

1 **Economic susceptibility of fire-prone landscapes in natural protected areas of the**
2 **southern Andean Range**

3 **Abstract**

4 Large fires are the most important disturbances at landscape-level due their ecological
5 and socioeconomic impacts. This study aimed to develop an approach for the
6 assessment of the socio-economic landscape susceptibility to fire. Our methodology
7 focuses on the integration of economic components of landscape management based on
8 contingent valuation method (CVM) and net-value change (NVC). This former
9 component has been estimated using depreciation rates or changes on number of arrivals
10 to different natural protected areas after large fire occurrence. Landscape susceptibility
11 concept has been motivated by the need to assist fire prevention programs and
12 environmental management.

13 There was a remarkable variation in annual economic value attributed to each protected
14 area based on the CVM scenario, ranging from 40,189-46,887 \$/year ("Tolhuaca
15 National Park") to 241,000-341,953 \$/year ("Conguillio National Park"). We added
16 landscape susceptibility using depreciation rates or tourists arrivals decrease which
17 varied from 2.04% (low fire intensity in "Tolhuaca National Park") to 76.67% (high
18 fire intensity in "Conguillio National Park"). The integration of this approach and future
19 studies about vegetation resilience should seek management strategies to increase
20 economic efficiency in the fire prevention activities.

21

22 **Keywords:** environmental susceptibility, landscape management, contingent valuation,
23 fire behavior, net-value change

24 **1 Introduction**

25 Forest fires are an active element in the configuration and shaping of wide variety of
26 ecosystems (FAO, 2007). In this sense, fire has played a keystone role in the shaping of
27 the heterogeneous Andean landscape (González et al., 2010) and its forest dynamics
28 (Veblen et al., 1995; Donoso, 1998). Although humans have used fire since the
29 Neolithic Era (Abrams, 1992), climate change and anthropic factors are transforming
30 fire into a threat to the biodiversity and conservation (Chavardes and Daniels, 2016).

31 Catastrophic forest fires have ravaged parts of Chile, Portugal, Spain and United States
32 this year. As an example, Chile was affected by severe forest fires between January and
33 February 2017 with more than 470,000 ha under different fire intensity levels (Rivera-
34 Careaga, 2017). The National Emergency Management Authority (LEMA) catalogued this fire
35 as a "firestorm", an unprecedented phenomenon in the history of humankind (European
36 Civil Protection Agency, 2017). They highlighted the fact in a single night the fire
37 consumed 8,000 ha per hour. Comparatively, France requested support for a fire that
38 burned a total of 8,000 ha and Spain's firefighting capacity collapsed with a fire that
39 involved just 25,000 ha. The experts hypothesize that the type of fire that is being seen
40 for the first time with Chile's "firestorm" will occur in the future in several countries
41 because it is partly due to phenomena such as climate change.

42 Forest fires constitute a worldwide problem, given their serious tangible assets,
43 environmental service and landscape goods impacts (Rodríguez-Silva and González-
44 Cabán, 2010). Therefore, an increase in economic losses from wildfires has been
45 corroborated from different studies (Román et al., 2013; Chuvieco et al., 2014). In this
46 sense, large wildfires could become a threat to social valuable landscapes because of
47 climate change and fire regime change (Molina et al., 2017a). Landscape resources don't

48 usually take the form of monetary values in wildfire impacts valuation. Although
49 indirect methods are challenging, forest management should involve intangible assets,
50 mainly in natural protected areas. The high socio-economic value of protected areas
51 should lead to preventive actions, in order to preserve its tourism activity, and as a
52 consequence, its economic value. It is essential that landscape resource can be fully
53 taken into account in planning and decision-making.

54 Although tangible assets and ecological losses have immediate short and medium-term
55 importance, the disappearance or changes in landscape give rise to additional long-term
56 impacts. However, in spite of some research approaches (Rodríguez y Silva et al., 2010;
57 Castillo et al., 2013), there is lack of knowledge of the long-term economic impacts,
58 mainly in natural protected areas. The conclusions of these former studies focuses on
59 the need of a detailed study of the economic susceptibility of forest landscapes against
60 wildfires. It is essential that the socioeconomic values of the environmental services and
61 landscape goods be fully taken into account in planning and decision-making (Costanza
62 et al., 2006; De Groot, 2006). Landscape can take the terms of monetary units though
63 indirect methods such as travel cost, hedonic technique and contingent valuation
64 (CVM). CVM is the main stated preference method over the last three decades
65 (González and León, 2003; MacMillan et al., 2006; Grammatikopoulou and Olsen,
66 2013; Chen and Hua, 2015; Chatterjee et al., 2017). In spite of the CVM limitations
67 (Schläpfer et al., 2004; Hynes et al., 2011), this methodology has been used in studies in
68 order to facilitate the comparison of different management alternatives to mitigate forest
69 fires (Molina et al., 2016).

70 Different studies have evaluated the economic damages caused by fire (Butry et al.,
71 2000; Morton et al., 2003; Barrio et al., 2007), and even some of them (Rodríguez y
72 Silva et al., 2010; Castillo et al., 2013) have been developed in Andean Range.

73 However, one of the most difficult things to do in valuing the economic impact of fire
74 on natural resources is to determine the economic value lost (Rodríguez y Silva and
75 González-Cabán 2010; Román et al. 2013). Potential damages can be quantified as the
76 percentage net value change (CNV) depending on fire intensity and resources
77 sensibility' (Thompson et al., 2011). In this sense, taking potential fire behavior into
78 account is fundamental to determine the economic efficiency of fire prevention and
79 suppression activities (Duguy et al., 2007; Thompson et al., 2013). Fire behavior was
80 included by fire intensity levels (FIL) which are closely related to the impact caused by
81 the amount of heat emitted (Rodríguez y Silva et al., 2012; Castillo et al., 2017). The
82 identification of CNV caused by wildfires was expressed as depreciation rates according
83 to FIL based on the simplicity required by forest managers (Zamora et al., 2010; Molina
84 et al., 2011). These depreciation ranges were identified based on the social perception
85 using the stated social preferences. In the last part of the contingent valuation
86 questionnaire, panoramic photographs were used to estimate depreciation rates or visits
87 frequency depending on three outstanding FIL (Molina et al., 2017b).

