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Predicting the impact of invasive trees from different measures of abundance --Manuscript Draft--

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Abstract:	<p>Biological invasions produce negative impacts worldwide, causing massive economic costs and ecological impacts. Knowing the relationship between invasive species abundance and the magnitude of their impacts (abundance-impact curves) is critical to designing prevention and management strategies that effectively tackle these impacts. However, different measures of abundance may produce different abundance-impact curves. Woody plants are among the most transformative invaders, especially in grassland ecosystems because of the introduction of hitherto absent life forms. In this study, our first goal was to assess the impact of a woody invader, <i>Pinus contorta</i> (hereafter pine), on native grassland productivity and livestock grazing in Patagonia (Argentina), building abundance-impact curves. Our second goal, was to compare different measure of pine abundance (density, basal area and canopy cover) as predictors of pine's impact on grassland productivity. Our third goal, was to compare abundance-impact curves among the mentioned measures of pine abundance and among different measures of impact: total grassland productivity, palatable productivity and sheep stocking rate (the number of sheep that the grassland can sustainably support). Pine canopy cover, closely followed by basal area, was the measure of abundance that best explained the impact on grassland productivity, but the shape of abundance impact curves differed between measures of abundance. While increases in pine density and basal area always reduced grassland productivity, pine canopy cover below 30% slightly increased grassland productivity and higher values caused an exponential decline. This increase in grassland productivity with low levels of pine canopy cover could be explained by the amelioration of stressful abiotic conditions for grassland species. Different measures of impact, namely total productivity, palatable productivity and sheep stocking rate, drew very similar results. Our abundance-impact curves are key to guide the management of invasive pines because a proper assessment of how many invasive individuals (per surface unit) are unacceptable, according to environmental or economic impact thresholds, is fundamental to define when to start management actions.</p>
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25 Highlights

26 • Invasive species abundance-impact curves are key to designing management strategies.

27 • We assessed how increasing abundance of invasive pines impacted grasslands.

28 • Pine canopy cover over 30% reduced grasslands productivity exponentially.

29 • Managers should start controlling invasive pines before this threshold is reached.

30

31 Abstract

32 Biological invasions produce negative impacts worldwide, causing massive economic costs and
33 ecological impacts. Knowing the relationship between invasive species abundance and the
34 magnitude of their impacts (abundance-impact curves) is critical to designing prevention and
35 management strategies that effectively tackle these impacts. However, different measures of
36 abundance may produce different abundance-impact curves. Woody plants are among the most
37 transformative invaders, especially in grassland ecosystems because of the introduction of
38 hitherto absent life forms. In this study, our first goal was to assess the impact of a woody invader,
39 *Pinus contorta* (hereafter pine), on native grassland productivity and livestock grazing in Patagonia
40 (Argentina), building abundance-impact curves. Our second goal, was to compare different
41 measure of pine abundance (density, basal area and canopy cover) as predictors of pine's impact
42 on grassland productivity. Our third goal, was to compare abundance-impact curves among the
43 mentioned measures of pine abundance and among different measures of impact: total grassland

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44 productivity, palatable productivity and sheep stocking rate (the number of sheep that the
45 grassland can sustainably support). Pine canopy cover, closely followed by basal area, was the
46 measure of abundance that best explained the impact on grassland productivity, but the shape of
47 abundance impact curves differed between measures of abundance. While increases in pine
48 density and basal area always reduced grassland productivity, pine canopy cover below 30%
49 slightly increased grassland productivity and higher values caused an exponential decline. This
50 increase in grassland productivity with low levels of pine canopy cover could be explained by the
51 amelioration of stressful abiotic conditions for grassland species. Different measures of impact,
52 namely total productivity, palatable productivity and sheep stocking rate, drew very similar
53 results. Our abundance-impact curves are key to guide the management of invasive pines because
54 a proper assessment of how many invasive individuals (per surface unit) are unacceptable,
55 according to environmental or economic impact thresholds, is fundamental to define when to
56 start management actions.

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58 Keywords: Impact-based management, grasslands, livestock grazing, Pinus, primary productivity,
59 woody invasions.

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61 1. Introduction

62 Biological invasions are one of the major components of global change (Díaz et al., 2019),
63 producing negative impacts globally (Cameron et al., 2016; Gallardo et al., 2016; Simberloff, 2011;
64 Vilà et al., 2011) and incurring high economic costs (Diagne et al., 2021; Pimentel et al., 2005). All
65 continents are affected by biological invasions (van Kleunen et al., 2015), even the remote
66 Antarctica (Frenot et al., 2005; Hughes et al., 2015), as are all types of ecosystems (Vilà and Hulme,

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67 2017), spanning tropical forests (Acurio et al., 2010) to deserts (Tellman, 2002). In addition to their
68 severe ecological impacts, the negative impacts produced by these invasive species threaten the
69 provision of essential ecosystem services and the economic activities that depend upon them (Vilà
70 and Hulme, 2017). Knowing the relationship between invasive species abundance and the
71 magnitude of their impacts (abundance-impact curves hereafter) is essential to forecasting and
72 effectively managing damages (Cassey et al., 2006; Gilbert and Levine, 2013; Norbury et al., 2015;
73 Thomsen et al., 2011). These relationships can inform economic analyses (costs vs. benefits) of
74 management actions (Sofaer et al., 2018; Yokomizo et al., 2009) that are instrumental for
75 designing management strategies that effectively mitigate these impacts (Bradley et al., 2019;
76 Byers et al., 2002).

77 Woody plants are among the most transformative plant invaders, especially in grassland
78 ecosystems because of the changes produced by the introduction of a new life form (Ehrenfeld,
79 2003; Rundel et al., 2014). This is a problem global in scope and growing, consequently requiring
80 immediate action (Archer et al., 2017; Simberloff et al., 2010; Van Auken, 2000). While many
81 impacts have been documented for invasive woody plants on natural grasslands (Davis et al.,
82 2019; Ferraina et al., 2021; Pawson et al., 2010), no study has yet quantified the relationship
83 between their abundance and the magnitude of their impacts on native grasslands productivity, or
84 how these impacts progressively affect economic activities which depend on this ecosystem
85 service, such as livestock grazing. For instance, a recent study has measured the impact of an
86 invasive woody plant (*Gleditsia triacanthos*) on the productivity of native grasslands, but only
87 comparing invaded vs uninvaded plots (i.e. with no gradient in invader abundance) (Ferraina et al.,
88 2021). Another study, assessed the impact of increasing *P. contorta* cover on native grasslands
89 species richness and relative cover (Taylor et al., 2016), which (although related) are different
90 from grassland productivity (Catchpole and Wheeler, 1992; Chiarucci et al., 1999; Jiang et al.,

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91 2017). Further, none of the mentioned studies focused on the invader impact on the productivity
92 of palatable grassland species (those foraged by livestock), which in turn would impact livestock
93 grazing. Management of invasive woody plants to tackle their impacts on grasslands productivity
94 and livestock grazing will benefit from detailed quantifications of their abundance-impact curves
95 (Bradley et al., 2019; Yokomizo et al., 2009).

