

## Voluntary, permanent land protection reduces forest loss and development in a rural-urban landscape

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Abstract:	Voluntary, permanent land protection is a key conservation process in many countries. Concerns with the effectiveness of such decentralized processes exist due to the potential for 1) selection bias, i.e., the protection of parcels whose land cover would have been conserved in the absence of protection, and 2) local spillover effects, i.e., protection increasing the likelihood that adjacent parcels lose land cover due to additional conversion. We examine the validity of both concerns using a quasi-experimental approach and a dataset of 220,187 parcels and 26 years of protection and land-cover change in Massachusetts. We find that land acquisitions and conservation restrictions implemented by state, local, and non-governmental actors reduced forest loss and conversion to developed uses without increasing either type of land-cover change on adjacent parcels. Our results suggest that voluntary, permanent land protection can make significant contributions in protecting land cover in landscapes dominated by private ownership.

1 *Voluntary, permanent land protection reduces forest loss*  
2 *and development in a rural-urban landscape*

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28

## 29 Abstract

30 Voluntary, permanent land protection is a key conservation process in many countries.  
31 Concerns with the effectiveness of such decentralized processes exist due to the potential for 1)  
32 selection bias, i.e., the protection of parcels whose land cover would have been conserved in  
33 the absence of protection, and 2) local spillover effects, i.e., protection increasing the likelihood  
34 that adjacent parcels lose land cover due to additional conversion. We examine the validity of  
35 both concerns using a quasi-experimental approach and a dataset of 220,187 parcels and 26  
36 years of protection and land-cover change in Massachusetts. We find that land acquisitions and  
37 conservation restrictions implemented by state, local, and non-governmental actors reduced  
38 forest loss and conversion to developed uses without increasing either type of land-cover  
39 change on adjacent parcels. Our results suggest that voluntary, permanent land protection can  
40 make significant contributions in protecting land cover in landscapes dominated by private  
41 ownership.

For Peer Review

## 42 Introduction

43 The voluntary, permanent protection of land by multiple, decentralized actors is an important  
44 conservation process in many parts of the world. Over the past two decades, willing private  
45 landowners have transferred ownership or partial rights to millions of hectares of land to  
46 governments and conservation non-governmental organizations (NGOs), often in response to  
47 financial incentives (Parker & Thurman 2018). Here we consider such transactions “voluntary” if  
48 landowners have the option to not give up their land rights, and “permanent”, if the transaction  
49 does not oblige recipients to return the land rights in the future. Significant volumes of  
50 voluntary, permanent land protection (VPLP) transactions are occurring in Australia, Brazil,  
51 Canada, Chile, Colombia, Denmark, Finland, and the United States, and individual deals have  
52 been reported from at least 20 tropical countries (Nolte 2018). With rising global pressures on  
53 ecosystems, climate-induced species migrations, and growing societal discomfort with  
54 compulsory approaches to protection, the importance of VPLP is likely to increase.

55  
56 In spite of the substantial growth in VPLP transactions, their actual effectiveness is rarely the  
57 subject of rigorous scholarly scrutiny. Empirical studies have examined different aspects of  
58 VPLP, including preferences of landowners (e.g., Knight et al. 2011; Bastian et al. 2017), spatial  
59 patterns of protection (e.g., Meyer et al. 2014), or effects of tax incentives (e.g., Parker &  
60 Thurman 2018). However, few ex-post analyses quantify the difference that VPLP have made  
61 for conservation outcomes. This contrast with a rapidly growing empirical literature on impacts  
62 of other conservation strategies, including 1) compulsory approaches, e.g., protected areas (see  
63 Oldekop et al. 2016 for a recent review) and regulatory zoning / enforcement (e.g., Wu & Cho  
64 2007; Sims & Schuetz 2009; Börner et al. 2015; Nolte et al. 2017), 2) voluntary, non-permanent  
65 protection, such as payments for environmental services (PES) (see Börner et al. 2017 for a  
66 recent review) and alternative income generation strategies (e.g., Weber et al. 2011), 3)  
67 approaches that grant or clarify land rights, such as indigenous lands (e.g., Nolte et al. 2013;  
68 Blackman et al. 2017) or improvements in land-tenure security (Robinson et al. 2017), as well as  
69 4) supply-chain interventions (see Lambin et al. 2018 for a recent review).

70  
71 Whether VPLP transactions make a difference for conservation is not a trivial question.

72 Observed conditions on protected lands are not a good proxy for impact for two reasons: First,  
73 impact is the difference between observed outcomes in the presence of an intervention and  
74 outcomes that would have occurred in its absence (Ferraro 2009). Protecting a property under  
75 low pressure will usually result in lower impact than protecting a similar property under higher  
76 pressure; it will also, on average, generate lower opportunity costs and thus be cheaper or less  
77 controversial. The coincidence of both has created a pervasive global bias in the allocation of  
78 stricter conservation towards low-pressure locations (Joppa & Pfaff 2009), which has been  
79 shown to lower the true effectiveness of parks, PES, and other voluntary, non-permanent  
80 conservation interventions (Ferraro et al. 2011; Mason & Plantinga 2013; Börner et al. 2017).

81 This problem is likely to be even more pronounced when all transactions are voluntary, yet few  
82 empirical studies have accounted for possible bias in estimating the effectiveness of VPLP.