88 Development of a multidiscipline forestry policy is not possible without considering
89 landscape susceptibility, because of the importance of recreation activities for rural
90 development and territorial planning (Molina et al., 2016). This paper aims to develop a
91 social approach for the economic assessment of the landscape susceptibility to fire. The
92 sense of this study is the identification of the landscape resource affectation and its
93 economic valuation based on tourism and recreational impacts using three important
94 natural protected areas in Chile. By extending landscape approach from the traditional
95 point of contingent valuation studies, we have incorporated landscape susceptibility in
96 order to identify effects of fire occurrence. Our approach proposes an economic
97 framework for annual landscape susceptibility (Scott and Thompson, 2015) based on

98 landscape value and net-value change (CNV). While landscape resource has been
99 valued according to CVM, CNV has been estimated based on three potential fire
100 intensity levels using estimated post-fire number of visitors. The landscape
101 susceptibility model is more complete than the former studies, since it includes
102 economic landscape value and potential fire impacts. The results could emphasize in the
103 meaningful role of the recreation resource on natural protected areas, and as a
104 consequence, the importance of fire prevention activities to landscape conservation.
105 Landscape susceptibility approach would add to learning community knowledge the
106 non-market fire impacts according to the higher probability of future large fires or
107 "firestorm" in several countries.

108

109 **2 Material and methods**

110 **2.1 Study area**

111 The climate of the Andean Range has a Mediterranean influence reflected by a winter-
112 maximum in precipitation and relatively dry summers. Annual precipitation varies
113 between 1500 and 3000 mm, although at higher altitudes the precipitation can reach
114 more than 4000 mm, the majority falling as snow. In this mountain range, most of the
115 soils are derived from ash deposited by volcanic activity (Donoso, 1998). About 97% of
116 the Araucaria forests are restricted to the upper elevations of the Andean mountain
117 range from Region VIII to Region XIV. In this study, we used three natural protected
118 areas of the IX Region of Chile ("Araucania Region") within the "Araucarias Biosphere
119 Reserve" (Figure 1).

120 - "Conguillio National Park": this area occupies about 608 km², formed mainly by
121 *Araucaria araucana* and *Nothofagus* spp. The shape of the Monkey Puzzle trees, lakes

122 and Llaima volcano increases the scenic beauty of this park. In this sense, Conguillio
123 was the most visited park in the IX Region (111,709 visitors in 2016). "China Muerta",
124 which is an adjoining National Reserve with similar landscapes, was severely burned in
125 2015 fire.

126 - "Tolhuaca National Park": this park encompasses part of the forested foothills and part
127 of the upper elevations of Andean mountain range covering about 6,500 ha. Their main
128 attractions are mixed forest landscape, wildlife, Tolhuaca volcano, small lakes and
129 thermal waters. The visitors' number was 11,270 in the last year. The Park and the
130 adjoining "Malleco National Reserve" were affected by severe forest fires in 2002 and
131 2015.

132 - "Malalcahuello National Reserve": this northern area combines *Araucaria-Nothofagus*
133 forests with a charcoal desert landscape of ash and sand (Lonquimay volcano and
134 Navidad Crater). The reserve has a surface area of about 13,800 ha including the ash
135 volcano landscape. In the border, "Nalcas National Reserve" is identified as the limit of
136 *A.araucana* distribution. In 2016, the number of visitors reached 108,618 people of
137 different nationalities.

138

139 **2.2 Economic valuation of recreational resources in natural protected areas**

140 In this study, landscape takes the form of monetary values through Contingent
141 Valuation. The purpose of this method is to obtain respondent's willingness to pay
142 (WTP) for the conservation of protected areas. We estimate WTP using the maximum-
143 likelihood method for interval-data model (Grammatikopoulou and Olsen, 2013). The
144 implicit assumption is that one underlying WTP value drives the responses to both
145 dichotomous-choice questions. If this is true, the following question provides a interval

146 around the true WTP value and the maximum-likelihood optimization model is
147 appropriate.

148 Contingent valuation information was obtained from 425 tourists in three protected
149 areas in the IX Region of Chile. Random samples of tourists were interviewed at
150 different park entrances (control points), hotels, thermal water centers and campsites
151 along the protected areas. We must note that our sample suffered from over-
152 representation of young adulthood tourists (aged between 20 and 40), and
153 underrepresented elder tourists (aged upper to 60). For this reason, we considered two
154 age ranges: < 40 years and > 40 years. The ratio of male and female respondents was
155 very equitable (208 women and 206 men). A total of 414 interviews were completed out
156 of 425, for a completion rate of 97%.

157 Firstly, the survey included a brief description of the project to prevent bias because of
158 the insufficient detail of bad-informed people who think that they will pay more money
159 in the future if they select this question option. The first part of the questionnaire
160 incorporated respondent's personal information (gender, age, job, place of origin and
161 travel motivation). The following question "Do you agree with the payment of an
162 amount of money to protected areas conservation due to the increase of the forest fire
163 risk due to the climate and socio-economic changes and its ecological value?" attempted
164 to minimize the rejection responses. The second part, using a discrete change in
165 entrance fees, was aimed at tourists and an estimation of WTP. Similar to other studies
166 (Vaux et al., 1984; Christie et al., 2006), respondents are asked whether or not they
167 would pay some specific sum or "bid". In the last part of the questionnaire, each tourist
168 was asked about its future visit or the change to other natural protected area on holidays
169 based on a fire occurrence.

170 WTP must be statistically analyzed to obtain an estimate of the mean WTP. However,
171 when respondents disagreed to pay any bid at all, different interpretations can be found:
172 enough taxes, lack of worth. Excluding these bids from the mean WTP calculation
173 would lead to biased estimates of population assessment (Hynes et al., 2011). In this
174 sense, we used two approaches (Molina et al., 2016): either taking all respondents into
175 consideration, valuing those who refuse to pay an entrance fee as zero WTP (known as
176 “all respondents”), or taking only affirmative answers into consideration (known as
177 “only affirmative respondents”). Mean WTP of each study area is multiplied by the
178 number of visitors annually to estimate its annual use value. This former data was
179 obtained from official statistics in 2016 of the Forestry Corporation
180 (<http://www.conaf.cl/parques-nacionales/visitanos/estadisticas-de-visitacion/>, August
181 2017).

