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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-phisicological 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

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41
1. Introduction

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

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

The understanding of the effects of forest management on ecosystem dynamics needs to consider more than just species richness and abundance. Diversity influences ecosystem functioning through the type, range and relative abundance of functional traits (Cadotte et al. 2011). Thus, the analysis of functional diversity allows us to understand the effect of species gains and losses on the 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; m^3 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 m^2. 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 mm^2 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 underlaying 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    $$CI = \frac{((C_{t_1} - T_{t_1}) - (C_{t_0} - T_{t_0}))}{((C_{t_1} - T_{t_1}) + (C_{t_0} - T_{t_0}))} * (-1)$$

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

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

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

4. Discussion

Land use change is among the most important drivers of biodiversity loss, and its effects strongly depend on the type, frequency and severity of disturbance (IPBES 2019). This opens challenges and opportunities for natural resource management on private lands. Most of the woody species that our study sites are heliophiles and resprouters (Rusch et al. 2017) and exhibit fast initial 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, Makinnen 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 a. 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

productivity) under low harvesting intensities the community was characterized by a resource

conservation strategy (high leaf chlorophyll content, shrubs), and higher harvesting intensity

changed community towards a fast resource use strategy (lower leaf chlorophyll content, herbs).

But the opposite response was found in low productivity sites. These results call for precaution in

the generalization of conclusions and contributes to forest management by pinpointing that some

important ecological responses to disturbances might be site-specific. Finally, the resilience of

taxonomical diversity and functional composition was not affected by harvesting intensity. For

functional diversity, we found that the capacity of these woodlands to resemble undisturbed sites

four years after the elimination of the disturbance will depend on disturbance intensity and site

productivity. Future studies considering longer periods of recovery time are needed to effectively

plan sustainable management, and the functional composition of the community needs to be

addressed to ensure the analysis of he dynamic of the system.

