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1 **Diversity, functionality and resilience under increasing harvesting intensities in woodlands of**
2 **Northern Patagonia**

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12
13 **Abstract**

14 Sustainable forest management relies on the understanding of biodiversity response to disturbance
15 and the ecological resilience of the system. The dynamic equilibrium hypothesis (DEM) predicts
16 that site productivity will modulate the effects of disturbance gradient on biodiversity. Also,
17 considering functional diversity (eco-morfo-physiological traits related to resource usage) is needed
18 to understand the effect of species gains and losses on ecosystem functionality. Here we assess the
19 response of understory plant taxonomic and functional diversity to increasing harvesting intensities
20 (0, 30, 50 and 70% of basal area removed) at three woodland sites of contrasting biomass growth
21 (productivity) in northern Patagonia. Also, we assessed resilience based on comparisons with
22 undisturbed treatments four years after initial harvest. In agreement with DEM, both taxonomic and
23 functional diversity peaked at high, medium, or low harvesting intensities in the high-, medium-, or
24 low-productivity site, respectively. Taxonomic composition was clearly determined by site identity,
25 while no pattern emerged for functional composition. Functional traits related to light use showed
26 different responses: specific leaf area was only affected by site identity while leaf chlorophyll
27 content was affected by an interaction between harvesting intensity and site identity. Interestingly,

28 there was no effect of harvesting intensity on the resilience of taxonomic diversity and functional
29 composition. Only for functional diversity, harvesting intensity was as important as site identity. In
30 the high and intermediate productivity sites the traits that characterizes the system were more
31 resilient and resembled the control treatment after four years of low or high (but not intermediate)
32 harvesting intensities. Our results support the use of the DEM on forest interventions and the
33 importance of considering both taxonomic and functional composition, as the consideration of
34 functional traits related to resource use strategies have different implications when considering the
35 resilience of the system.

36

37 **Keywords**

38 dynamic equilibrium model, functional diversity, functional trait, recovery, Patagonia, forest
39 management, temperate woodlands, thinning

40

41

42 **1. Introduction**

43 Understanding the effect of forest management on biodiversity is important for designing
44 sustainable silvicultural practices. Forest management (e.g. thinning, pruning, harvesting, etc) has
45 been traditionally orientated mainly to products (e.g. timber, biomass), while consequences on
46 biodiversity has been less studied (Puettmann et al. 2015). Nevertheless, nowadays there is
47 consensus that sustainable forestry must maintain and even favor biodiversity for achieving
48 sustainability goals (The Montreal Process 2015). Since in many countries most of forest are on
49 private lands (Mayer and Tikka 2006), linking biodiversity and forest management becomes a prior
50 objective.

51 Forest harvesting is a common silvicultural practice, which reduces tree density to improve
52 tree growth, affecting biodiversity and community structure according to stand complexity and
53 management strategies (Verschuly et al. 2011; Duguid and Ashton 2013; Root and Betts 2016). In
54 temperate forests, where light availability is a key driver of community composition, canopy
55 opening can increase species diversity, as it promotes a variety of new habitat types and resources
56 (Lencinas et al. 2011; Chillo et al. 2018; Nacif et al. 2020). Yet disturbance effects on biodiversity
57 are not independent of site productivity (Thomas et al. 1999; Cingolani et al. 2005), although few
58 studies consider both factors simultaneously. The Dynamic Equilibrium Model (DEM, Huston
59 2014) propose that growth rates (productivity) and mortality associated to the disturbance are key
60 processes interacting in their effects on species diversity. This model predicts that site productivity
61 will modulate the effects of the disturbance gradient on diversity (Huston, 2014). Accordingly, plant
62 diversity might peak at high disturbance intensities in high productivity forests while peaking at
63 medium intensities in lower productivity forests (Cornell, 1978).

64 The understanding of the effects of forest management on ecosystem dynamics needs to
65 consider more than just species richness and abundance. Diversity influences ecosystem functioning
66 through the type, range and relative abundance of functional traits (Cadotte et al. 2011). Thus, the
67 analysis of functional diversity allows us to understand the effect of species gains and losses on the
68 functioning of the ecosystem as a whole. For example, functional traits related to the leaf economic

69 spectrum (LES) such as specific leaf area and photosynthetic capacity gives information about
70 resource use strategies of understory plant community (Wright et al. 2004). These traits have a
71 strong correlation that suggest a constrained set of options regarding strategies in leaf production.
72 This is, an economic spectrum ranging from leaves with a quick return on dry mass and nutrient
73 investment (i.e. high specific leaf area and photosynthetic rate, with low life span) to leaves with a
74 slow potential rate of return (i.e. low specific leaf area and photosynthetic rate, but long life span)
75 (Shipley et al. 2006). When considering harvesting intensity as a disturbance, this information can
76 be related to the composition of the community that grew after the disturbance in terms of primary
77 productivity and nutrient cycling, as well as the ecosystem services that depend upon those
78 processes (Lavorel and Grigulis 2012).

79 Other important issue for forest management is the understanding of the ecological
80 resilience of the ecosystem, known as its capacity to return to a pre-disturbed condition and still
81 maintain its essential structure and function (Holling 1973). In general, more productive sites are
82 expected to be more resilient than less productive ones (Stone et al. 1996; Thompson et al. 2009).
83 Nevertheless, few attempts had been done for understanding the consequences of disturbance
84 intensity and site productivity on forest ecosystem resilience (Clarke et al. 2005; Kohv et al. 2013).
85 To add complexity to this issue, both taxonomic and functional identity of the resulting community
86 are important, because species richness may be similar but if the composition differs, the
87 community may change in its functionality (Lipoma et al. 2017). For example, traits such as life
88 form and dispersal mechanisms represent an adaptive response to disturbances and had been
89 proposed to be central for ecosystems resilience (Johnstone et al. 2016). Therefore, in order to
90 design environmentally friendly forest management practices, it is necessary to understand how
91 disturbance and productivity interact for shaping different components of biodiversity (taxonomic
92 and functional) and ecosystem resilience.