182

183 **2.3 Landscape susceptibility in natural protected areas**

184 Fire behavior is not homogeneous in forest fires depending on meteorological,
185 physiographic and fuel model conditions (Finney, 1998). The level of damage could be
186 determined by the fire intensity levels (FIL) (Zamora et al., 2010; Rodríguez y Silva et
187 al., 2012; Castillo et al., 2017). In our study, Fire Intensity Levels (FIL) were identified
188 based on photographs with different fire behavior in study area fires:

189 - FIL I: surface and passive fire behavior

190 - FIL II: active fire with unburned islands and attractive elements

191 - FIL III: active fire without unburned areas

192 Depreciation rates were estimated based on the contingent valuation questionnaire
193 where three panoramic photographs were affected by these different FIL (Appendix I).
194 On this FIL ladder, each photograph represented progressive higher fire impacts.
195 Although low-intensity fires could have positive effects on fire-prone ecosystems
196 (Smucker et al., 2005), in this study fire impacts on a landscape level were viewed as a
197 negative decrease in visitors number in a short-term perspective. The depreciation rate
198 (%) was identified as the difference in the number of arrivals according to each FIL.
199 Respondents were asked about trip changes based on a fire occurrence. These
200 recreational changes or deterioration rates (DR) provide a versatile assessment tool
201 (Equation 1) for landscape susceptibility assessment based on fire intensity (Molina et
202 al., 2017b):

$$203 \quad LS = L * DR \quad (1)$$

204 where "LS" is the annual landscape susceptibility of each protected area (€), "L" is the
205 estimated annual landscape value (€) and "DR" is the depreciation rate in visitors (%)
206 based on fire intensity.

207

208 **2.4 Statistical analysis**

209 Firstly, a logit regression model was employed to test the sensitivity of our respondents
210 in relation to their socio-economic characteristics on the probability of giving protest
211 responses or non-protest responses. In general, logit analysis would be preferable in
212 situations where the normality assumption (of the sample) are violated and many of the
213 independent variables are qualitative (Chen and Hua, 2015). It is appropriate for the
214 present study in which subgroups have been clearly defined in terms of protest and non-
215 protest responses. The dependent variable takes the value 1 if the respondent states a

216 zero bid and 0 if the respondents states a positive WTP amount. In a sensitivity analysis,
217 we compared WTP for those people expressing the highest level of confidence with all
218 others. Previous studies have found that estimated WTP is more consistent with theory
219 for respondents who reported greater confidence in their answers (Grammatikopoulou
220 and Olsen, 2013; Chen and Hua, 2015).

221 One-way analysis of variance (ANOVA) was used to determinate if significant
222 differences ($p < 0.05$) existed in gender (female and male), age (< 40 years and > 40
223 years), job or economic condition (students and unemployed respondents, conventional
224 workers and high level workers) and place of origin (Araucania Region, Bio-Bio
225 Region, Metropolitana Region, Other Regions and Foreign visitors) for each natural
226 protected area and CVM scenario. If significant differences were detected, a Tukey
227 HSD test was performed to determine which specific study area and CVM scenario was
228 different from another.

229 The significant differences among the mean deterioration rates according to each FIL
230 were calculated using the non-parametric analysis. In this case, Wilcoxon test was used
231 to identify if significant differences ($p < 0.05$) existed in depreciation rate for each Fire
232 Intensity Level and different respondent characteristics. CVM scenario, gender, age, job
233 and place of origin were tested using non-parametric test. SPSS^(C) was used in all
234 analysis.

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236

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238

239 **3 Results**

240 **3.1 Economic valuation of recreational resources in natural protected areas**

241 Three subgroups can be categorized In this sense, based on respondents' answer:
242 legitimate zero respondents ("respondents cannot afford due to my budget constrains"),
243 protest respondents induced by "distrust of government" and positive WTP responses.
244 We considered two logit models based on all zero respondents and only protest zero
245 respondents. For the reduced model, we use the same variables as in the previous logit
246 model. There explanatory variables showed a statistically significant bearing on WTP:
247 education, income and place of origin (Table 1). Respondents with relative high levels
248 of education stated a significant higher WTP since education is often found to have a
249 positive impact on WTP. Higher education was found to have significantly lower
250 probability of protesting. In the case of positive WTP and protest zero respondents, the
251 level of income was significant, as economic theory would prescribe, WTP increases
252 with increasing income. The increased personal experience with the area (IX Region of
253 Chile), induced more lexicographical preferences that could translate into an increase in
254 protest zero respondents. The signs of other coefficients estimates were as expected,
255 though not statistically significant at conventional levels.

256 WTP was obtained from a conservation free payment that ranged from 0 to 45.94 US
257 dollar (\$). The percentage of respondents that proposed to abstain from paying for the
258 conservation areas was 22.52%. Most of these disagreed respondents (75.01%) were
259 related to the enough taxes paid to the government. In this sense, differences were
260 performed between the least favorable contingent valuation scenario ("all respondents")
261 and the most favorable scenario ("affirmative respondents") (Table 2). There was a
262 notable increase between 14.28% ("Tolhuaca National Park") and 29.03% ("Conguillio

263 National Park") depending on the CVM scenario. In a similar CVM scenario, WTP
264 identified two significant groups: "Conguillio National Park" and "Malalcahuello
265 National Reserve" and "Tolhuaca National Park" (Table 2). The maximum WTP was
266 attributed to this former national park in both CVM scenarios.