96 The *Pinus* genus is an ideal study system to assess the abundance-impact relationship that
97 drives the impact of woody invasives on grassland productivity. Firstly, numerous species of this
98 genus have become invasive in grassland ecosystems around the world (Nuñez et al., 2017;
99 Richardson et al., 1994; Simberloff et al., 2010). Further, as most pine invasions start from
100 plantations maintained for economical purposes (Nuñez et al., 2017), their spread away from
101 these plantations, through wind dispersal, creates abundance gradients with the highest
102 abundance near the source of propagules and lower abundance with increasing distance (Langdon
103 et al., 2010; Taylor et al., 2016). These abundance gradients offer an exceptional opportunity to
104 measure *Pinus* invasion impacts at varying levels of invader abundance. Finally, the invasion of
105 *Pinus* species has notorious impacts in grasslands, and these impacts may affect grassland
106 productivity, namely a strong reduction in plant species richness and cover (Davis et al., 2019;
107 Franzese et al., 2017; Taylor et al., 2016), a steep increase in fuel load which affects the ecosystem
108 fire regime (Paritsis et al., 2018; Taylor et al., 2017) and an increase in litter depth (Taylor et al.,
109 2016) which may affect nutrients cycles (Araujo and Austin, 2015).

110 Abundance of invasive *Pinus* species can be measured in different ways, with a trade-off
111 between simplicity and the detail in the information obtained. On one hand, density is the most
112 easily obtained but it does not account for the difference in size between seedlings and adults,
113 which is key when assessing the impact of invasive woody plants (Franzese et al., 2017). Basal area
114 (the sum of the cross sections of all trees within a hectare) incorporates pine size through the

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115 additional measurement of individual diameters through a simple procedure that requires no
116 specialized or expensive equipment. On the other hand, canopy cover is more resource and time-
117 consuming, obtained through a more technical procedure which includes taking hemispheric
118 pictures and analyzing the obtained images. This requires relatively expensive (photographic
119 camera) and specialized (fisheye lens) equipment.

120 In this study, our first objective was to quantify the impact of *Pinus contorta* (hereafter
121 pine) invasion on native grassland productivity in northwest Patagonia. For this purpose, we built
122 abundance-impact curves, assessing the impact of pine invasion on grassland productivity along a
123 gradient of increasing pine abundance. Our second objective was to compare different measures
124 of pine abundance (density, basal area and canopy cover) as predictors of pine's impact on
125 grassland productivity. While pine density is quick and easy to measure, we expected that it would
126 underestimate the individual impact of adults (which are considered equivalent to seedlings). We
127 expected that basal area would be a better predictor of pine impact, since it incorporates
128 individual size, while still being fairly straightforward to measure. Canopy cover moves a step
129 further, since it integrates the distribution of pine sizes, according to the distance of each
130 individual to the point of measurement, by recording the sky obstruction by pine canopies (Rich,
131 1990). Therefore, we expected that canopy cover would provide the best prediction of pine
132 impacts. Last but not least, our third objective was to compare abundance-impact curves among
133 different measures of pine abundance (previously mentioned) and among measures of impact,
134 including total grassland productivity, productivity of palatable species (i.e. those that are foraged
135 by sheep, the main livestock in the region (DNAP, 2018a, 2018b, 2018c)), and sheep stocking rate
136 (i.e. the number of sheep that can be supported by this palatable productivity). Since these are
137 semiarid grasslands, we expected that low pine abundance could facilitate grassland biomass
138 growth (Belsky, 1994; Blaser et al., 2013; Dohn et al., 2013; Mazía et al., 2016) by ameliorating

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139 stressful abiotic conditions through the reduction of air temperature and soil water evaporation
140 (Belsky et al., 1993) and via the pumping of water (by hydraulic lift) from deeper soil (Neumann
141 and Cardon, 2012) (Figure S1). We also expected that low pine abundance would favor the
142 proportion of palatable species in the grasslands (Bernardi et al., 2016; Peterson et al., 2007;
143 Scholes and Archer, 1997), possibly by providing microsite conditions more favorable for these
144 species, concomitantly increasing palatable productivity, even more than total productivity via
145 positive effects on total productivity and the proportion of palatable productivity. Additionally, we
146 also expected that high pine abundance would reduce both total and palatable productivity (Davis
147 et al., 2019; Franzese et al., 2017; Taylor et al., 2016), mainly through the reduction in solar
148 radiation (Rago et al., 2021).

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150 2. Materials and Methods

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152 2.1 Study area

153 We conducted our research on five sites (called A1, A2, A3, C1 and C2) in northwest Patagonia
154 (Argentina; Figure 1a & 1b, Table 1) where plantations of *Pinus contorta* were established
155 approximately 30 years ago, replacing areas of native grasslands. Each of these five plantations has
156 invaded adjacent native grasslands producing a gradient of invasion abundance, with the highest
157 abundance near the plantations and decreasing with distance. Mean annual precipitation is 900
158 mm and mean annual temperature is 8.3 °C (Bariloche Airport meteorological station, located
159 within our study area). Vegetation cover in these native grasslands is on average 60%, and
160 dominated by *Pappostipa speciosa* and *Festuca pallescens*, both perennial tussock grasses and
161 palatable species to sheep, supplemented with other plant species present in lower abundances

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162 (Anchorena and Cingolani, 2002; Bonvissuto et al., 2008). These grasslands cover a large
163 proportion of the Patagonian region and are among the most susceptible ecosystems to pine
164 invasions (Richardson et al., 1994; Simberloff et al., 2010). Pine plantations have replaced large
165 areas of these grasslands (Schlichter and Laclau, 1998), becoming a source for invasions that, in
166 turn, further replace areas of these native grasslands (Nuñez et al., 2017). The most important
167 economic activity in these grasslands is extensive livestock grazing, mainly by sheep (DNAP, 2018b,
168 2018a, 2018c), which depends almost exclusively on native rangelands for forage (Golluscio et al.,
169 1998; Soriano and Paruelo, 1990).

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171 2.2 Experimental design

172 2.2.1 Assessing pine impacts on grassland productivity using abundance-impact curves

173 To quantify the impact of invasive pines on grassland productivity, we estimated grassland
174 productivity along transects of increasing pine abundance, in the five sites described before (A1,
175 A2, A3, C1 and C2). We defined grassland productivity as the dry aerial biomass produced annually
176 per surface unit (kg /hectare/year). We used green biomass in the peak production as a metric for
177 aboveground annual productivity, as proposed by Sala and Austin (2000) in sites where the
178 growing season is brief and marked. In the region where our sites are located, the growing season
179 spans from December to March (Heinemann and Kitzberger, 2006). In each of our five sites, we
180 built transects (five transects per site) radiating from the edge of a pine plantation, following the
181 main dispersal direction, until we reached a position beyond the invasion front, where invasive
182 individuals were absent (Figure 1c and 1d). On each transect, we compared grassland productivity
183 in the absence of pine invasion (control plot) with grassland productivity in paired plots with
184 increasing levels of pine abundance (invaded plots). Since the distance between the plantation and

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185 invasion front varied across sites, we adjusted the length of our transects accordingly. Therefore,
186 in sites A2 and C2 our transects were 200-m long, in site C1 transects were 350-m long, and in sites
187 A1 and A3 transects were 500 m long. The distance between the five transects within the same
188 site was 50 m. Along each independent transect, we set circular plots at the following distances
189 from the pine plantation (propagule source): 0, 25, 50, 75, 100, 150, 200, 350 and 500 m (these
190 two last distances only for sites with transects that reached 350 and 500 m) based on visible
191 thresholds in pine density. Distance between plots was smaller closer to the pine plantation than
192 near the invasion front because the greatest variation in invasion density occurs near the seed
193 source and we wanted to capture this variability.