93 Here we analyze the response of taxonomic and functional biodiversity and composition,
94 and resilience to harvesting intensity in Northern Patagonian mixed secondary woodlands, and its
95 recovery after 4 years. We used an experimental approach of systematic harvesting treatments in

96 strips of increasing width and constant length and replicated on three sites with different
97 productivity (growing rates of dominant tree species) (Coulin et al. 2019). Our working hypothesis
98 were: 1) taxonomic and functional diversity will vary with harvesting intensity, and the pattern of
99 the response will depend on site productivity. We predict that sites with greater productivity will
100 tolerate more harvesting intensity, showing higher values of diversity than sites with less
101 productivity sites; 2) harvesting releases dominant competitor pressure (for light resource), thus we
102 expect changes in community composition regarding different resource use strategies (mainly light
103 use availability) along harvesting intensity gradient; and 3) the resilience of the ecosystem to
104 harvesting intensity will mainly depend on site productivity. We expect higher resilience at greater
105 productivity sites than at less productivity sites.

106

107 **2. Methods**

108 The north Patagonian Andean region is a Mediterranean type-climate region, with
109 annual precipitation ranging from 920 mm to 1,300 mm in the *Nothofagus antactica* distribution,
110 with average annual maximum temperature of 15 °C and minimum temperature of 1.5 °C. Frosts
111 occur about 120 days a year, with 0.5 days' hail, annual relative humidity 65 %, and an annual dew
112 temperature of 2 °C (Reque et al. 2007). Presence of frost is longer in valley bottom, being the less
113 favorable sites for tree growth due to cold air accumulation (Davel and Ortega 2003).

114 Three sites with different environmental conditions in the province of Rio Negro, Argentina,
115 were chosen to conduct the study (Figure S1, Table S1). The categorization of site productivity was
116 based on several indicators of site condition (Table S1), but the main index used to determine site
117 productivity was Mean Annual Increment (MAI; m³ ha⁻¹ year⁻¹), which describes increments in
118 firewood biomass. This index reflects biomass volume by stand age and it is a direct measure of site
119 quality regarding biomass production. Thus, high, intermediate and low productivity sites refer to
120 sites with greater, intermediate and less firewood biomass growth. Several other variables were
121 considered for the categorization, such as the mean height of dominant trees and on-site exposure,
122 one of the main environmental factors driving forests physiognomy in this region. In southern-

123 exposure hillsides, soils are deeper, have greater development and higher moisture retention than
124 northern hillsides, where soils are drier because are exposed to the dominant northwestern winds
125 and intense summer droughts (Davel and Ortega, 2003). The valley bottom site present specific
126 environmental conditions that are limiting for biomass growth, such as lower temperatures and less
127 precipitation (Table S1). Hence, the high productivity sites, with greater MAI index values, was
128 located on a southern slope (*N. antarctica* dominant height of 6.1 m), the intermediate productivity
129 site was placed on a northern slope (dominant height of 3.4 m), and the low productivity site, with
130 less MAI index values, was placed on a valley bottom (dominant height of 3.1 m) (Table S1)(Coulin
131 et al. 2019).

132 Vegetation was dominated by mixed *N. antarctica* in the high and intermediate productivity
133 sites were *Schinus patagonicus*, *Lomatia hirsuta*, and *Embothrium coccineum* codominate the stand,
134 and by pure *N. antarctica* in the low productivity site; the only tree species present in all three sites
135 was *N. antarctica*. In the high and intermediate productivity sites, soils were dominated by the
136 group of Hapludands, with dark color, sandy texture, lose structure and abundant presence of roots.
137 The intermediate productivity site had shallower soils and presence of rocks. The groups of
138 Udivitrands were dominant in the low productivity site with ocher color, poor abundance of roots
139 and less soil depth (Table S1). Elevation across sites range from 790 m to 840 m.

140

141 2.1. Harvesting treatments

142 At each site eight 31.5 m x 45 m plots were selected. Between 2013 and 2014 six plots
143 were harvested in six strips of increasing width (1.5, 2.5 and 3.5 m; along the plots) with the two
144 remaining plots serving as controls, resulting in 0, 30 %, 50 % and 70 % approximately of basal
145 area removal, respectively. All stems with more than 4 cm of diameter were classified as firewood
146 leaving smaller branches and leaves in the intervention strips (Coulin et al. 2019; Nacif et al. 2020;
147 Carrón et al. 2020). The design of the interventions were relatively conservative, as we performed a
148 high number of harvesting strips of relatively low width (i.e. low space factor: relation between
149 strips width and canopy height) when compared with common management in other forest

150 ecosystems (Ishii et al 2008, Makinnen et al. 2006).

151

152 2.2. *Sampling design*

153 In all plots each species cover was determined by visual estimation in quadrants of 1 m². We
154 used four quadrants per plot which were placed in the four cardinal points at 2.5 meters from the
155 plot center. Measurements were done during spring of 2015 and 2018. In each measurement all
156 species were identified.

157 We chose 6 functional traits related to community response to harvesting disturbance
158 (canopy opening), which involves mainly higher light resource availability but also lower soil
159 humidity and desiccation risk. Trait assessment was divided based on trait intra-specific variability.
160 Traits were life form (annual, perennial), growth form (tree, shrub, bambusoid semi-woody,
161 palmoid semi-woody, tussock, rhizomatous herb, rosette herb, extensive-stemmed herb), leaf
162 texture (membranous, intermediate, thought), and seed dispersal mechanism (wind, hydrochory,
163 ballistichory, zoochory, mixed). These traits have low intraspecific variability and were recorded
164 from bibliography and herbarium data, regarding life history information. We also measured
165 specific leaf area (SLA mm² mg⁻¹) and leaf chlorophyll content (SPAD® units). These traits have
166 higher intra-specific variability and were measured in the field, in 6 different individuals per site of
167 the most abundant species, following standard methodologies (Pérez-Harguindeguy et al. 2013).