267 On the other hand, WTP was significant higher in female respondents of "Tolhuaca
268 National Park" when compared to the other two study areas (Table 3). In "Conguillio
269 National Park" and "Malalcahuello National Reserve", it could be observed a higher
270 male WTP than female WTP. According to the age, < 40 years respondents of
271 "Tolhuaca National Park" presented significant differences with the others (Table 4).
272 While in selected respondents of "Tolhuaca National Park" and "Malalcahuello National
273 Reserve", < 40 years increased the economic value, in selected respondents of
274 "Conguillio National Park", > 40 years showed a higher WTP.

275 Significant differences were shown between "Tolhuaca National Park" and the other
276 study areas based on students and unemployed respondents and high level workers
277 (Table 5). In all areas, the highest WTP was found in high level workers according to its
278 quality life. Therefore, significant differences were performed based on the place of
279 origin. Bio-Bio, Metropolitano, others Chilean regions and foreign respondents
280 identified significant groups among the study areas (Table 6). Selected foreign WTP in
281 "Malalcahuello National Reserve" and selected Metropolitano WTP in "Tolhuaca
282 National Park" surpassed the rest of monetary values.

283 In 2016, the number of visitors reached more than 100,000 people from different
284 nationalities in two selected areas. There was a notable variation in the annual value
285 attributed to each natural protected area depending on the CVM scenario: "all
286 respondents" and "affirmative respondents" (Table 7). The maximum economic impact

287 (341,953 \$/year) was obtained in "Conguillio National Park" according to the most
288 favorable CVM scenario. However, "Malalcahuello National Reserve" reached the
289 higher valuation (251,036 \$/year) based on the least favorable CVM scenario.

290

291 **3.2 Landscape susceptibility in natural protected areas**

292 Wilcoxon test showed the presence of significant differences ($p < 0.05$) among FIL. In
293 this sense, depreciation rate (%) was proportional to Fire Intensity Level (FIL)
294 providing the maximum landscape impact to FIL III. In terms of visit frequency, the
295 presence of tourists could decline annually from 2.04% to 76.67% according to FIL and
296 protected area (Figure 2). Non-parametric test identified two significant area groups due
297 to the similar behavior for depreciation rates. In FIL II and FIL III, we could observe a
298 group of respondents who would visit the area depending on the entrance fee and the
299 price of recreational activities.

300 There are many respondent characteristics that are likely to be related to the propensity
301 to respondent. We found significant differences in rate of depreciation focusing on
302 CVM scenario and economic condition of the respondents (Table 8). Under this former
303 respondent characteristic, two groups (student and unemployed and conventional and
304 high level workers) were performed. Therefore, depreciation rate according to place of
305 origin was very heterogeneous based on each protected area. Finally, there are not
306 significant differences based on gender and age classification (Table 8).

307 We provided the landscape susceptibility using the annual economic value that was
308 generated from each WTP scenario (Table 7) and the mean depreciation rate for each
309 FIL and protected area (Figure 2). Furthermore, differences of depreciation rates could
310 be observed using or not using the conditional respondents who would visit the affected

311 area depending on the entrance fee and the price of recreational activities. Remarkable
312 differences were observed in annual landscape susceptibility according to each
313 protected area and FIL (Table 9). In this sense, "Conguillio National Park" annual
314 susceptibility varied considerably, ranging from \$9,182-13,028 to \$184,778-262,175
315 (Table 9). While in "Tolhuaca National Park" landscape susceptibility ranged from
316 \$820-956 to \$28,707-34,448, in "Malalcahuello National Reserve" varied from
317 \$23,120-27,390 to \$ 133,877-166,422.

318

319 **4 Discussion**

320 Forest fires constitute a worldwide problem according to its associated socio-economic
321 and ecological impacts (Román et al., 2013; Chuvieco et al., 2014). The current large
322 fire frequency and fire intensity are increasingly becoming a growing global concern for
323 woodlands. Fire regime change has homogenized forests affecting its landscape value
324 and bio-diversity (Chavardes and Daniels, 2016; Molina et al., 2017a). In this sense, our
325 study area has been globally designated as a main conservation eco-region, "hotspot" of
326 biodiversity (Myers et al., 2000). This approach shows the potential fire impacts
327 associated to recreational resources in three "hotspot" or natural protected areas.

328 Annual landscape susceptibility varied considerably in the Chilean protected areas
329 ranging from \$34,448 to \$262,175 (maximum recreational impact). However,
330 contingent valuation method (CVM) is a stated preference methodology that provides
331 respondents the possibility to refuse a payment for protected area conservation. When
332 respondents disagreed to paying any entrance fee (22.52% of the respondents) are
333 excluding of the statistical analysis, a selection bias problem could be generated
334 (Schläpfer et al., 2004; Hynes et al., 2011). For this reason, this approach allows us to

335 compare two scenarios (all respondents and affirmative respondents) of WTP and
336 annual value showing significant differences between them. The annual recreation value
337 increased from 14.38% ("Tolhuaca National Park") to 29.16% ("Conguillio National
338 Park") using only affirmative respondents. Differences were performed based on the
339 economic condition and place of origin for all selected areas. The economic condition is
340 associated to the amount of money that can be paid for annual conservation. The place
341 of origin could be related to the economic ranking of each Chilean Region and foreign
342 countries.

343 Landscape goods are rarely incorporated into territorial planning, even though this
344 resource could constitute a large proportion of the ecosystem value, mainly in protected
345 areas (Costanza et al., 2006; De Groot, 2006; Román et al., 2013). An adequate
346 preventive management of Mediterranean landscape requires the knowledge of
347 landscape susceptibility (Scott and Thompson, 2015). This research has proposed an
348 integrated landscape susceptibility framework from landscape valuation and net-value
349 change (NVC) using fire intensity (Molina et al., 2017b). Fire intensity can be
350 represented by intensity scales similar to other European approaches (Rodríguez y Silva
351 et al., 2012) and Chilean studies (Castillo et al., 2017). The use of three FIL or fire types
352 belongs to the simplicity required by respondents to identify differences in the impact
353 caused by each fire type. These FIL can directly support the estimation of the NVC of
354 the natural resources (Zamora et al., 2010). In this paper, the NCV was expressed in
355 term of a reduction on protected area visits using social stated preferences. As an
356 example, tourists number could decrease in "Conguillio National Park" from 3.81%
357 (FIL I) to 76.67% (FIL III). Respondents, who would return to the burned area
358 depending on the entrance fee and the price of recreational activities, were higher in FIL

359 II compared to the rest. There is an increase of the number of post-fire visits in FIL III
360 according to the furthest respondents (Other Regions and Foreign tourists).