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195 2.2.2 Comparing the performance of different measures of pine abundance as predictors of their
196 impact on grassland productivity

197 We measured three different metrics of pine abundance, to find out which one best predicted
198 their impact in grassland productivity:

199 a) Density (trees/ha): number of pines per hectare. Within each circular plot (10 m diameter), we
200 counted the number of pine individuals and calculated the corresponding number of individuals
201 for a hectare.

202 b) Basal area (m²/ha): the sum of the cross sections of all pines per hectare. Within each plot, we
203 recorded each individual diameter at ground level to calculate every cross section and add them
204 up to obtain basal area (m²/ha).

205 c) Canopy cover (%): the percentage of the sky that is blocked by the projection of the pine
206 crowns. Within each circular plot, we randomly distributed six square subplots (0.25 m²). In each

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207 of these subplots we took a hemispheric picture to estimate pine canopy cover using a Nikon
208 Coolpix P80 camera with an Opteka fisheye lens 0.20x.

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210 2.2.3 Comparing abundance-impact curves across metrics of pine abundance and among pine
211 impacts

212 We compared abundance-impact curves across the metrics of pine abundance described
213 above and across three measures of pine impacts on grasslands:

214 1) Total productivity: aerial green biomass per hectare, considering all grassland species. In the
215 same subplots where we measured pine canopy cover (by taking hemispheric pictures), we
216 estimated total grassland productivity by harvesting (using pruning scissors) all the green biomass
217 of grasses, forbs and shrubs at the peak of the growing season (early summer) (Sala and Austin,
218 2000). We classified each species into palatable and non-palatable (foraged and not foraged by
219 sheep, respectively) according to a handbook for natural grassland condition in Patagonia
220 (Bonvissuto et al., 2008). We oven-dried the harvested biomass at 60°C for 48h and used a
221 precision scale (0.001 grams) to measure the dry weight.

222 2) Palatable productivity: aerial green biomass per hectare, only considering species foraged by
223 sheep. We followed the procedure described to estimate total productivity but here we only
224 considered palatable grassland species, including grasses, forbs and shrubs.

225 3) Sheep stocking rate: number of sheep that can be sustained by the palatable productivity. To
226 assess the impact of pine invasions on livestock grazing, mainly sheep, which relies almost
227 exclusively on rangelands for forage, we calculated the sheep stocking rate that the grasslands of
228 our study can sustainably support based on the feeding requirement of an Ovine Livestock Unit

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229 (OLU). The OLU represents an individual Merino wether (castrated male sheep) with an average
230 live weight of 40 kg that consumes 365 kg of dry forage in a year in Patagonia grasslands. We used
231 this livestock equivalence as a reference that establishes a correspondence between different
232 animal species. We also considered the vegetation Use Factor (UF) of each site, a value
233 corresponding to a proportion of effectively consumable forage that can be grazed by sheep
234 without compromising a sustainable forage production over time (Table 1) (Bonvissuto et al.,
235 2008; Golluscio et al., 2009).

236 Because pines show rare events of long distance dispersal (Langdon et al., 2010), and even
237 though our transects went beyond the invasion front (which is driven by the predominant short
238 distance dispersal kernel), in some specific cases the plot that was most distant form the
239 plantations ended up located in an area with small groups of invasive pine individuals. We did not
240 consider for analyses the transects where pines were present in all plots along the transect
241 because these transects lacked a control plot, making it impossible to assess their impact. The
242 total number of transects (throughout the five sites) not considered for analyses was five: three in
243 site A2, one in site A3 and one in site C2.

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245 2.3 Statistical analyses

246 To evaluate the impact of increasing pine abundance on grassland productivity we used
247 linear regression models (lm function in R) (Chambers and Hastie, 1992; Wilkinson and Rogers,
248 1973), with pine abundance as predictor variable and their impacts as response variables
249 (abundance-impact curves). To compare different measures of pine abundance as predictors of
250 their impact in grassland productivity we used the three different measures of pine abundance
251 previously described (density, basal area and canopy cover) as predictor variables in our regression

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252 models. We only used one predictive variable for each regression model. We compared regression
253 models with different prediction variables based on the Akaike Information Criterion (AIC).

254 To compare abundance-impact curves among different metrics of pine abundance, we
255 used these regression models where the predictive variable was either pine density, pine basal
256 area or pine canopy cover. To compare abundance-impact curves among measures of impact we
257 used the three measures of impact described before (total grassland productivity, productivity of
258 palatable species, and sheep stocking rate) as response variables in our regression models (one
259 response variable for each regression model). In all cases, we built both a linear and a polynomial
260 (order 2) regression, since both linear and nonlinear relationships between abundance and impact
261 of invasive species have been documented (Bradley et al., 2019; Sofaer et al., 2018). For each
262 predictive variable, we selected between the linear and nonlinear regressions based on AIC.

263 Since grassland productivity is variable across sites (Figure S2), we standardized our impact
264 measures by calculating effect sizes. For each subplot we calculated three different effect sizes
265 (ES), focusing on the three different impacts on grasslands mentioned before (Figure 2), using log-
266 response ratios:

267 1) The effect size of pine invasion on grassland total productivity (EStp):

268
$$EStp = \ln (Bi/Bc).$$

269 2) The effect size of pine invasion on grassland palatable productivity (ESpp):

270
$$ESpp = \ln (Pi/Pc)$$

271 3) The effect size of pine invasion on the sheep stocking rate that the grassland can
272 sustainably support (ESsr):

273
$$ESsr = \ln (Si/Sc)$$

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274 where B_i is the total grassland dry aerial green biomass (including grasses, forbs and shrubs, both
275 palatable and non-palatable) for subplots located in plots invaded by pines, P_i is the total dry aerial
276 green biomass of palatable species for the same subplots, and S_i is the sheep stocking rate that
277 can be sustainably supported by the total dry aerial green biomass of palatable species for the
278 same subplots. B_c is the total grassland dry aerial green biomass for subplots located in the paired
279 control plot of the same transect, which is the most distant from the pine plantation and has not
280 been invaded yet (Figure 1d), P_c is the total dry aerial green biomass of palatable species for
281 subplots located in the control plot, and S_c is the sheep stocking rate that can be sustainably
282 supported by the total dry aerial green biomass of palatable species for subplots located in the
283 control plot.