168

169 2.3. *Data analysis*

170 Taxonomic diversity was assessed by estimating a Chao's q1 index per plot according with
171 the Hill numbers framework proposed by Chao et al (2014). This index can be interpreted as the
172 effective number of common species in the plot where the species are weighted proportionally to
173 their frequencies. To evaluate functional diversity, we calculated two different indexes in order to
174 represent trait diversity and dominant trait values. We decided to use Rao's quadratic entropy index
175 as it considers richness and relative abundance of traits and it is a good measure for identifying
176 assembly patterns underlying community structure (Mouchet et al. 2010). We decided to use the

177 community weighted mean (CWM) index of different traits, as it calculates the averaged trait value
178 in a given community, weighted by species abundance (Garnier et al. 2004). All index where
179 estimated using FD package (Laliberté et al. 2015) with R software (R Core Team 2017).

180 To assess patterns of diversity along disturbance gradients (harvesting intensity) and site
181 productivity on taxonomic (Chao q1) and functional diversity (Rao Q), we fitted linear mixed-
182 effects models. We used lme() function from nlme package for normal distributed data and glmer()
183 function and lme4 package for non-normal distribution data (Bates et al., 2014; Pinheiro et al.,
184 2018). At the plot level, the models considered the fixed effect of harvesting (quantitative
185 predictor), square harvesting (quantitative predictor, to show non-lineal responses to harvesting),
186 site productivity (categorical predictor), year since harvesting (categorical predictor) and plot as a
187 random effect for avoiding pseudo-replications (Zuur et al., 2009, Pinheiro et al., 2018; Coulin et al.
188 2019). Variances were modeled using VarIdent() function. Multimodel inference was performed and
189 AICc criteria was used to selected the best models following a parsimonious criterion using
190 dredge() function and MuMin package (Barton, 2009). AICc is a correction for small sample size
191 applied to the Akaike information criterion (AIC). Relative importance values of all analyzed
192 variables were calculated with the importance() function in the MuMin package, where the Akaike
193 weights are summed between all models for each explanatory variable (Coulin et al. 2019).

194 To assess patterns of community composition along harvesting intensity and site
195 productivity, we fitted linear mixed-effect models to the community weighted mean of specific leaf
196 area (CWM.SLA) and of leaf chlorophyll content (CWM.LCC). Also, a non-metric
197 multidimensional scaling (nMDS) analysis was done to qualitatively assess changes in community
198 composition considering harvesting intensity (% basal area), year since harvesting and site
199 productivity (Legendre and Legendre 2012).

200 To assess community resilience to harvesting for firewood we calculated a change index
201 proposed by Lipoma (2018), which considers the change in time (4 years) of the difference between
202 a given harvesting intensity and the control treatment (no harvesting).

$$203 \quad CI = \frac{((C_{t1} - T_{t1}) - (C_{t0} - T_{t0}))}{((C_{t1} - T_{t1}) + (C_{t0} - T_{t0}))} * (-1)$$

204 Where CI is the change index, C is the control treatment, T is a given harvesting intensity treatment
205 (30%, 50% or 70%), $t1$ refers to measurement of 2018 and $t0$ refers to the measurements of 2015.
206 The index values vary between 1 y -1; where 0 indicates no change, positive values indicate that the
207 treatment values get closer (similar) to the control values, and negative values indicates that the
208 treatment values differentiate from the control values (Lipoma 2018). We calculated this change
209 index for taxonomic diversity (Chao q1) and functional diversity (Rao Q), and for the community
210 weighted mean of specific leaf area (CWM.SLA) and leaf chlorophyll content (CWM.LCC).
211 Finally, we fitted lineal models considering treatment and site as independent variables and CI as
212 dependent variable, multi-model inference was performed and AICc criteria selected the best
213 models following a parsimonious criterion using dredge() function and MuMin package.

214

215 **3. Results**

216 The best-fit models for taxonomic diversity (Chao q1) always included the year since
217 harvest as a predictor, but year had no effect on functional diversity (Figure 1). In general, diversity
218 had different patterns of response depending on site identity, where high, intermediate and low
219 productivity sites were categorized based on biomass growth through the MAI index. For the high
220 productivity site, both taxonomic and functional diversity increased with increasing harvesting
221 intensity, while the intermediate and low productivity sites showed diversity peaking at intermediate
222 harvesting intensities. Also, the low productivity site showed the lower levels of taxonomic
223 diversity at higher harvesting intensities (Figure 1).

224 The change in community composition under different harvesting intensities and site
225 identity can be seen by analyzing the community weighted mean of traits related to strategies on the
226 use of light and space resources. CWM.LCC varied with both site identity and year since harvest.
227 While the highest value of LCC was found at lowest harvesting intensities in the high productivity,
228 the opposite was found for the intermediate and low productivity sites (Figure 2). CWM.SLA was
229 not explained by site identity, and only a small effect of year since harvesting was found (Figure
230 S2). All models and the relative importance of the evaluated variables are presented in Table S2 and

231 S3. Also, we present the CWM of categorical traits in Table S4, in a way to visually link the most
232 abundant category of traits such as leaf texture, growth form and seed dispersal mechanism with the
233 harvesting intensity gradient and different site identities. Finally, the nMDS analysis showed
234 different patterns of response between taxonomical and functional composition. While a clear
235 pattern separating composition based on site identity and year since harvesting was found for
236 taxonomic composition, there is no clear pattern of response for functional composition (Figure 3
237 and S3). For taxonomic composition, sites clearly group based on identity, and almost all showed a
238 similar trajectory with time, approaching each other after 4 years of initial harvesting (Figure 3).