361 The depreciation rate varied significantly according to fire intensity similar to other
362 studies (Zamora et al., 2010; Rodríguez y Silva et al., 2012). While FIL II would
363 decrease number of tourists from 10.09% ("Malalcahuello National Park") to 32.66%
364 ("Tolhuaca National Park") in relation to surface fire or FIL I, FIL III would reduce the
365 number of visitors from 46.75% ("Malalcahuello National Park") to 72.86%
366 ("Conguillio National Park"). "Conguillio National Park" and "Tolhuaca National Park"
367 have shown the highest depreciation rates. This fact could be related to the recent
368 occurrence of large fires in these protected areas or their surrounding areas, where fire
369 impacts are still observed.

370 Although there was an increase of 20.77% of the number of tourists in protected areas
371 in IX Region of Chile (2015), Forestry Corporation statistics
372 (<http://www.conaf.cl/parques-nacionales/visitanos/estadisticas-de-visitacion/>, August
373 2017) showed a reduction of 6.33% tourists after "Tolhuaca fire". If we added both
374 values, (tourists increase in the Region and visitors decrease) we would reach 27.09% of
375 depreciation rate that is very similar to the one observed in social analysis from FIL II
376 (Figure 2). In the case of "Malleco National Reserve", after fire occurrence (2015),
377 there was a decrease of 26.46% of visitors. If regional tourists increase is considered,
378 the reduction would increase at 47.23%. These fire impacts differences could be
379 associated to the burned area and fire intensity in both Chilean protected areas
380 (differences in FIL). In spite of the differences, highest fire intensities (FIL II and FIL
381 III) would play a keystone role in the economic rural development of these Chilean
382 wilderness areas.

383 In European natural parks, large fires on some protected areas caused a dramatic
384 reduction in number of arrivals. The number of arrivals decreased between 23.65% and
385 91.01% (Molina et al., 2017b). According to the Association of Rural Hotels, “fires
386 stopped the arrival of thousands of tourists. Despite the number of confirmed
387 reservations, eco-tourism decreased 40% since fire occurrence. This decrease in the
388 annual number of visitors had a significant impact on new enterprises projects, which
389 focused on the development of rural economy” (Molina et al., 2017b).

390 Landscape susceptibility is a meaningful component of forest fire management (Castillo
391 et al., 2013; Chuvieco et al., 2014). We used two scenarios ("all respondents" and
392 "affirmative respondents") based on uncertainty associated with the sampling bias and
393 CVM method similar to other approaches (Molina et al., 2016). There was an
394 outstanding difference in annual landscape susceptibility per unit area depending on the
395 natural protected area (Figure 3). The snow centre and thermal water resorts make
396 Malalcahuello area attractive to respondents as a travel destination, and as a
397 consequence, one of the most visited protected areas in Chile despite of its limited size.
398 According to the highest fire intensity, while "Malalcahuello National Reserve" annual
399 susceptibility ranged from 10.47 \$/ha*year to 13.01\$ /ha*year, "Conguillio National
400 Park" (3.04-4.31 \$/ha*year) and "Tolhuaca National Park" (4.43-5.32 \$/ha*year)
401 reached values closer than European natural parks (Molina et al., 2017b). One Virginia
402 study (Morton et al., 2003) showed a middle annual value per unit area (7 \$/ha) in
403 relation to "Conguillio National Park" and "Tolhuaca National Park" and
404 "Malalcahuello National Reserve".

405 There are other studies which obtain the total recreational losses based on the
406 integration of landscape value and vegetation resilience or the time needed by a
407 landscape to recuperate its original scenic beauty and recreational value because of fire

408 (Butry et al., 2000; Barrio et al., 2007). When comparing our results with these former
409 studies, we observed an infra-valuation of the monetary values. Knowing the annual
410 landscape value, fire behavior and vegetation resilience, landscape vulnerability could
411 be represented by updating the economic value over the years necessary for restoring
412 the original landscape quality (Molina et al., 2017b). Future studies should contemplate
413 the vegetation resilience of the different landscapes on the study areas. As an example,
414 if we consider vegetation resilience between 13 to 44 years for *Nothofagus* forests
415 (Molina et al., 2017a), we would estimate a similar value per hectare than those studies
416 (Butry et al., 2000; Barrio et al., 2007).

417 Despite the limitations of CVM and social preferences method (González and León,
418 2003; MacMillan et al., 2006; Grammatikopoulou and Olsen, 2013; Chen and Hua,
419 2015; Chatterjee et al., 2017), they have become important tools to economic valuation
420 of environmental services and landscape goods. Bias resulting from the insufficient
421 detail of bad-informed people could be resolved by the study design and
422 implementation. Former CVM studies have typically found that WTP is less sensitive to
423 the stated magnitude of fire risk than standard economic theory would predicted. A
424 plausible explanation for this inadequate sensitivity is that respondents may not
425 understand the magnitude of the described fire risk (Molina et al., 2017b). Therefore,
426 important differences in the effect of alternative visual aids could be found based on the
427 photographs used. For a subsample of respondents that received no visual aids or other
428 photographs, it could not be performed statistically significant difference. Replication of
429 these results in a context other than fire-prone landscapes is needed before these results
430 can be generalized. Consequently, the effect of visual aids on sensitivity to magnitude
431 of fire risk may be quite different than considered in Chile and Spain studies.

432 Forest managers require information on the socioeconomic consequences of landscape
433 alteration. Considering this, landscape susceptibility provides a tool to improve fuel
434 treatment optimization and budget allocation in order to ensure the cost-efficient of
435 management activities. Landscape susceptibility approach would help pointing out the
436 situations where fuel management may be useful in reducing fire impacts. The
437 landscape model provided here permits the extrapolation of this landscape susceptibility
438 approach to any territory and scale using social questionnaires in the natural protected
439 areas. Further studies are required to identify a proportional allocation of the economic
440 annual value according to landscape quality and the location of recreational activities
441 based on contingent rating. Experiences of large wildfires (2017) and potential fire
442 impacts associated with Chilean protected areas should lead to fire management
443 decisions by prioritizing more valuable and susceptible areas.

