii	Anacardiaceae	39	20
97	Homalium ceylanicum	Salicaceae	6	3
98	Hydnocarpus pentandrus	Achariaceae	17	7
99	Ixora brachiata	Rubiaceae	376	45

100	Irora pavatta	Rubiaceae	7	3
100	Ixora pavetta	Kublaceae	/	5
101	Jatropha curcas	Euphorbiaceae	1	1
102	Justicia adhatoda	Acanthaceae	9	1
103	Knema attenuata	Myristicaceae	24	4
104	Kydia calycina	Malvaceae	1	1
105	Lagerstroemia microcarpa	Lythraceae	61	27
106	Lagerstroemia parviflora	Lythraceae	1	1
107	Lannea coromandelica	Anacardiaceae	12	7
108	Lasiosiphon glaucus	Thymelaeaceae	32	12
109	Leea indica	Vitaceae	111	34
110	Lepisanthes tetraphylla	Sapindaceae	23	11
111	Ligustrum robustum	Oleaceae	23	10
112	Litsea glutinosa	Lauraceae	2	1

113	Litsea nigrescens	Lauraceae	3	1
114	Litsea stocksii	Lauraceae	23	7
115	Lophopetalum wightianum	Celastraceae	18	2
116	Macaranga peltata	Euphorbiaceae	125	29
117	Machilus glaucescens	Lauraceae	11	5
118	Mallotus nudiflorus	Euphorbiaceae	1	1
119	Mallotus philippensis	Euphorbiaceae	86	27
120	Mallotus resinosus	Euphorbiaceae	9	2
121	Mammea suriga	Calophyllaceae	49	3
122	Mangifera indica	Anacardiaceae	49	20
123	Mappia nimmoniana	Icacinaceae	25	9
124	Margaritaria indica	Phyllanthaceae	2	2
125	Melastoma malabathricum	Melastomataceae	3	1

126	Melia dubia	Meliaceae	1	1
127	Melicope lunu-ankenda	Rutaceae	1	1
128	Memecylon talbotianum	Melastomataceae	8	4
129	Memecylon umbellatum	Melastomataceae	778	56
130	Memecylon wightii	Melastomataceae	1	1
131	Meyna laxiflora	Rubiaceae	32	15
132	Microcos paniculata	Malvaceae	64	19
133	Miliusa tomentosa	Annonaceae	14	6
134	Mimusops elengi	Sapotaceae	25	14
135	Mitragyna parvifolia	Rubiaceae	3	2
136	Monoon fragrans	Annonaceae	2	2
137	Morinda coreia	Rubiaceae	3	3
138	Murraya paniculata	Rutaceae	3	3

139	Myristica beddomei	Myristicaceae	21	4
140	Myristica malabarica	Myristicaceae	12	2
141	Neolamarckia cadamba	Rubiaceae	3	2
142	Neolitsea zeylanica	Lauraceae	5	2
143	Nothopegia castaneifolia	Anacardiaceae	75	31
144	Ochna obtusata	Ochnaceae	4	2
145	Pavetta sp	Rubiaceae	1	1
146	Phyllanthus emblica	Phyllanthaceae	17	7
147	Pongamia pinnata	Fabaceae	3	1
148	Prunus ceylanica	Rosaceae	1	1
149	Psydrax dicoccos	Rubiaceae	16	11
150	Pterocarpus marsupium	Fabaceae	6	4
151	Pterospermum diversifolium	Malvaceae	5	3

152	Putranjiva roxburghii	Putranjivaceae	3	3
153	Sageraea laurina	Annonaceae	48	12
154	Sapindus trifoliatus	Sapindaceae	1	1
155	Saraca asoca	Fabaceae	86	6
156	Schleichera oleosa	Sapindaceae	39	9
157	Securinega sp	Phyllanthaceae	1	1
158	Semecarpus anacardium	Anacardiaceae	4	1
159	Senegalia catechu	Fabaceae	29	6
160	Solenocarpus indicus	Anacardiaceae	2	2
161	Spondias pinnata	Anacardiaceae	2	2
162	Sterculia foetida	Malvaceae	1	1
163	Sterculia guttata	Malvaceae	9	8
164	Sterculia urens	Malvaceae	1	1