239 The change index (CI) indicates changes in diversity and community composition after five
240 years of initial harvesting disturbance. We evaluated this index for Chao q1, Rao Q, CWM.SLA and
241 CWM.LCC. For both community weighted mean indexes (CWM.SLA and CWM.LCC), best-fit
242 models were null models with low relative weight of potential explanatory variables (Table S2 and
243 S3). The best-fit model of the change index of taxonomic diversity (Chao q1) included only site
244 identity as an explanatory variable (Table S2 and S3, Figure S4). The best-fit model of functional
245 diversity (Rao Q) included site identity, harvesting intensity and the interaction between them as
246 explanatory variables, although the importance of the effects was low (Table S2 and S3). High and
247 intermediate productivity sites presented a similar response, with the highest and lowest harvesting
248 intensities being most similar to the control than intermediate harvesting intensities after four years
249 since disturbance; while low productivity sites showed that at lower harvesting intensities the
250 community differentiates the most from control sites (Figure 4).

251

252 **4. Discussion**

253 Land use change is among the most important drivers of biodiversity loss, and its effects
254 strongly depend on the type, frequency and severity of disturbance (IPBES 2019). This opens
255 challenges and opportunities for natural resource management on private lands. Most of the woody
256 species that our study sites are heliophiles and resprouters (Rusch et al. 2017) and exhibit fast initial
257 growth after disturbance (Landesmann et al. 2016), but the response of the community to initial

258 disturbance and potential recovery is not well understood. We designed this study in order to assess
259 patterns and mechanisms on community response and resilience to different intensities of harvesting
260 in a temperate mixed secondary woodland of northwest Patagonia. Our main results show that both
261 taxonomic and functional diversity have similar responses to harvesting intensity under different
262 site identity (firewood biomass growth), and that these responses supports the prediction of the
263 dynamic equilibrium model. This means that in sites with higher productivity (biomass growth)
264 diversity is not threaten by intermediate or high harvesting intensities for firewood (branches with
265 diameters higher than 4cm), when harvest occurs in small in strips of relatively low space factor
266 (relation between strips width and canopy height) compared with other thinning strips widely
267 performed in other forest ecosystems (Ishii et al 2008, Makinen et al. 2006), and leaving smaller
268 branches and leaves in the intervention strips. But community composition did not respond in the
269 same way than diversity indices. When considering traits related to the leaf economic spectrum
270 such as leaf chlorophyll content the response seems to be site-specific. Interestingly, the resilience
271 of taxonomic diversity and specific traits was not affected by harvesting intensity, and only the
272 change index of functional diversity could be explained by an interaction of harvesting intensity and
273 site identity. The specificity of these results shows the importance of considering local conditions
274 and disturbance intensity when planning harvesting for firewood interventions in temperate forests.

275 Taxonomic and functional diversity changed according to our first working hypothesis; thus,
276 we present evidence for the use of the dynamic equilibrium model (DEM, Huston 2014) in natural
277 resources interventions. We found that the response of taxonomic and functional diversity to
278 harvesting intensity varies according to site productivity, a key finding for sustainable management.
279 Specifically, sites with higher biomass growths (high productivity site) showed a positive response
280 to harvesting intensity, while sites with intermediate and lower biomass growth showed higher
281 diversity at intermediate and low harvesting intensities, respectively. Our results agree with
282 previous work in Patagonia's mix temperate woodlands regarding pollinators diversity (Coulin et al.
283 2019) and plant taxonomic diversity (Goldenberg 2020) and expands on functional diversity.
284 Similar results had been reported for other type of environments (Agard et al. 1996; Laliberté et al.

285 2013), but to our best knowledge this is the first study to evaluate diversity and composition, taking
286 in consideration ecosystem functioning, and accept DEM as a useful tool for guiding management
287 decisions in temperate forests.

288 For the analysis of functional diversity, we chose Rao's quadratic entropy because it allowed
289 us to consider both functional richness and divergence of the community, as it considers species
290 abundance in the analysis (Botta-Dukát 2005). Moreover, the use of Rao's index had been found to
291 properly identify assembly rules as it can differentiate limiting similarity from niche filtering and
292 random assembly (Mouchet et al. 2010). In our case study, higher values of functional diversity
293 imply higher functional richness and/or functional divergence of traits related to the use resources
294 such as light and space. High harvesting intensities may reduce competitive ability of dominant
295 species, resulting in a release of available resource that enable the coexistence of species with a
296 broad range of functional traits (Miedema et al. 2019; Williams et al. 2020).

297 Bringing functional diversity and composition together at the community level shed light on
298 the process behind community change after disturbance. We predicted changes in community
299 composition along the leaf economic spectrum (mainly light use availability) under different
300 harvesting intensities, but our hypothesis was partly demonstrated as no straightforward pattern
301 emerged from the analysis. A trend can be seen, where the change in understory composition after
302 harvesting varied depending on site identity. In the high productivity sites under low harvesting
303 intensities the community was characterized by a resource conservation strategy (high leaf
304 chlorophyll content, shrubs). Higher harvesting intensity changed community towards a fast
305 resource use strategy, with lower leaf chlorophyll content. In this new community, rhizomatous
306 herbs and other species with intermediate to membranous leaf texture gain importance. But there
307 was no effect of site identity or harvesting intensity on specific leaf area. Notably, the opposite
308 response was found in low productivity sites, where undisturbed communities are characterized by
309 lower levels of leaf chlorophyll content and higher harvesting intensity changed the community
310 towards a resource conservation strategy, mainly dominated by bambusoid semi-woody and shrubs.
311 These trade-offs in competitive ability had been recognized as important in predicting diversity

312 response when disturbance returns resources to the ecosystem (Haddad et al. 2008), which is the
313 case of harvesting for firewood in temperate forests. The lack of a clear pattern to the predictions of
314 the leaf economic spectrum trade-off at a local scale agrees with recent findings, where trait
315 plasticity seems to play an important role in the response to local environmental responses, and
316 similar species may respond differently to different drivers (Wright and Sutton-Grier 2012;
317 Williams et al. 2020). In this sense, our results call for precaution in the generalization of
318 conclusions and contributes to forest management by pinpointing that some important ecological
319 responses to disturbances might be site-specific.