444

445 **Conclusions**

446 Landscape goods could reactive the economy of rural wilderness areas, mainly in
447 natural protected areas. In this sense, results reflect the relevance of landscape and
448 leisure activities provided by Chilean protected areas. For decision-making, the
449 economic valuation of landscape and recreation resources is useful and important,
450 because it provides managers information necessary to evaluate potential tradeoffs when
451 proposing fire reduction programs for protected areas conservation. The economy
452 relevance of landscape goods would justify greater investments in fire prevention
453 programs.

454 A model of evaluating the landscape susceptibility using social stated preferences and
455 potential fire impacts is of great importance for the comprehensive management of the

456 territory. The proposed methodology can be extrapolated to other regions and countries,
457 although contingent valuation is required for the inclusion of landscape value in the
458 economic assessment. Expressing the landscape susceptibility in terms of the
459 deterioration rate or visit frequency decrease responds to a needed simplicity required
460 by the questionnaire respondents. The potential impacts associated with fire occurrence
461 in natural protected areas should lead to fire prevention treatments such as fuel
462 reduction and prescribed fire programs to mitigate potential fire impacts. The reduction
463 of fire vulnerability under different management alternatives is a keystone to sustainable
464 landscape and forest planning.

465

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468 vulnerability of the recreational resources against forest fires in the 21st century" of the
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471 anonymous reviewers and the Associate Editor for their help in improving presentation
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607 **Figure captions**

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609 Figure 1. Study area location

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611 Figure 2. Depreciation rate in the number of arrivals according to each protected area
612 and fire intensity (FIL)

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614 Figure 3. Annual landscape susceptibility (\$/ha) according to each protected area and
615 fire intensity (FIL)

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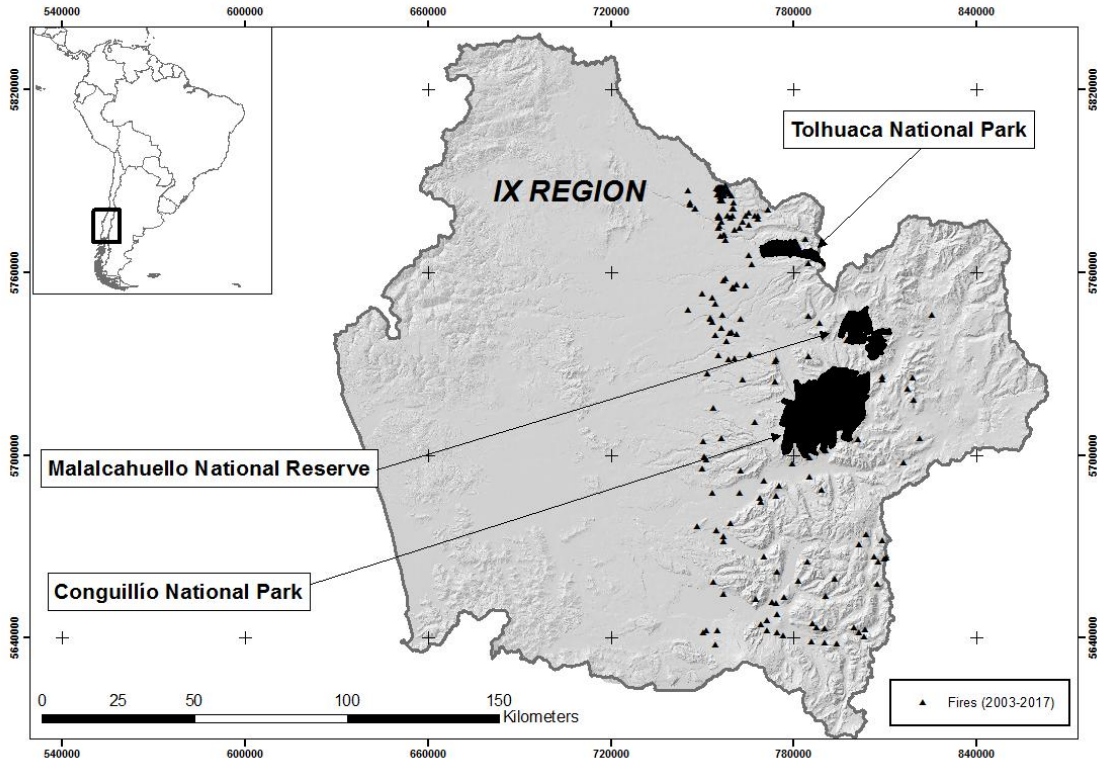
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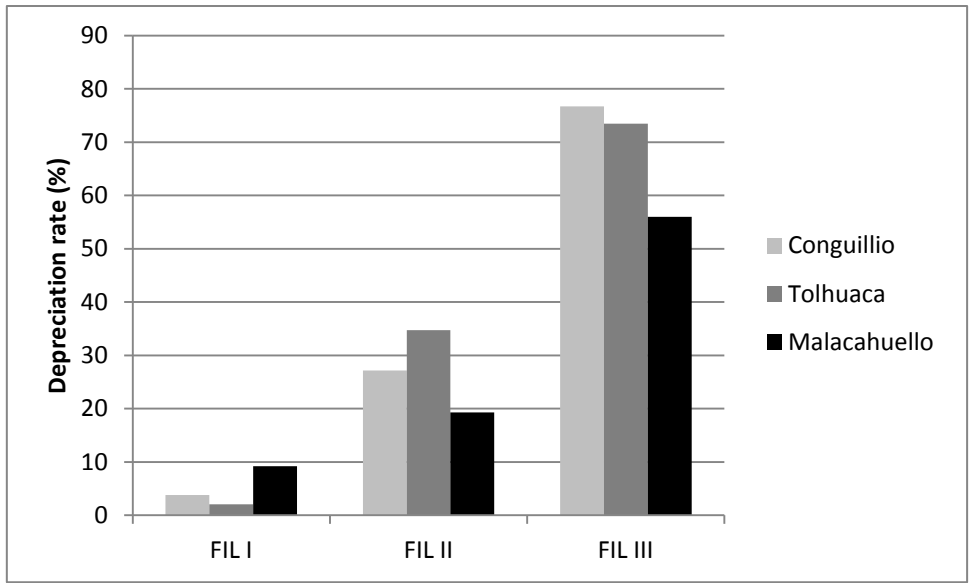
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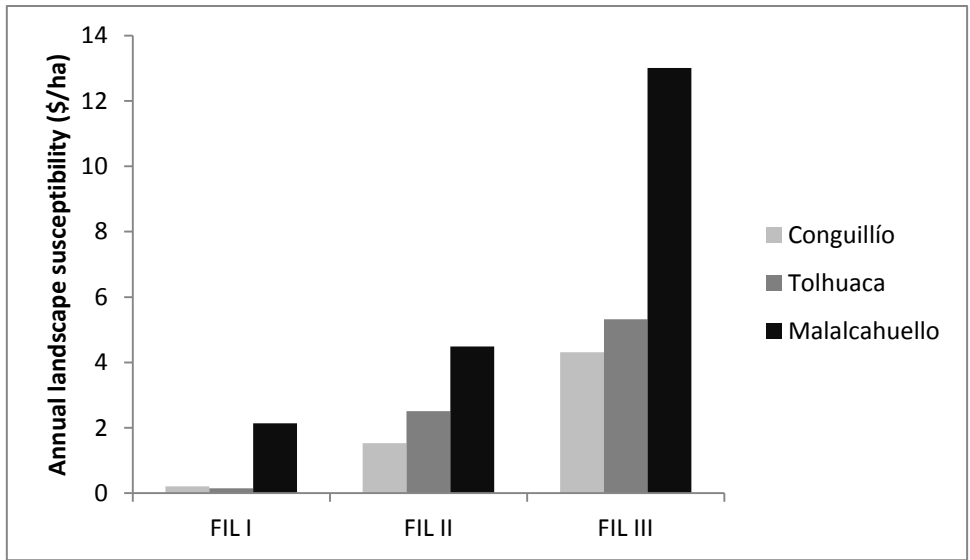
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682 Table 1. Logit model results for all respondents (positive WTP, legitimate zero and
 683 protest respondents) and selected model (positive WTP and protest zero respondents)