284 For each plot, we calculated an average for each of the three effect sizes (EStp, ESpp and
285 ESsr) using the data from the six subplots. A positive effect size would indicate that pine invasion
286 increased grassland productivity, while a negative effect size would indicate that grassland
287 productivity is negatively affected by the invasion of pines. The absolute value of the effect size
288 indicates the magnitude of the response of grassland productivity to the invasion of pines, either
289 positive or negative.

290 To understand how different measures of pine abundance relate to each other we built
291 three regression models. In the first case, we built a generalized linear model, with pine density as
292 predictive variable and pine basal area (rounded up to the nearest integer) as response variable,
293 assuming a Poisson distribution. In the second model, we used pine density as predictive variable
294 and pine canopy cover (converted from percentage to proportion data) as response variable,
295 assuming a Beta distribution (“betareg” function from the “betareg” package) (Cribari-Neto and
296 Zeileis, 2010). For the third model, we considered pine basal area as predictive variable and pine

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297 canopy cover (converted from percentage to proportion data) as response variable, assuming a
298 Beta distribution.

299 Finally, for each subplot we assessed how pine density, basal area and canopy cover
300 influenced the proportion of palatable (out of the total) productivity. For this purpose, we built a
301 regression model assuming a Beta distribution (“betareg” function from the “betareg” package)
302 (Cribari-Neto and Zeileis, 2010), with proportion of palatable productivity as response variable and
303 pine density, basal area, or canopy cover as predictive variables. Each regression included only one
304 pine abundance measure as predictive variable. We performed all statistical analyses using R
305 v.4.1.1 (R Development Core Team, 2021).

306

307 3. Results

308 3.1 Assessing pine impacts on grassland productivity using abundance-impact curves

309 We found a predominantly negative effect of pine invasions on grassland total productivity,
310 palatable productivity and sheep stocking rate, with increasing levels of pine abundance causing
311 further declines on each of the measures of impact studied here (Figure 3). However, this pattern
312 was affected by the measure of pine abundance with which abundance-impact curves were built.

313 On one hand, pine density and pine basal area both produced only declines of grassland total
314 productivity, palatable productivity and sheep stocking rate across the range of pine abundance
315 explored here. On the other hand, low levels of pine canopy cover (below 30%) produced an
316 increase of grassland total productivity, palatable productivity and sheep stocking rate, while
317 higher levels of pine canopy cover caused a strong decline.

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319 3.2 Comparing the performance of different measures of pine abundance as predictors of their
320 impact on grassland productivity

321 We found variation among our measures of pine abundance in their adequacy as predictors of
322 impact on grassland productivity (Table 3). The measure of pine abundance that best explained
323 variability in the impact of pine invasion (across all three impact measures) was canopy cover,
324 followed by basal area, and pine density came last, as we expected.

325 While all measures of pine abundance showed a positive relationship with each other,
326 increases in one measure were not associated with proportional increases in other measures
327 (Figure 4). While pine basal area increased exponentially with density, pine canopy cover at first
328 increased exponentially with density and basal area, to then stabilize at high levels of pine density
329 (approximately 3500 pines/ha) and basal area (approximately 60 m²/ha), without reaching 100 %
330 cover.

331

332 3.3 Comparing abundance-impact curves across metrics of pine abundance and among pine
333 impacts

334 We found different relationships between abundance and impact for different pine
335 abundance measures. When considering total productivity, the impact of pine density and basal
336 area was best predicted by a linear function, while that of pine cover was best predicted by a
337 nonlinear (polynomial order 2) function (Table 2a). In the case of palatable productivity and sheep
338 stocking rate, the impact of pine density and cover were also best captured by a nonlinear
339 function, while that of pine basal area was still best described by a linear relationship (Table 2b &
340 2c).

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341 We found that pine density ($R^2 = 0.46$, $p < 0.0001$, Figure 3a) and basal area ($R^2 = 0.53$, $p <$
342 0.0001 , Figure 3a) showed a strong negative linear relationship with grassland total productivity
343 throughout the range of pine abundance surveyed. On the other hand, we found that the best
344 model including pine cover as predictive variable had a negative linear component and a negative
345 polynomial component, resulting in a nonlinear relationship where pine cover below $\sim 30\%$ slightly
346 increased total productivity and higher values strongly reduced it ($R^2 = 0.53$, $p < 0.0001$, Figure 3a).
347 In the case of palatable productivity, the best model including density as measure of abundance
348 had a negative linear and a positive polynomial component, which resulted in a nonlinear
349 relationship where the most rapid decline in palatable productivity occurred at low pine density
350 ($R^2 = 0.42$, $p < 0.0001$, Figure 3b). On the other hand, pine basal area showed a strong negative
351 linear relationship with palatable productivity ($R^2 = 0.43$, $p < 0.0001$, Figure 3b) and pine canopy
352 cover showed a nonlinear relationship where cover values below $\sim 30\%$ slightly increased it and
353 higher values strongly reduced it ($R^2 = 0.43$, $p < 0.0001$, Figure 3b).

354 When considering the sheep stocking rate that can sustainably be supported by these
355 grasslands, pine density also showed a nonlinear relationship where the most rapid decline
356 occurred at low density ($R^2 = 0.36$, $p < 0.0001$, Figure 3c), while pine basal area still showed a
357 strong negative linear relationship ($R^2 = 0.40$, $p < 0.0001$, Figure 3c). Pine cover showed the same
358 nonlinear relationship where values below $\sim 30\%$ slightly increased stocking rate and higher values
359 strongly reduced it ($R^2 = 0.41$, $p < 0.0001$, Figure 3c). We found no significant effects of pine
360 density ($p = 0.1727$), basal area ($p = 0.4236$) or canopy cover ($p = 0.7536$) on the proportion of
361 palatable (out of the total) productivity.

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363 4. Discussion

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364 We have described the relationship between invasive pine abundance and the response of
365 natural grassland productivity using abundance-impact curves. We found that pine canopy cover,
366 closely followed by basal area, was the abundance measure that best explained the impact of pine
367 invasion on grassland productivity. This is likely indicative of the effects of competition for light, as
368 pine canopy cover quantifies the availability of light in the community (Rago et al., 2021; Trentini
369 et al., 2017). We obtained different shapes (linear and nonlinear) for our abundance-impact curves
370 depending on our different measures of pine abundance, highlighting the importance of defining
371 how to measure invasive species abundance. Remarkably, our analyses revealed that while
372 increases in pine density and basal area produced a negative response of grassland productivity
373 throughout the range of pine abundance assessed, pine canopy cover below 30% produced a slight
374 increase in productivity and higher values caused an exponential decline. Different measures of
375 impact such as total productivity, palatable productivity and sustainable sheep stocking rate
376 yielded very similar results.

377 Different measures of invasive pine abundance showed different relationships with
378 grassland response to invasion (i.e., different abundance-impact curves). While pine density and
379 basal area showed mostly negative impacts on grassland productivity, canopy cover showed
380 positive and negative impacts at low and high pine abundances, respectively. Invasive species
381 show multiple abundance-impact curves (Sofaer et al., 2018; Strayer, 2020; Yokomizo et al., 2009)
382 due to variations in the invaded habitat (Thiele et al., 2010) and the impact measured (Kelemen et
383 al., 2016; Robinson et al., 2005; Strayer, 2020). However, there is a glaring gap in our
384 understanding of how abundance-impact curves for the same invasive species can be affected by
385 the measure of abundance used. This novel result highlights the importance of defining an
386 appropriate measure for invader abundance, an aspect intimately related to the impact studied
387 (Strayer, 2020).