320 Other key issue for planning sustainable natural resource management is the resilience of the
321 system (Gunderson and Holling 2002; Folke et al. 2004). Interestingly, the resilience of community
322 structure (i.e. resilience in taxonomic diversity) and of specific functional traits related to the leaf
323 economic spectrum were not affected by harvesting intensity. Only site identity (relative to biomass
324 production) was an important factor for the resilience in taxonomic diversity; while both site
325 identity and harvesting intensity (and the interaction) were important factors accounting for
326 differences in functional resilience (i.e. resilience in functional diversity), although with a low
327 magnitude. Under lower and higher harvesting intensities functional diversity resembles a non-
328 disturbed community more than under intermediate harvesting intensities in high and intermediate
329 productivity sites. But in low productivity sites the response to lower and higher harvesting
330 intensities presents opposite patterns, and sites with higher harvesting intensities were the ones that
331 resembles the most to a non-disturbed community. Thus, we reject our hypothesis, as no clear
332 pattern could be identified regarding important factors affecting the resilience of these woodlands to
333 harvesting intensity. Here we used the change index as a metric of resilience (Lipoma 2018) by
334 focusing on recovery based on a baseline from an undisturbed control treatment after 4 years since
335 the disturbance. This index does not allow us to analyze the trajectory of each site after disturbance
336 but rather compare it with a reference situation. In this sense, it gives useful information regarding
337 the recovery component of resilience to harvesting for firewood (Ingrisch and Bahn, 2018), because
338 treatments were completely randomized and control sites were part of the same stand as harvested

339 sites, with similar ecological (species richness, seed bank) and climatic conditions.

340 . Considering the productive, social and environmental value of native shrublands of north
341 Patagonia, the management of forests with energy purposes should be based on partial cuttings,
342 leaving permanent forest cover to ensure the provision of ecosystem services and system
343 sustainability (*sensu* National Law 26.331). In this way, harvesting in strips is an innovative
344 management approach in north Patagonia that could be both environmental and economically
345 sustainable (Goldenberg et al. 2018).

346 Finally, to effectively apply these results to forest management, information about
347 ecosystem dynamic is needed. In the case of mixed woodlands in northwest Patagonia, a state and
348 transition model had been proposed based on several background knowledge (Peri et al. 2017;
349 Rusch et al. 2017). Our results shed light on the mechanisms behind the proposed positive and
350 negative transitions of the reference state. For example, we show that in sites with greater firewood
351 biomass production (high productivity), high harvesting intensities may not generate a negative
352 transition from mixed shrubland to an open grasslands-woodland, as diversity peaked at the
353 beginning and at the end of the experiment, and the traits that characterizes a shrub land resembled
354 the control treatment in four years. But caution should be taken when managing a lower
355 productivity woodland, as the response pattern is the opposite.

356

357 **5. Conclusion**

358 This study was designed to understand patterns and mechanisms of biodiversity response to
359 harvesting for bioenergy in temperate woodlands (Goldenberg et al. 2018), taking into consideration
360 the complexity of the disturbance in space (harvesting intensity and site identity) and time
361 (resilience). We show evidence for the use of the dynamic equilibrium model (Huston 2014) in
362 natural resources interventions considering both taxonomic and functional diversity and validate it
363 as a useful tool for guiding management decisions in temperate woodlands. But when functional
364 composition was analyzed, the response pattern was not as straightforward as the one found for
365 diversity indices. For example, in sites with greater firewood biomass production (high

366 productivity) under low harvesting intensities the community was characterized by a resource
367 conservation strategy (high leaf chlorophyll content, shrubs), and higher harvesting intensity
368 changed community towards a fast resource use strategy (lower leaf chlorophyll content, herbs).
369 But the opposite response was found in low productivity sites. These results call for precaution in
370 the generalization of conclusions and contributes to forest management by pinpointing that some
371 important ecological responses to disturbances might be site-specific. Finally, the resilience of
372 taxonomical diversity and functional composition was not affected by harvesting intensity. For
373 functional diversity, we found that the capacity of these woodlands to resemble undisturbed sites
374 four years after the elimination of the disturbance will depend on disturbance intensity and site
375 productivity. Future studies considering longer periods of recovery time are needed to effectively
376 plan sustainable management, and the functional composition of the community needs to be
377 addressed to ensure the analysis of the dynamic of the system.

378

379 **6. CRediT authorship contribution statement**

380 Verónica Chillo: Conceptualization; Methodology, Software, Data curation; Formal analysis;
381 Investigation, Writing - original draft, Writing - review & editing. Matías Goldenberg:
382 Conceptualization, Methodology, Software, Writing - review & editing. Néstor Pérez Méndez:
383 Software, Data curation; Formal analysis; Investigation; Writing- review & editing. Lucas
384 Alejandro Garibaldi: Funding acquisition; Supervision; Project administration; Conceptualization;
385 Investigation, Writing- review & editing.