165	Stereospermum colais	Bignoniaceae	38	22
166	Strombosia ceylanica	Olacaceae	12	3
167	Strychnos nux-vomica	Loganiaceae	8	7
168	Symplocos macrophylla	Symplocaceae	89	16
169	Syzygium caryophyllatum	Myrtaceae	18	9
170	Syzygium cumini	Myrtaceae	249	64
171	Syzygium gardneri	Myrtaceae	1	1
172	Syzygium hemisphericum	Myrtaceae	48	12
173	Syzygium rubicundum	Myrtaceae	9	4
174	Tabernaemontana alternifolia	Apocynaceae	116	40
175	Tamilnadia uliginosa	Rubiaceae	3	3
176	Tectona grandis	Lamiaceae	118	13
177	Terminalia bellirica	Combretaceae	60	28

178	Terminalia chebula	Combretaceae	25	11
179	Terminalia elliptica	Combretaceae	298	41
180	Terminalia paniculata	Combretaceae	437	49
181	Tetrameles nudiflora	Tetramelaceae	1	1
182	Tetrapilus dioicus	Oleaceae	125	31
183	Tritaxis glabella	Euphorbiaceae	45	5
184	Vitex altissima	Lamiaceae	7	2
185	Wendlandia thyrsoidea	Rubiaceae	2	2
186	Wrightia arborea	Apocynaceae	22	6
187	Wrightia tinctoria	Apocynaceae	23	9
188	Xantolis tomentosa	Sapotaceae	151	37
189	Xylia xylocarpa	Fabaceae	120	16
190	Zanthoxylum rhetsa	Rutaceae	45	15

191	Ziziphus rugosa	Rhamnaceae	1	1
192	Ziziphus xylopyrus	Rhamnaceae	1	1

717

718 Table S2. Treatment contrast table showing coefficient estimates and associated 95% CI for the

719 Generalized linear model with Poisson error structure that examined relationship between

720 observed species richness per plot (not including lianas) (≥ 10 cm GBH) and climatic water

721 deficit and land-use categories. Pseudo $R^2 = 0.09$.

Predictor variable	Estimate (95% CI)	p
Intercept: Less-disturbed forests	5.17 (3.41-6.94)	< 0.001
Historically-disturbed forests	-0.73 (-3.54 – 2.07)	0.61
Repeatedly-disturbed forests	-3.12 (-5.680.58)	0.016
Climatic water deficit (CWD)	0.002 (0.001 – 0.004)	0.006
CWD × Historically-disturbed forests	-0.001 (-0.004 - 0.002)	0.62
CWD × Repeatedly-disturbed forests	-0.003 (-0.0060.0005)	0.02

- 723 **Table S3.** Treatment contrast table showing coefficient estimates and associated 95% CI for the
- 724 General linear model with Gaussian error structure that examined the relationship between the
- square root of basal area (m²ha⁻¹) and CWD and land-use categories. $R^2 = 0.51$.

Predictor variable	Estimate (95% CI)	р
Intercept: Less-disturbed forests	13.23 (5.84 – 26.13)	< 0.001
Historically-disturbed forests	2.76 (-1.8×10 ⁻⁴ – 0.02)	0.68
Repeatedly-disturbed forests	-29.17 (-44.919.00)	< 0.001
Climatic water deficit (CWD)	0.01 (-15.9 – 10.36)	0.1
$CWD \times Historically-disturbed$ forests	0.003 (-0.0480.019)	0.66
$CWD \times Repeatedly-disturbed$ forests	-0.03 (-0.018 – 0.01)	< 0.001

- **Table S4.** Table summarises information (Mean \pm SE) on tree density, basal area, canopy cover
- and number of cut stems per plot across different land-use categories in low and high elevations.