	All respondents		Positive WTP and protest zero respondents	
	Coef.	Std. error	Coef.	Std. error
Constant	1.88	0.34	17.84	1.14
Income	-0.43	0.39	-21.07 ^a	0.01
Education	0.43 ^b	0.24	1.08	1.06
Age	-0.11	0.31	0.40	0.59
Gender	0.09	0.22	-0.84	1.06
Place	0.35 ^b	0.24	0.89 ^b	0.48
Chi-square	15.89		36.16	
Pseudo-R2	0.09		0.34	

684 Income: 1 = less than \$1,000, 0 = otherwise

685 Education: 1 = university degree o higher, 0 = otherwise

686 Age: 1 = age is less than 40 years, 0 = ages is more than 40 years

687 Gender: 1 = female, 0= male

688 Place: 1 = IX region of Chile, 0 = otherwise

689 a 5% significance level

690 b 10% significance level

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699 Table 2. WTP differences (\$) between the least and the most favorable CVM scenarios

Area	WTP "all respondents"	WTP "affirmative respondents"
Conguillio	2.2(±3) ^a	3.1(±3.2) ^a
Tolhuaca	3.6(±5) ^b	4.2(±5.4) ^b
Malacahuello	2.3(±2.7) ^a	2.7(±2.2) ^a

700 Mean values in a column followed by the same letters are not significant different ($p < 0.05$,
 701 ANOVA Tukey HSD)

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723 Table 3. WTP differences (\$) based on gender distinction and protected area

Scenario*	Gender distinction		
	Conguillio	Tolhuaca	Malalcahuello
WTP F1	2(±2.1) ^a	4.8(±9.5) ^b	2.1(±1.9) ^a
WTP F2	2.8(±2.1) ^a	5.9(±10.3) ^b	2.5(±1.9) ^a
WTP M1	2.3(±3.8) ^a	2.6(±2) ^a	2.5(±2.6) ^a
WTP M2	3.4(±4.2) ^a	2.9(±1.8) ^a	3(±2.6) ^a

724 Mean values in a row followed by the same letters are not significant different ($p < 0.05$,
 725 ANOVA Tukey HSD)

726 *WTP F1: Willingness to Pay of all female respondents; WTP F2: Willingness to Pay of
 727 affirmative female respondents; WTP M1: Willingness to Pay of all male respondents; WTP
 728 M2: Willingness to Pay of affirmative male respondents

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747 Table 4. WTP differences (\$) based on age and protected area

Scenario*	Age (years)		
	Conguillio	Tolhuaca	Malalcahuello
< 40 all	2.1(±2.5) ^a	4.3(±8.7) ^b	2.4(±2.3) ^a
< 40 selected	2.8(±2.5) ^a	4.8(±9.1) ^b	2.8(±2.3) ^a
> 40 all	2.3(±3.7) ^a	2.6(±2.5) ^a	2.3(±2.5) ^a
> 40 selected	3.3(±4.1) ^a	3.1(±2.4) ^a	2.7(±2.5) ^a

748 Mean values in a row followed by the same letters are not significant different (p < 0.05,
 749 ANOVA Tukey HSD)

750 *All: Willingness to Pay of all respondents; Selected: Willingness to Pay of affirmative
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770 Table 5. WTP differences (\$) based on job or economic condition and protected area