6. CRediT authorship contribution statement

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

7. Declaration of Competing Interest

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

8. Acknowledgments

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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 specif 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 ± 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.
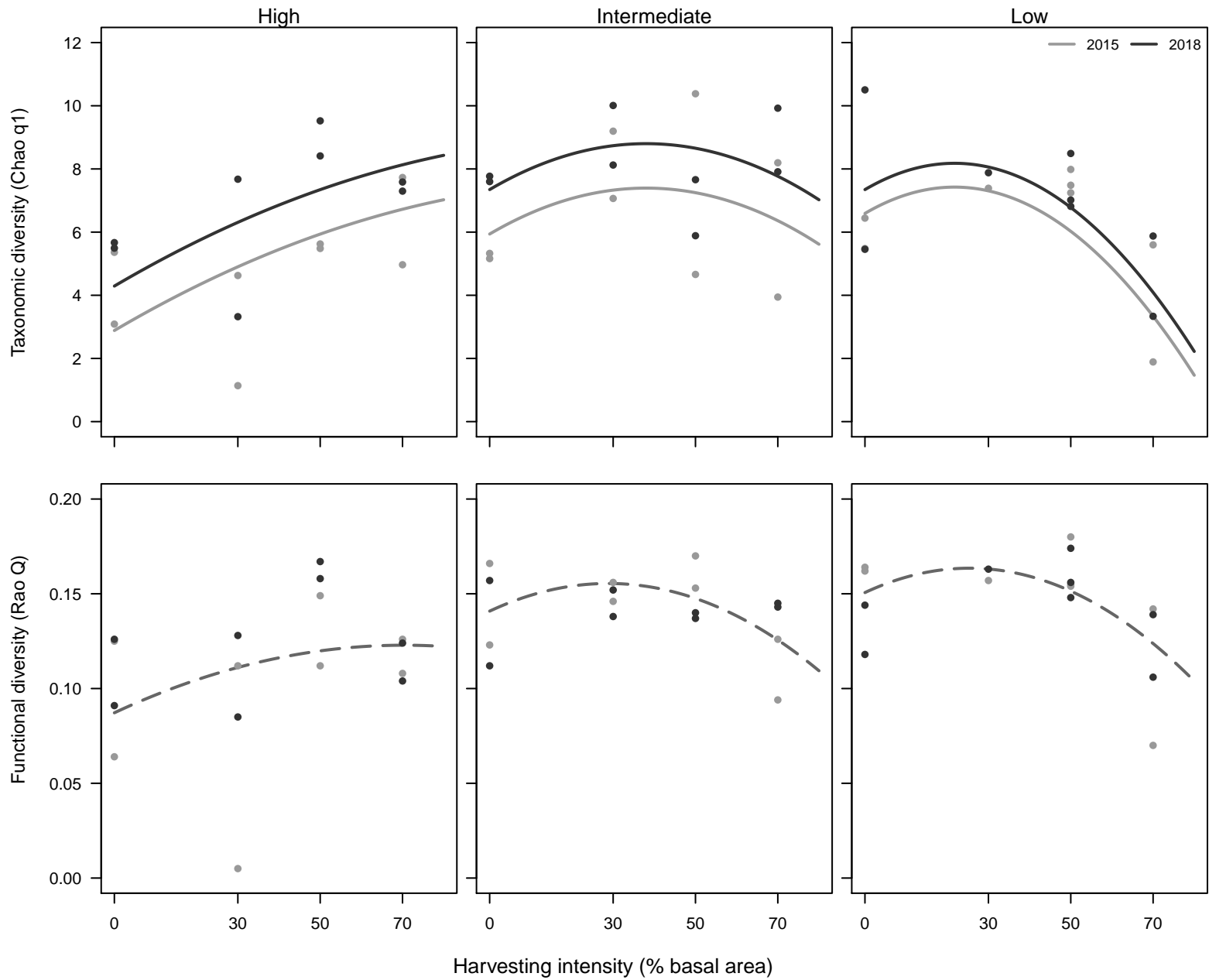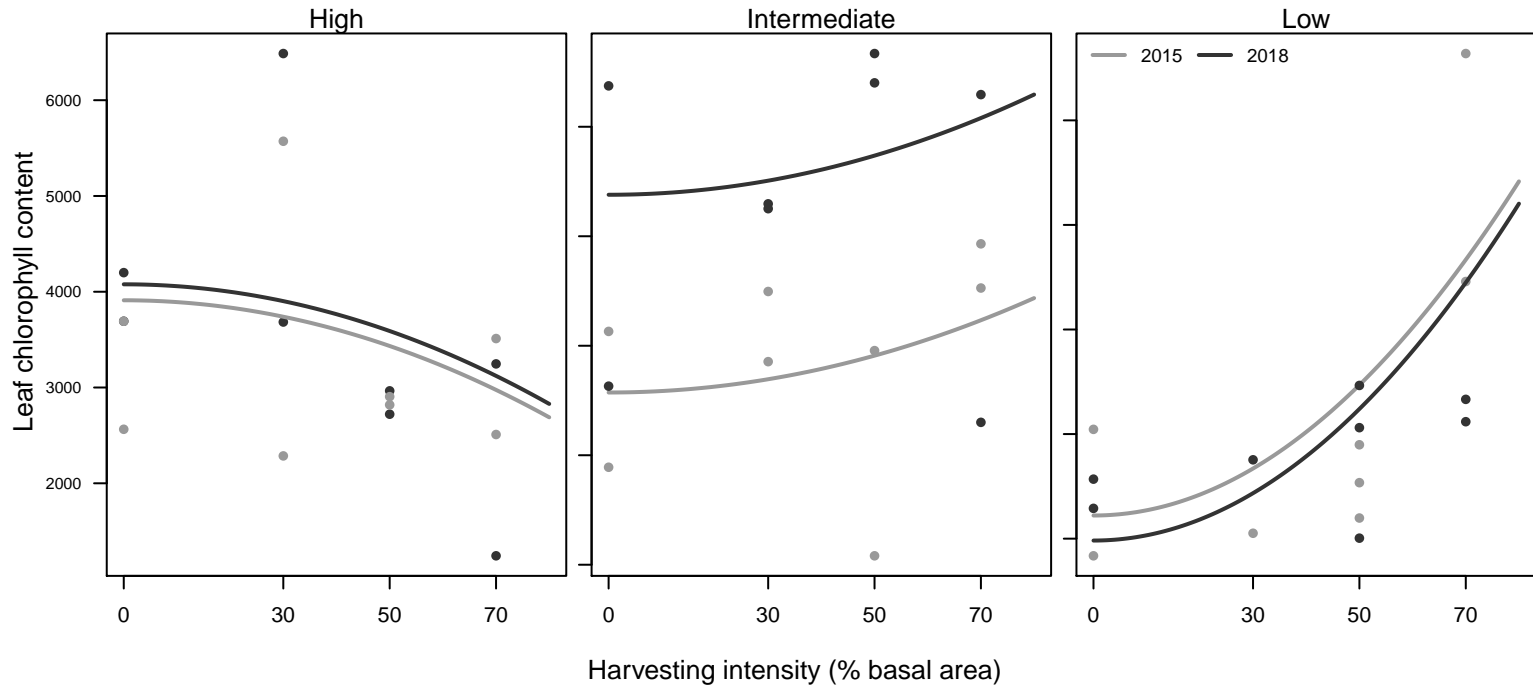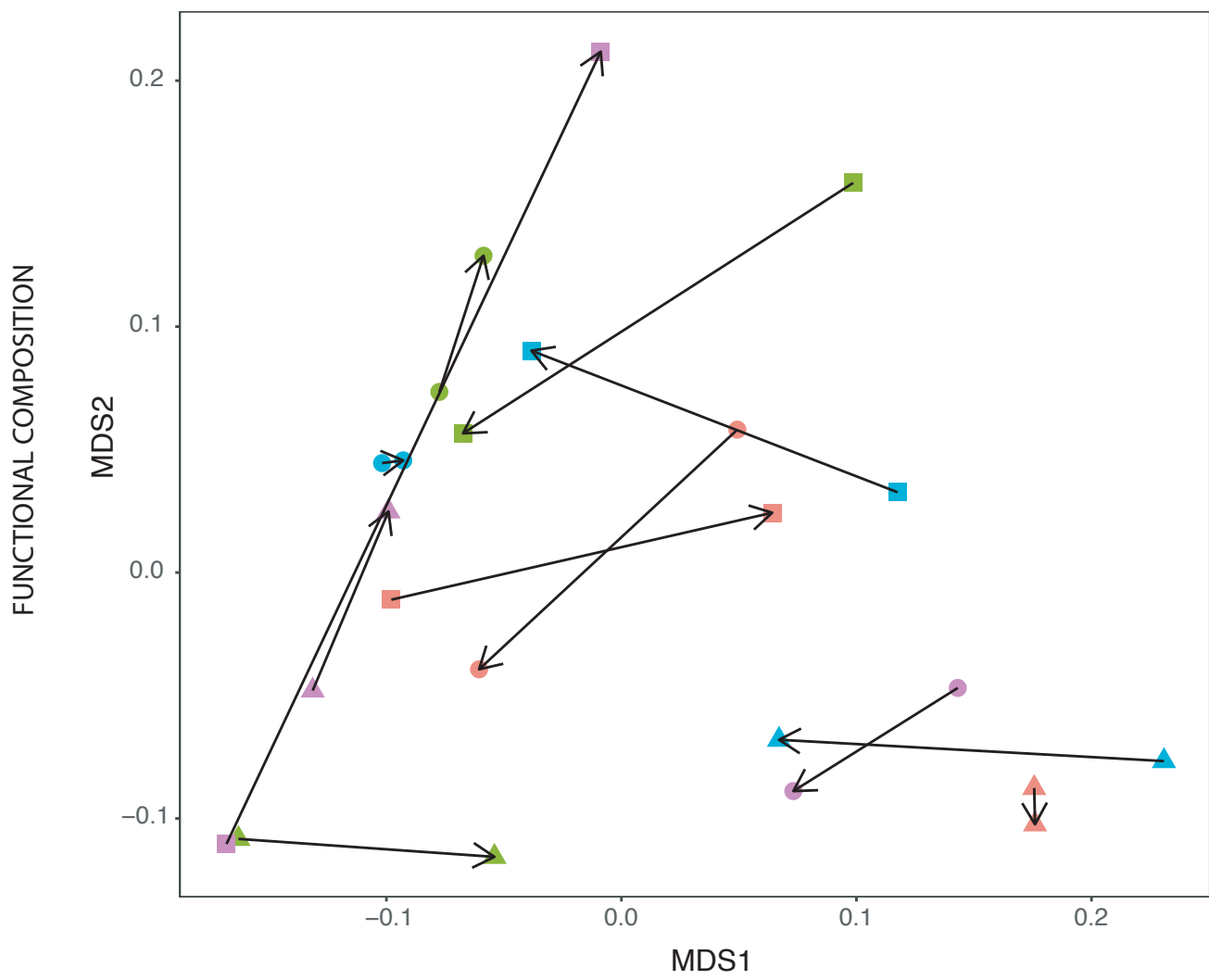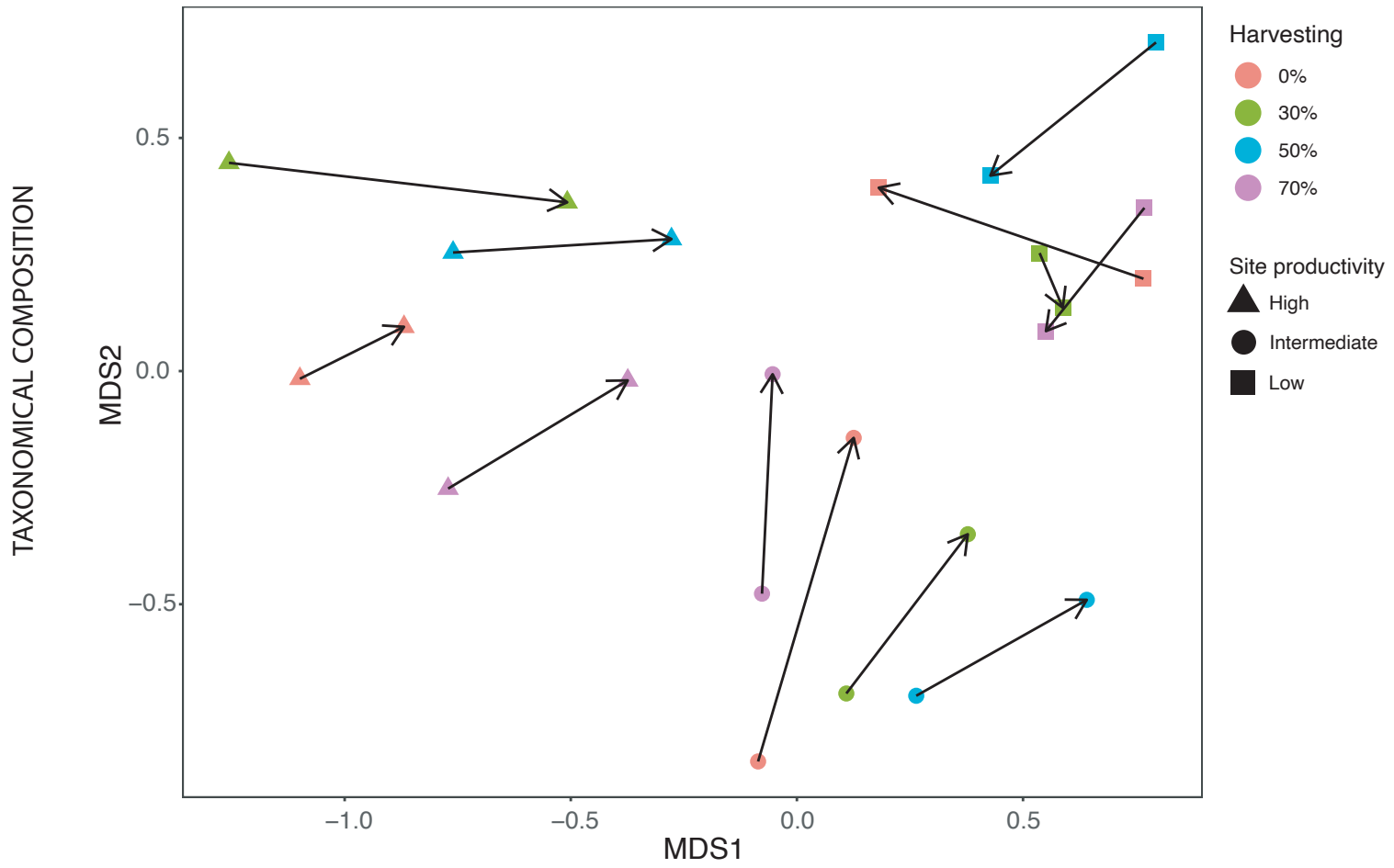
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Figure 1

Figure 1

Figure 2

Leaf chlorophyll content vs. Harvesting intensity (% basal area) across High, Intermediate, and Low panels, with 2015 and 2018 trends.

Figure 3

Figure 4