Structural	Repeatedly	Repeatedl	Historica	Historically	Less-	Less-
variables	-disturbed	у-	lly-	-disturbed	disturbed	disturbed
	forests	disturbed	disturbed	forests	forests (Low)	forests
	(Low)	forests	forests	(High)		(High)
		(High)	(Low)			

Tree density	1081	889	1137	1565	1041	1288
(ha ⁻¹)	(±90.7)	(±99.2)	(±81.9)	(±185.8)	(±106.9)	(±112.4)
Basal area (m ²	24.3 (±	11.2 (±	32.3 (±	42.6 (± 2.7)	36.2 (± 2.6)	46.4 (± 5.5)
ha ⁻¹)	2.9)	2.3)	3.3)			
Canopy cover	39.6 (±2.2)	14.1	48.6	45.4 (±1.3)	46.5 (±2.6)	50 (±1.5)
(%)		(±2.9)	(±2.4)			
Number of cut	15.8 (±	24.7 (±	6.2 (±	2.2 (± 1.3)	5.0 (± 1.1)	10 (± 2.4)
stems per plot	3.3)	3.3)	1.3)			

729

Table S5. Treatment contrast table showing coefficient estimates and associated 95% CI for the Generalized linear model with binomial error structure that examined relationship between proportion of evergreen individuals (≥ 10 cm GBH) and climatic water deficit and land-use categories. Pseudo $R^2 = 0.38$.

Predictor variable	Estimate (95% CI)	р
Intercept: Less-disturbed forests	28.44 (25.28 - 31.74)	< 0.001
Historically-disturbed forests	-7.88 (-11.92 – 3.92)	< 0.001
Repeatedly-disturbed forests	-12.52 (-16.44 – -8.69)	< 0.001
Climatic water deficit (CWD)	0.03 (0.026 - 0.034)	< 0.001

CWD × Historically-disturbed forests	-0.008 (-0.0120.003)	< 0.001
CWD × Repeatedly-disturbed forests	-0.012 (-0.0160.008)	< 0.001

734

- **Table S6.** Treatment contrast table showing coefficient estimates and associated 95% CI for the
- 736 Generalized linear model with Poisson error structure that examined the relationship between the
- number of evergreen tree species (not including lianas) (≥ 10 cm GBH) and climatic water
- 738 deficit and land-use categories. Pseudo $R^2 = 0.33$.

Predictor variable	Estimate (95% CI)	р
Intercept: Less-disturbed forests	9.96 (7.85 – 12.11)	< 0.001
Historically-disturbed forests	1.16 (-2.36 – 4.69)	0.52
Repeatedly-disturbed forests	-3.99 (-7.41 – -0.56)	0.02
Climatic water deficit (CWD)	0.009 (0.006 – 0.01)	< 0.001
$CWD \times Historically-disturbed$ forests	0.0014 (-0.002 - 0.005)	0.48
$CWD \times Repeatedly-disturbed$ forests	-0.004 (-0.008 – 0.00005)	0.052

739

Table S7. Treatment contrast table showing coefficient estimates and associated 95% CI for the
Generalized linear model with Poisson error structure that examined the relationship between the

- number of deciduous tree species (not including lianas) (≥ 10 cm GBH) and climatic water
- 743 deficit and land-use categories. Pseudo $R^2 = 0.36$.

Predictor variable	Estimate (95% CI)	p
Intercept: Less-disturbed forests	-11.4 (-15.4 – -7.5)	< 0.001
Historically-disturbed forests	1.89 (-3.8 – 7.5)	0.51
Repeatedly-disturbed forests	8.5 (3.8 – 13.3)	< 0.001
Climatic water deficit (CWD)	-0.014 (-0.020.01)	< 0.001
$CWD \times Historically-disturbed$ forests	0.002 (-0.004 - 0.008)	0.56
$CWD \times Repeatedly-disturbed$ forests	0.009 (0.004 – 0.014)	< 0.001