386

387 **7. Declaration of Competing Interest**

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

390

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527

528 **Figure 1.** Best-fit models of changes in taxonomic diversity (Chao q1) and functional diversity
529 (Rao's Q) depending on harvesting intensity (% of basal area) for sites with different productivity
530 (high, intermediate and low biomass growth). Colors of lines and dots represents different years
531 since harvesting: first year in gray (2015), fourth year in black (2018). Dashed line represents a
532 model without year effect.

533

534 **Figure 2.** Best-fit models of changes in community weighted mean values of leaf chlorophyll
535 content (CWM.LCC) depending on harvesting intensity (% of basal area) for sites with different
536 productivity (high, intermediate and low biomass growth). Colors of lines and dots represents
537 different years since harvesting: first year in gray (2015), fourth year in black (2018).

538

539 **Figure 3.** Non-metric multidimensional scaling (n-MDS) analysis grouping sampling sites with
540 different harvesting intensities and site identity (high, intermediate and low biomass growth) based
541 on taxonomic and functional composition.

542

543 **Figure 4.** Change index of functional diversity (Rao's Q) as a function of harvesting intensity (% of
544 basal area) and site identity (high, intermediate and low biomass growth). Positive values of CI
545 indicate higher similarity with not-harvested sites, 0 indicates no change, negative values indicate
546 differentiation from not-harvested sites.

547

548

549 **Supporting information**

550 **Figure S1:** Experimental sites along Rio Negro province (Argentina).

551

552 **Figure S2.** Community weighted mean of specific leaf area as a function of harvesting intensity and
553 site identity in north-western Patagonia

554

555 **Figure S3.** Non-Metric Multidimensional Scaling (nMDS) analysis showing the dispersion of
556 functional traits in harvesting treatments and site identity.

557

558 **Figure S4.** Change index of taxonomic diversity as a function of harvesting intensity and site
559 identity in north-western Patagonia

560

561 **Table S1.** Study sites characteristics. Mean values \pm standard deviations.

562

563 **Table S2.** Relative importance of each predictor variable for the general model

564

565 **Table S3.** Estimated values for each predictor variable for the general models.

566

567 **Table S4.** Most abundant categories at each site, based on CWM index of categorical traits.