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388 Different abundance-impact curves could be related to different mechanisms
389 underpinning the impact of pine invasions in grasslands. There may be different species
390 interactions behind each measure of abundance. Competition has been proposed as one of the
391 key mechanisms affecting invaded grasslands (Levine et al., 2003; Simberloff, 2011; Strayer et al.,
392 2006; Vilà et al., 2011) and pine invasions reduce grassland species richness and cover (Davis et al.,
393 2019; Franzese et al., 2017; Taylor et al., 2016). However, facilitation may be the main interaction
394 between trees and grasses under stressful abiotic conditions (Belsky, 1994; Blaser et al., 2013;
395 Dohn et al., 2013; Mazía et al., 2016). In this sense, high pine canopy cover is an indicator of
396 competition for light (Araujo and Austin 2015, Taylor et al. 2016, Rago et al. 2021) but low canopy
397 cover may be associated with facilitation by reducing stressful abiotic conditions. On the other
398 hand, density and basal area are indicators of competition for different resources (water,
399 nutrients, light) (Álvarez Taboada et al., 2004; Biging and Dobbertin, 1995; Contreras et al., 2011;
400 Tomé and Burkhart, 1989). Each new tree, and its concurrent increase in basal area, requires
401 resources for growth, which if consumed by trees become unavailable for grassland species (Rago
402 et al., 2021; Trentini et al., 2017).

403 Our results broadly concur with previous research focused on the impact of invasive pines
404 in grasslands. For instance, the strong reduction in grassland productivity we found here parallels
405 the sharp decrease in plant richness and cover caused by pine invasions (Davis et al., 2019;
406 Franzese et al., 2017; Taylor et al., 2016). However, we also found positive impacts on grassland
407 productivity when pine abundance was low, confirming the results from studies reporting the
408 facilitation of grassland productivity by trees through the reduction of abiotic stress (Belsky, 1994;
409 Blaser et al., 2013; Dohn et al., 2013; Mazía et al., 2016). Another possible mechanism by which
410 pines are facilitating grassland species growth is by reducing soil erosion, which can be achieved at
411 levels of plant cover below 30% (Eshghizadeh et al., 2018).

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412 The abundance-impact curves described here may be useful to guide management of
413 invasive pines. In particular, low pine canopy cover (below 30%) seems to slightly promote
414 grassland productivity. Therefore, 30% cover is a good candidate threshold to trigger management
415 actions (e.g. start control interventions) to prevent strong increases in negative impacts (Sofaer et
416 al., 2018; Yokomizo et al., 2009). However, 30% cover is rapidly reached, at relatively low pine
417 densities (such as 1000 pines/ha), which may help explain why grassland productivity shows a
418 strong decrease with increasing pine density even for low values of density. In such cases, early
419 detection and rapid response will be the most effective way to manage the impacts of pine
420 invasions (Bradley et al., 2019).

421 For basal area, we found a linear relationship with impact, as reported by Bradley et al
422 (2019) for cases where the invasive species and the impacted native populations are at the same
423 trophic level, and competition is the prevalent interaction. In such cases, management of invasive
424 populations may be beneficial at any level of basal area (Bradley et al., 2019). However, pine basal
425 area increases exponentially with pine density, so there is a high risk of achieving high levels of
426 basal area before being noticed, especially if abundance is monitored using density. Based on this,
427 a key target for management should be to define a limit of grassland productivity loss or sheep
428 stocking rate reduction beyond which further losses are not affordable, either from an economic
429 (Barney, 2016; Higley and Pedigo, 1996; Sofaer et al., 2018; Yokomizo et al., 2009) or from an
430 ecological point of view (i.e. “breakpoints”, Vilà et al. 2011). This limit could be context-specific,
431 specified according to the impact measure (total grassland productivity, palatable productivity or
432 sheep stocking rate), or according to the economic context and the trade-offs between cost of
433 control and benefits of increases in grassland productivity (i.e. reduction of impacts) (Higley &
434 Predigo 1996, Yokomizo et al. 2009, Soafer et al. 2018). Nevertheless, it is paramount that the pine

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435 abundance measure used to guide management matches those of the abundance-impact curves
436 and is also used to monitor invasive pine abundance.

437 We have inferred possible mechanisms behind our results, but further research will be
438 needed to test these ideas. For example, measuring and comparing abiotic conditions across
439 gradients of pine abundance could help describe the growing conditions of grassland species. Our
440 abundance-impact curves may be applicable to other invasive woody species, especially other
441 members of the *Pinus* genus, and to grasslands in other temperate regions. However, some of our
442 results may be specific to our study species and our study region. Therefore, replications of our
443 study may be needed in new regions and with other invasive species to assess if our abundance-
444 impact curves hold similar.

445

446 5. Conclusions

447 The abundance-impact curves we have built here may be useful to guide the management of
448 invasive woody plants in grasslands with similar abiotic conditions. In temperate regions with
449 lower summer abiotic stress, grassland productivity may be reduced by invasive woody plants
450 throughout the whole range of woody canopy cover. Conversely, in regions with higher summer
451 water deficit invasive woody canopy cover above 30% may still increase grassland productivity.
452 While we show a predominant negative impact of invasive woody plants on native grassland
453 productivity and their capacity to sustain livestock grazing, our abundance-impact curves differ
454 among measures of invader abundance, and the timing of management actions could differ
455 substantially according to the measure of abundance used to monitor the invader. Ideally,
456 managers should consider using multiple measures of invader abundance to monitor invasions
457 since they provide complementary information. However, in general, time and money are limiting

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458 and only one measure of abundance can be frequently assessed. In such cases, we suggest using
459 invader canopy cover to monitor and guide the management of woody plant invasions in
460 grasslands.

461

462 Declaration of competing interest

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

465

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474

475 Data availability statement

476 The data are freely available from Environmental Information Data Centre (EIDC)

477 (<http://eidc.ceh.ac.uk/>) for non-commercial use under Open Government Licence terms and

478 conditions. <https://doi.org/10.5285/54fe47f3-778e-4e0b-8cf5-b2fda2473b7f>.

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727 Tables

728 Table 1: Location (latitude and longitude), mean individual tree diameter (and standard error; SE),
729 surface covered (hectares) for each *Pinus contorta* plantation, as well as mean (and SE) grassland
730 palatable productivity and its corresponding vegetation use factor (UF), for each study site.