744

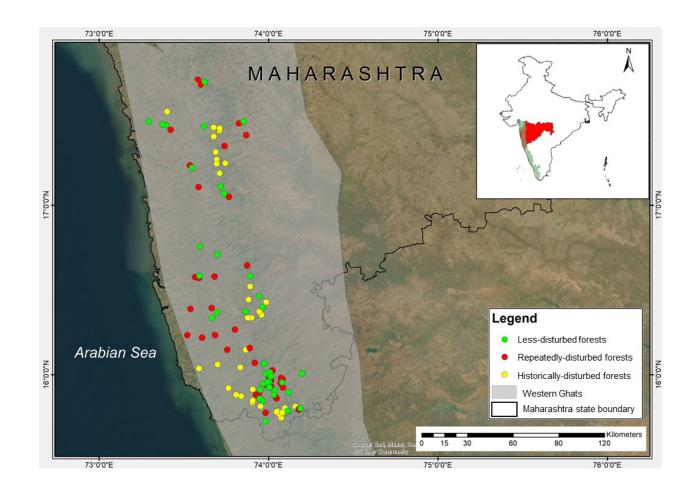
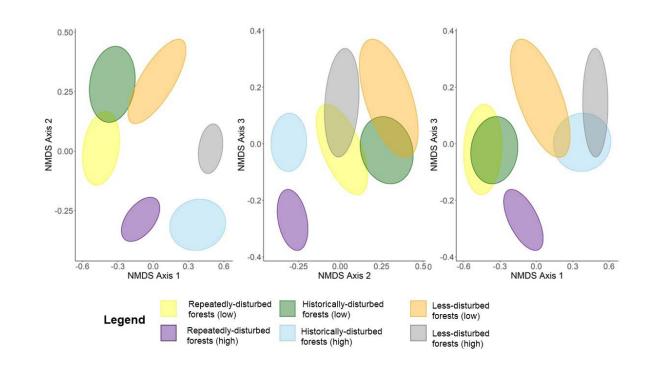
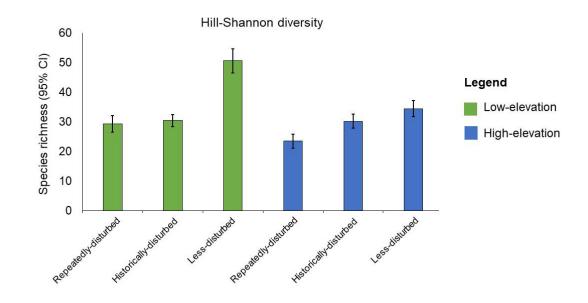


Figure S1. Map showing the vegetation plots distributed in our study region in Maharashtra state
in India. The three forest categories are spread across low (8–514 m ASL) and high elevations
(577–1054 m ASL).



751 Figure S2. Plot of Non-metric Multidimensional Scaling showing differences in plant

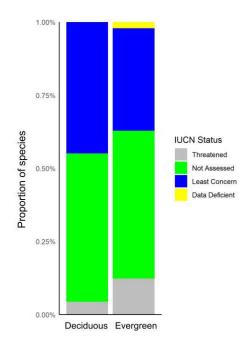
composition between low and high elevations and between the different land-use categories.

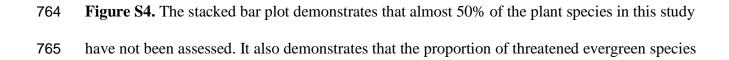


753

754	Figure S3. Hill-Shannon diversity (95% CI) of the three different land-use types, across low and
755	high elevations clearly showing significantly higher diversity of woody plants in less-disturbed
756	low elevation forests. We used the sample-coverage-based method to estimate the species
757	richness of plant communities in the different forest categories (low and high, less-, historically-
758	and repeatedly-disturbed forests) (Roswell et al. 2021). The least sample coverage was in the
759	low-elevation sacred groves (97%). Therefore, we rarefied the diversity measure of all other
760	categories to 97% sample coverage. We bootstrapped the data 50 times to estimate 95%
761	confidence intervals.

762





- is almost twice that of the deciduous species. This study will contribute to generating important
- information about the not-assessed, Data Deficient and Threatened species for the region.