Scenario*	Job (economic condition)		
	Conguillio	Tolhuaca	Malalcahuello
S and U all	1.9(±2.3) ^a	3.8(±2.8) ^b	2(±1.9) ^a
S and U selection	2.6(±2.3) ^a	4.5(±2.5) ^b	2.4(±1.9) ^a
W all	2.1(±3.4) ^a	2.3(±2) ^a	2.3(±2.7) ^a
W selection	3.1(±3.8) ^a	2.9(±1.7) ^a	2.7(±2.7) ^a
HLW all	2.5(±2.8) ^a	5.2(±10.7) ^b	2.4(±2.1) ^a
HLW selection	3.3(±2.8) ^a	5.6(±11) ^b	2.9(±2) ^a

771 Mean values in a row followed by the same letters are not significant different (p < 0.05,
 772 ANOVA Tukey HSD)

773 *S and U all: Willingness to Pay of all student and unemployed respondents; S and U
 774 selection: Willingness to Pay of affirmative student and unemployed respondents; W all:
 775 Willingness to Pay of all conventional workers; W selection: Willingness to Pay of affirmative
 776 conventional workers; HLW all: Willingness to Pay of all high level respondents based on its
 777 medium salary (engineer, doctor, pilot, architect, banker, politician, manager, vet, lawyer and
 778 broker); HLW selection: Willingness to Pay of affirmative high level respondents

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794 Table 6. WTP differences (\$) based on place of origin and protected area

Scenario*	Place of origin		
	Conguillio	Tolhuaca	Malalcahuello
Araucania all	1.8(±2.2) ^a	2.2(±2.3) ^a	2.4(±2.8) ^a
Araucania selection	2.6(±2.2) ^a	2.9(±2.2) ^a	2.6(±2.8) ^a
Bio-Bio all	3.1(±4.9) ^a	1.9(±1) ^b	2.2(±2.2) ^b
Bio-Bio selection	3.8(±5.2) ^a	2.3(±0.8) ^b	2.6(±2.2) ^b
Metropolitano all	2.5(±2.9) ^a	5.3(±9.9) ^b	2.4(±1.8) ^a
Metropolitano selection	3.5(±2.8) ^a	5.9(±10.3) ^b	2.8(±1.6) ^a
Others all	1(±2) ^a	2.3(±0.5) ^b	2.8(±3) ^b
Others selection	1.1(±0.6) ^a	3.1(±0.8) ^b	3.6(±3) ^b
Foreign all	1.8(±1) ^a	3.1(±0.9) ^a	4.1(±3.6) ^b
Foreign selection	2.1(±0.7) ^a	3.1(±0.9) ^a	6.1(±1.1) ^b

795 Mean values in a row followed by the same letters are not significant different (p < 0.05,
 796 ANOVA Tukey HSD)

797 *Araucaria all: Willingness to Pay of all Araucaria Region respondents; Araucaria selection:
 798 Willingness to Pay of affirmative Araucaria Region respondents; Bio-Bio all: Willingness to
 799 Pay of all Bio-Bio Region respondents; Bio-Bio selection: Willingness to Pay of affirmative
 800 Bio-Bio Region respondents; Metropolitano all: Willingness to Pay of all Metropolitano
 801 Region respondents; Metropolitano selection: Willingness to Pay of affirmative
 802 Metropolitano Region respondents; Others all: Willingness to Pay of all other Regions
 803 respondents; Others selection: Willingness to Pay of affirmative other Regions respondents;
 804 Foreign all: Willingness to Pay of all foreign respondents; Foreign selection: Willingness to
 805 Pay of affirmative foreign respondents

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817 Table 7. Annual valuation of recreational resources in the protected areas based on the
 818 CVM scenario

Area	WTP (\$/visitor*year)*	Visitors	Annual economic value (\$)*
Conguillio	2.2-3.1	111,709	241,004-341,953
Tolhuaca	3.6-4.2	11,270	40,189-46,887
Malacahuello	2.3-2.7	108,618	251,036-297,394

819 * The first value is in relation to "all respondents" scenario and the second value is based on
 820 "affirmative respondents" scenario

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840 Table 8. Depreciation rate (%) according to respondent characteristics

Respondents	FIL I	FIL II	FIL III
CVM scenario			
All respondents	9.78/9.78 ^a	28.26/35.87 ^a	70.65/70.65 ^a
Affirmative respondents	4.37/4.37 ^b	19.06/25 ^b	75.94/77.19 ^b
Gender distinction			
Female	5.34/5.34 ^a	23.3/27.67 ^a	78.15/79.6 ^a
Male	5.85/5.85 ^a	19.02/27.31 ^a	71.22/71.71 ^a
Age (years)			
< 40 years	5.49/5.49 ^a	19.61/25.1 ^a	72.55/74.12 ^a
> 40 years	5.77/5.77 ^a	23.72/31.41 ^a	78.2/78.2 ^a
Job (economic condition)			
Student and unemployed	4.94/4.94 ^a	16.05/22.22 ^a	65.43/67.9 ^a
Conventional workers	5.58/5.58 ^b	21.23/29.05 ^b	78.21/79.33 ^b
High level workers	6/6 ^b	24/28.67 ^b	75.33/75.33 ^b
Place of origin			
Araucania	4.9/4.9 ^a	21.57/27.45 ^a	77.45/77.45 ^a
Bio-Bio	7.75/7.75 ^b	20.15/28.68 ^a	80.62/82.17 ^a
Metropolitana	2.58/2.58 ^c	24.14/29.31 ^b	71.55/73.27 ^a
Others Regions	7.14/7.14 ^b	19.05/19.05 ^c	64.28/64.28 ^b
Foreign	9.52/9.52 ^d	14.28/28.56 ^d	61.91/61.91 ^b

841 Mean values for each respondent characteristic in a column followed by the same letters are not
 842 significant different ($p < 0.05$, Wilcoxon test)

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855 Table 9. Annual landscape susceptibility according to protected area and FIL

Area	FIL I (\$)	FIL II (\$)	FIL III (\$)
Conguillio	9,182-13,028	44,754-92,806	184,778-262,175
Tolhuaca	820-956	10,590-16,270	28,707-34,448
Malacahuello	23,120-27,390	41,848-57,397	133,877-166,422

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879 **Appendix I (Questionnaire photographs)**



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