568

569

570

Figure 1

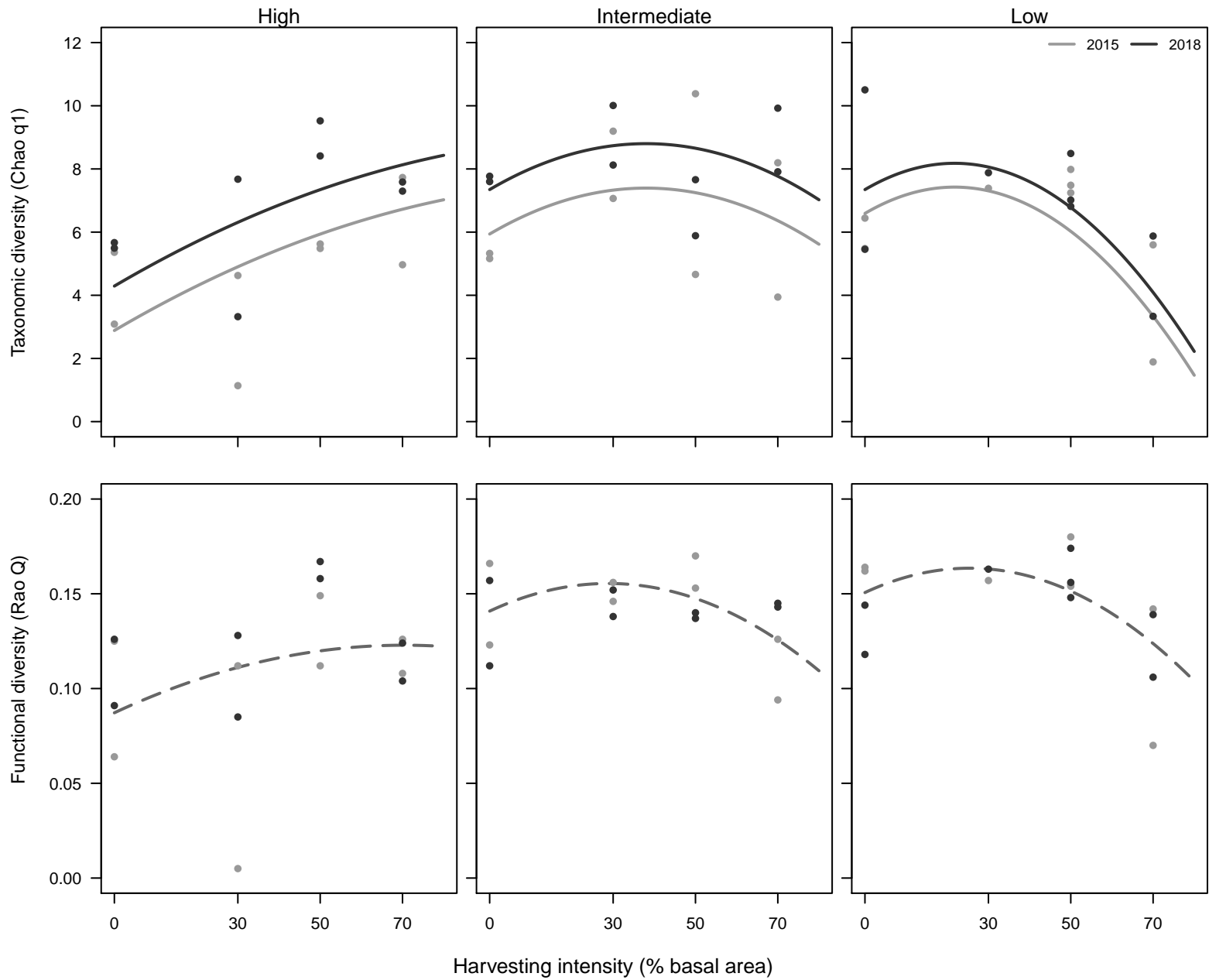


Figure 2

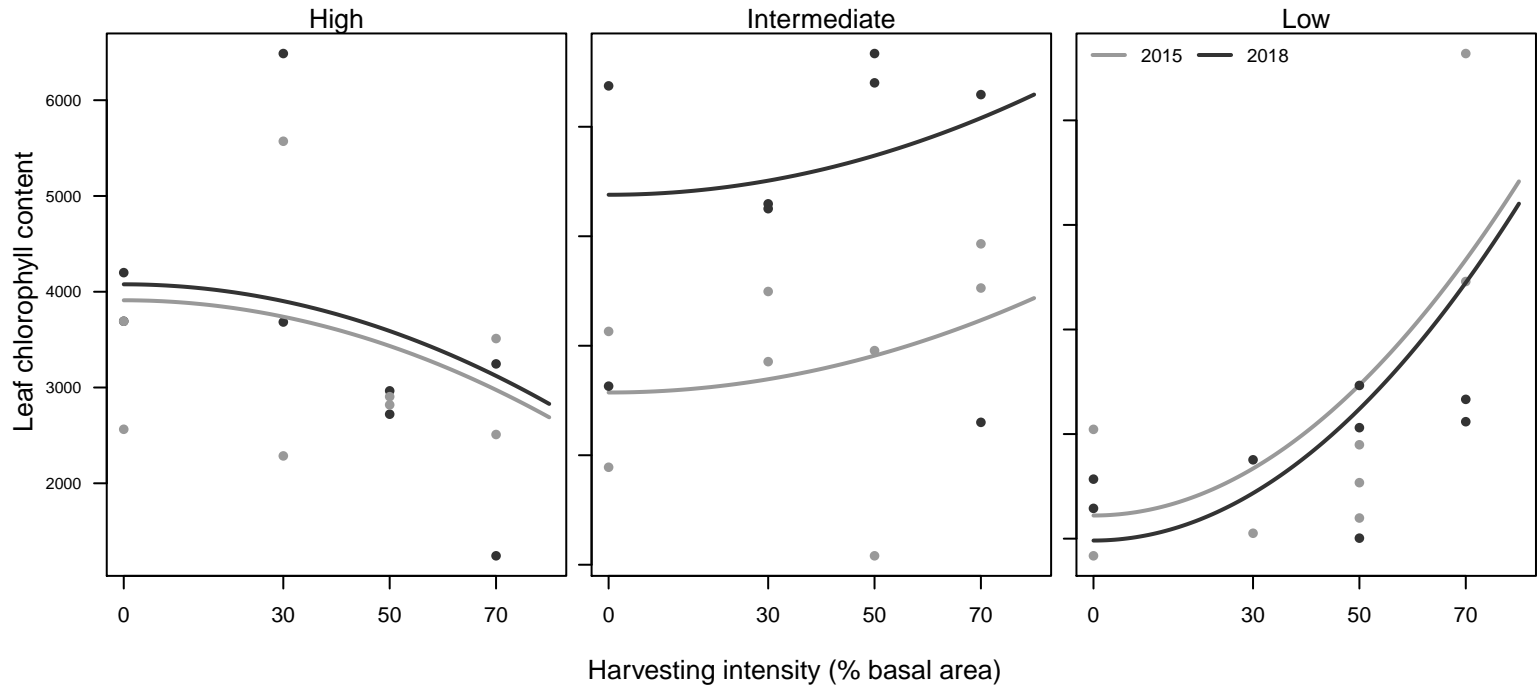