Site	Latitude	Longitude	Diameter (cm)	Surface (ha)	Palatable productivity (kg/ha)	Vegetation use factor (%)
A1	41° 9' 48.13" S	71°15' 32.84" O	44.9 (2.4)	3.206	479 (96)	40
A2	41°12' 50.99" S	71°13' 10.64" O	29.9 (1.2)	2.171	281 (145)	30
A3	41°12' 25.95" S	71°13' 59.91" O	36.1 (1.2)	2.778	153 (73)	20
C1	41° 7' 33.55" S	71°12' 36.86" O	35.7 (2.2)	1.811	394 (119)	40
C2	41° 8' 53.55" S	71°13' 54.18" O	34.5 (1.7)	3.332	119 (25)	20

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733 Table 2: Parameters of our models for different measures of *Pinus contorta* abundance as
734 predictive variables and the following response variables:

735 a) Effect size on total productivity (EStp)

Model	Variable	Function	Estimate (linear)	Estimate (poly)	SE	t (linear)	p (linear)	t (poly)	p (poly)	R ²	AIC
1	density	linear	-0.002	-	0.0002	-10.873	< 0.0001	-	-	0.46	656
2		poly	-26.884	4.7271	2.4751	-10.862	< 0.0001	1.910	0.0583	0.48	654
3	basal area	linear	-0.145	-	0.0117	-12.392	< 0.0001	-	-	0.53	638
4		poly	-28.830	-2.1582	2.3277	-12.386	< 0.0001	-0.93	0.3550	0.53	639
5	cover	linear	-0.081	-	0.0091	-8.885	< 0.0001	-	-	0.37	670
6		poly	-24.030	-16.043	2.3370	-10.282	< 0.0001	-6.86	< 0.0001	0.53	631

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737 b) Effect size on palatable productivity (EspP)

Model	Variable	Function	Estimate (linear)	Estimate (poly)	SE	t (linear)	p (linear)	t (poly)	p (poly)	R ²	AIC
7	density	linear	-0.002	-	0.0002	-9.666	< 0.0001	-	-	0.40	685
8		poly	-26.588	5.906	2.7143	-9.795	< 0.0001	2.176	0.0313	0.42	682
9	basal area	linear	-0.139	-	0.0135	-10.277	< 0.0001	-	-	0.43	677
10		poly	-27.554	0.3774	2.6906	-10.241	< 0.0001	0.14	0.8890	0.43	679
11	cover	linear	-0.082	-	0.0098	-8.376	< 0.0001	-	-	0.34	691
12		poly	-24.388	-12.268	2.7250	-8.950	< 0.0001	-4.5	< 0.0001	0.43	673

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739 c) Effect size on the stocking rate that the grassland can sustainably support (Essr)

Model	Variable	Function	Estimate (linear)	Estimate (poly)	SE	t (linear)	p (linear)	t (poly)	p (poly)	R ²	AIC
13	density	linear	-0.001	-	0.0001	-8.399	< 0.0001	-	-	0.34	619
14		poly	-18.310	5.0612	2.1447	-8.537	< 0.0001	2.36	0.0197	0.36	616
15	basal area	linear	-0.100	-	0.0105	-9.592	< 0.0001	-	-	0.40	606
16		poly	-19.911	1.5593	2.0790	-9.577	< 0.0001	0.75	0.455	0.40	607
17	cover	linear	-0.062	-	0.0074	-8.373	< 0.0001	-	-	0.34	612
18		poly	-18.344	-8.5883	2.0711	-8.857	< 0.0001	-4.15	< 0.0001	0.41	598

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742 Figure Legends

743 Figure 1: a) Map of Argentina showing the location of our study region (red square). b) Map of our
744 study region showing the location of our study sites (yellow dots). c) Scheme of the experimental
745 design, showing the five transects of 500 m included in one site and the location of sampling
746 circular plots (10 m diameter) along each transect. d) Scheme showing one transect of 500 m as an
747 example. Each circle represents a plot, including a representative picture. In both c) and d) the
748 dotted line shows the location of the invasion front.

749

750 Figure 2: Conceptual model showing the different impacts measured here (orange bubbles) and
751 which factors (green boxes) are considered in their calculation.

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753 Figure 3: Estimate, 95% confidence and predictive intervals of the relationship between effect size
754 of pine invasion on a) total grassland productivity, b) palatable productivity and c) sheep stocking
755 rate, and pine density (left pane), basal area (center pane) and canopy cover (right pane). For each
756 predictive variable we show the best regression model (based on AIC). While the confidence
757 intervals show the likely range of values that contain each of the mentioned response variables
758 (effect sizes), the prediction intervals predict in what range a future individual observation will fall.
759 Prediction intervals are much wider than confidence intervals, since they show higher uncertainty.

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761 Figure 4: Estimate and 95% confidence intervals of the relationship between pine basal area and
762 density (left pane), between pine canopy cover and density (center pane) and between pine
763 canopy cover and basal area (right pane).

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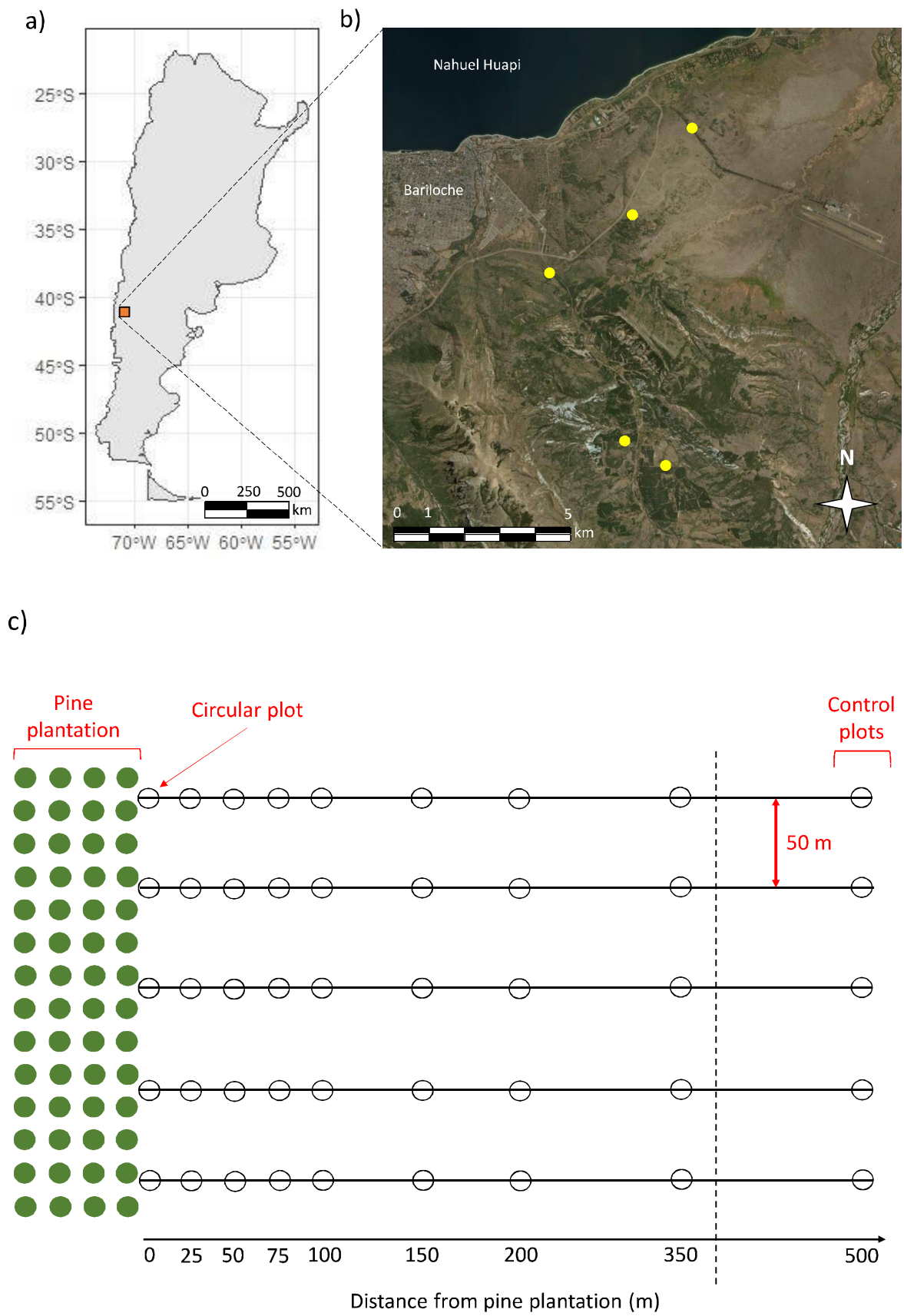
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d)

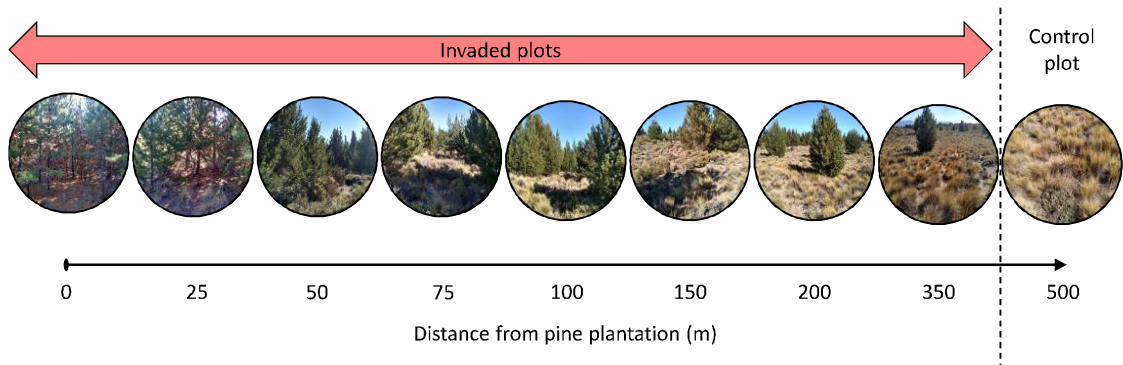


Figure 1: a) Map of Argentina showing the location of our study region (red square). b) Map of our study region showing the location of our study sites (yellow dots). c) Scheme of the experimental design, showing the five transects of 500 m included in one site and the location of sampling circular plots (10 m diameter) along each transect. d) Scheme showing one transect of 500 m as an example. Each circle represents a plot, including a representative picture. In both c) and d) the dotted line shows the location of the invasion front.

Figure 2

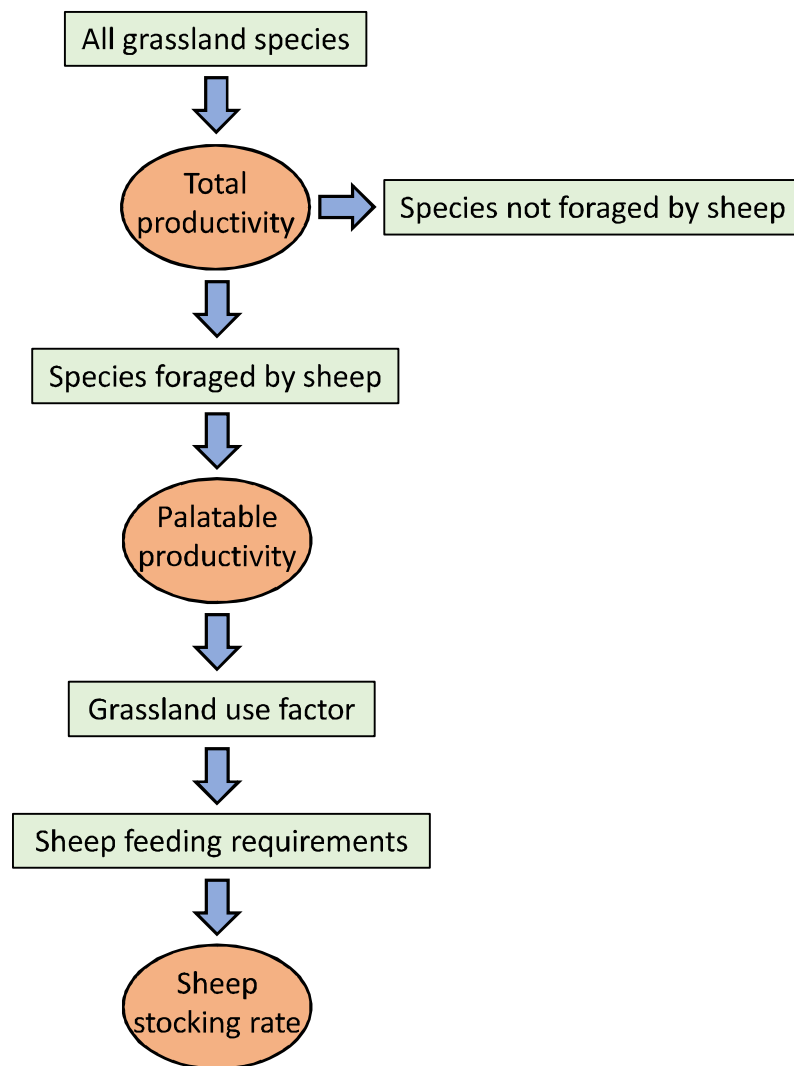
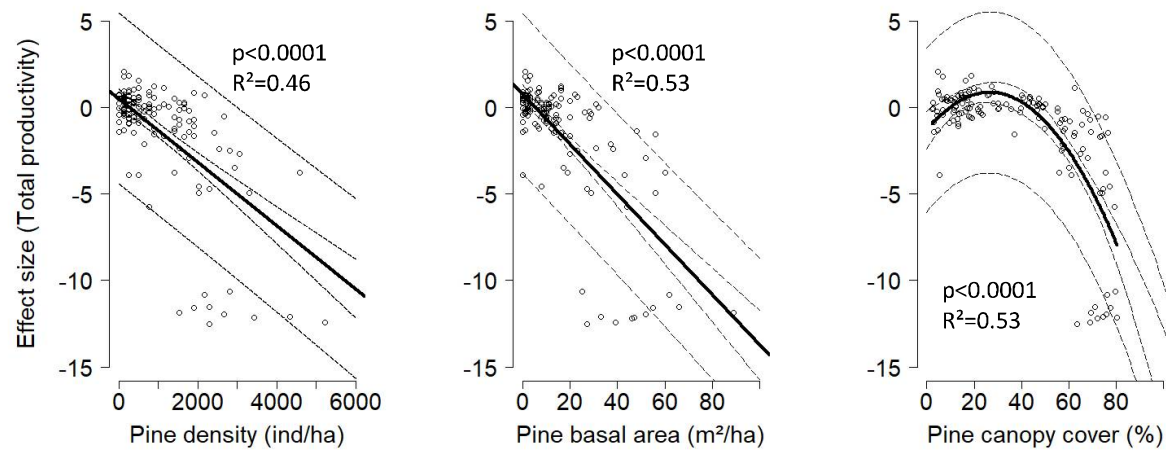