Figure 3

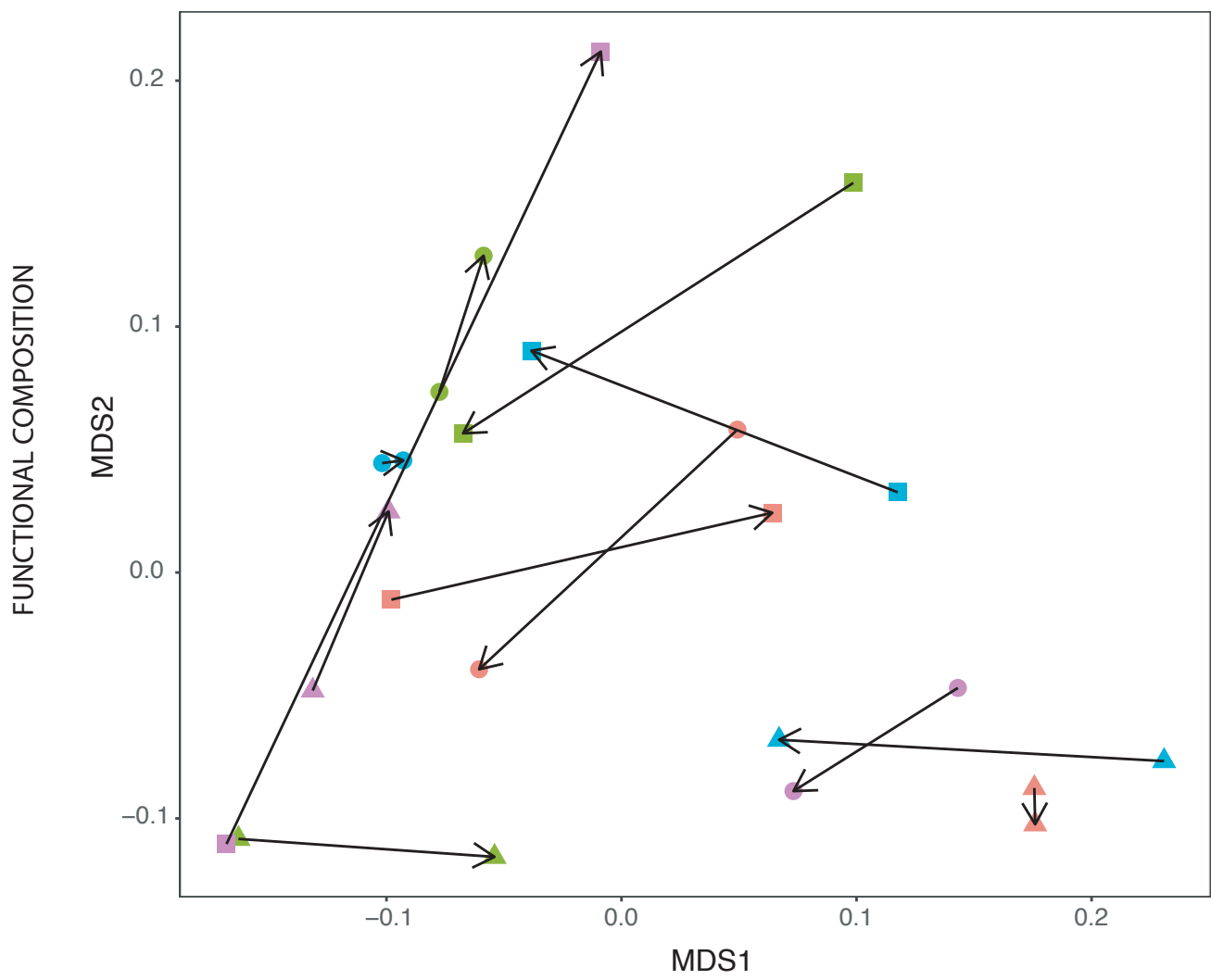
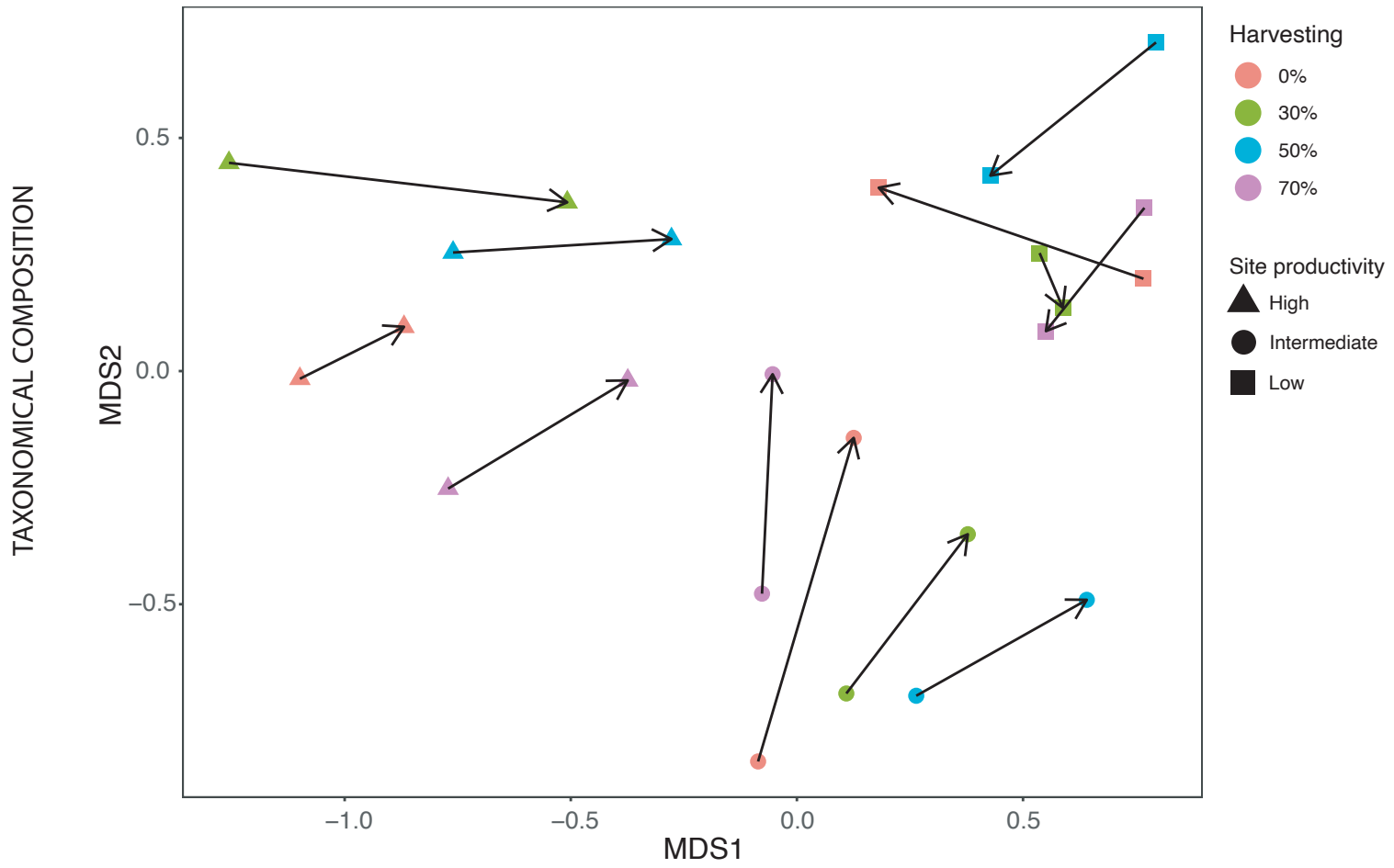


Figure 4