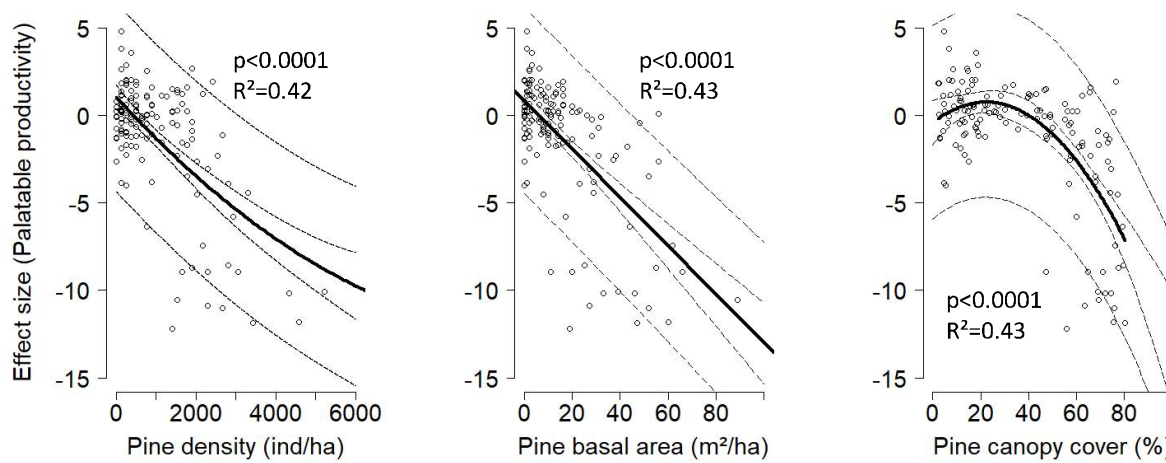
Figure 2: Conceptual model showing the different impacts measured here (orange bubbles) and which factors (green boxes) are considered in their calculation.

Figure 3

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b)



c)

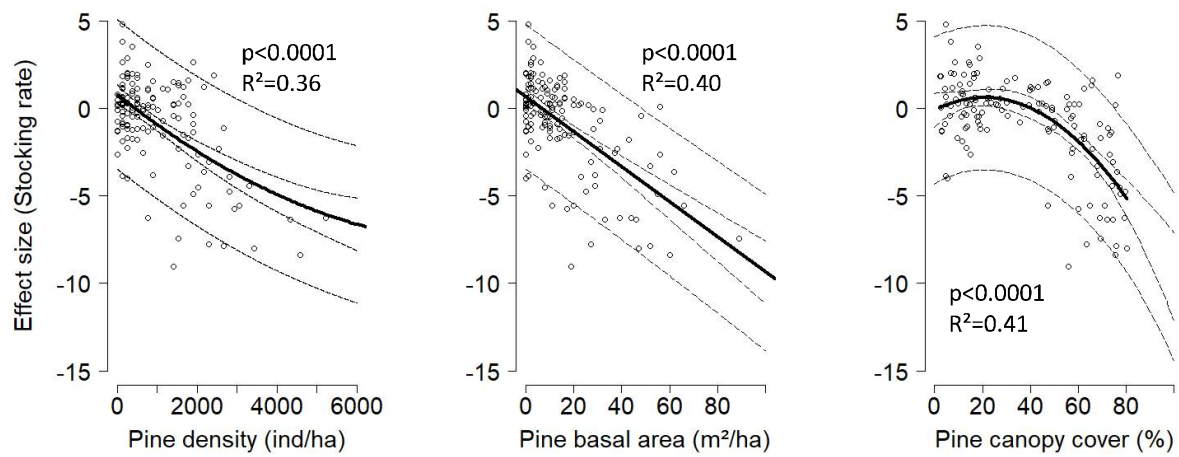
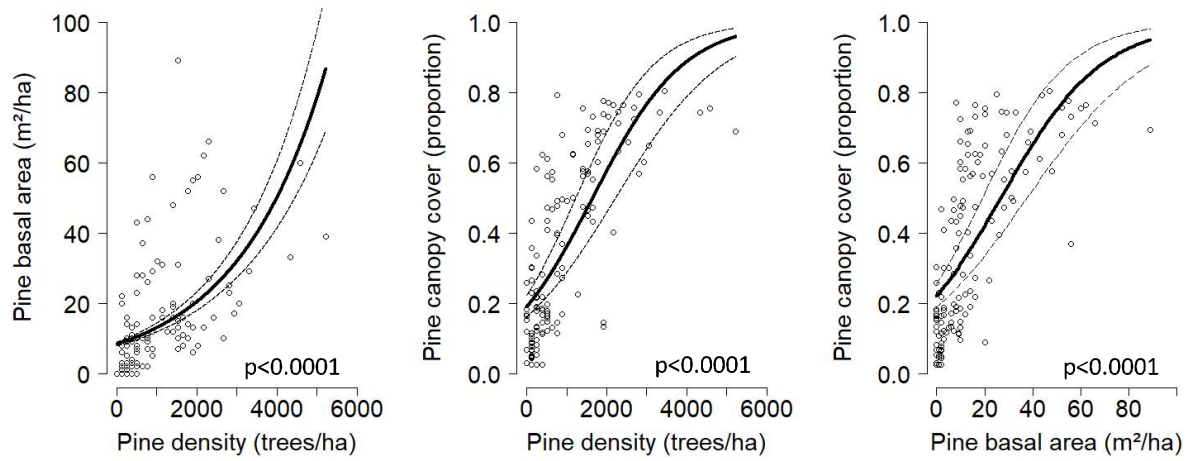


Figure 3: Estimate, 95% confidence and predictive intervals of the relationship between effect size of pine invasion on a) total grassland productivity, b) palatable productivity and c) sheep stocking rate, and pine density (left pane), basal area (center pane) and canopy cover (right pane). For each predictive variable we show the best regression model (based on AIC). While the confidence intervals show the likely range of values that contain each of the mentioned response variables (effect sizes), the prediction intervals predict in what range a future individual observation will fall. Prediction intervals are much wider than confidence intervals, since they show higher uncertainty.

1 Figure 4



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4 Figure 4: Estimate and 95% confidence intervals of the relationship between pine basal area and

5 density ([left pane](#)), between pine canopy cover and density ([center pane](#)) and between pine6 canopy cover and basal area ([right pane](#)).

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