83

84 Second, any conservation intervention that focuses on a subset of properties has the potential  
85 to affect outcomes on properties that were not subject to it. Such spillover effects, also termed  
86 leakage or slippage, can manifest through diverse channels and across different scales (Wu  
87 2000; Alix-Garcia et al. 2012; Atmadja & Verchot 2012; le Polain de Waroux et al. 2017). Long-  
88 distance spillovers of environmental policies are commonly studied using economic models  
89 (Sohnngen et al. 1999; Wu & Plantinga 2003; Searchinger et al. 2008), whereas local spillovers  
90 can be investigated directly using spatial data (e.g. Wu 2000; Sims & Schuetz 2009; Alix-Garcia  
91 et al. 2012). In the context of VPLP, local spillovers are of concern as they can undermine the  
92 provision of spatially dependent ecosystem services (e.g., contiguous scenic landscapes or  
93 connected habitats). Empirical work has shown that protection increases sales value of adjacent  
94 properties (Reeves et al. 2018), and some scholars argue that this phenomenon can undermine  
95 the cost-effectiveness of future protection efforts (Armsworth et al. 2006). Furthermore, if  
96 increased property values increased the likelihood of habitat loss, this could attenuate the net  
97 impacts of protection. Using data from three U.S. counties and a regression framework,  
98 McDonald et al. (2007) find proximity to protected areas to be associated with higher rates of  
99 development in two counties, but not the third. In a study of one U.S. county, Zipp et al. (2017)  
100 find protected open space to reallocate parcel subdivision within a small neighborhood, which  
101 reduces the net impacts of protection. These initial findings suggest a need for more large-scale  
102 empirical studies to identify where and under which conditions local spillovers occur.

103  
104 Here we investigate the validity of both concerns in a setting spanning rural to urban land uses  
105 with a high incidence of VPLP. Our study area, Massachusetts, is an exemplar of the private land  
106 conservation movement in the United States, with 120 active land trusts (Land Trust Alliance  
107 2016), substantial direct public funding (\$53 million annually, 1998-2011) (The Trust for Public  
108 Land 2017), and tax incentives for charitable land donations. As with much of New England,  
109 Massachusetts experienced two centuries of deforestation, followed by 150 years of forest  
110 regrowth, and, since the 1980s, a slow but continuous loss of forest cover, mostly due to low-  
111 density development (Olofsson et al. 2016). VPLP occurs for diverse reasons, including species  
112 conservation, local recreation, the preservation of cultural landscapes, and, more recently, the  
113 maintenance of carbon stocks. It can involve full acquisition by NGOs or public actors or the  
114 transfer of partial rights (known as “conservation restrictions” in Massachusetts, and  
115 “conservation easements” elsewhere in the U.S). Using a rich parcel dataset from the entire  
116 state, we estimate whether protection helped slow down forest loss and development. We find  
117 that most types of protection significantly reduced land-cover change on protected parcels,  
118 without leading to an increase of land-cover change on adjacent, unprotected parcels. Our  
119 results suggest that VPLP can play an important role in protecting conservation values along  
120 rural-urban gradients in high-income countries.

## 121 Methods

### 122 Data

123 We use spatial boundaries of parcels from the public MassGIS system (Commonwealth of  
124 Massachusetts 2018). MassGIS aggregates parcel layers from 351 towns, 66% of which reflect

125 recent conditions (2017 or 2018), while some date back to 2010. We include in our analysis all  
126 220,187 parcels in the state with an area of no less than 1 ha (2.47 ac). The threshold ensures  
127 that parcels contain sufficient units of land-cover observations (pixels) to reliably observe  
128 change at 30m resolution. To ensure this size threshold does not affect results, we supplement  
129 the parcel-based analysis with a pixel-based analysis (see below).

130

131 All protection data – including spatial boundaries, years of protection, conservation actors  
132 (local, state, non-governmental), and instruments (full acquisition or conservation restriction) –  
133 come from a database maintained by the Harvard Forest and the Highstead Foundation, which  
134 aggregates multiple public sources and supplements them with information from private land  
135 trusts, and was last updated in 2018.

136

137 Land-cover change estimates are derived from a dataset developed by Oloffson et al. (2016),  
138 which uses Landsat time series and a spectral break detection algorithm to map annual changes  
139 in 12 land-cover categories across New England from 1985 to 2012 at 30m spatial resolution.  
140 For each parcel, we extract annual % forest cover (deciduous, coniferous, and mixed) and %  
141 developed land (commercial, high-density, and low-density). Oloffson et al.'s data is known to  
142 underestimate conversion of forests to low-density development, but this underestimation is  
143 not known to be spatially biased (P. Oloffson, personal communication, 2018). We therefore  
144 consider the data suitable for quantifying impacts in relative terms (observed outcomes as % of  
145 estimated pressure), but caution against interpreting rates of avoided land-cover change in  
146 absolute terms.

147

148 We assume that probabilities of protection and land-cover change are influenced by a parcel's  
149 potential returns from alternate uses, which in turn is a function of its physical properties (e.g.,  
150 terrain, proximity to water), accessibility, and socio-economic setting (Irwin & Bockstael 2004).  
151 To control for key differences that might affect both protection and land-cover change, we  
152 compute a range of covariates for each parcel (discussed below and in Table 1). More details on  
153 data sources are provided in the Supporting Information.

## 154 Impact Estimation

155 We use quasi-experimental pre-matching followed by regression analysis to estimate the  
156 impact of voluntary, permanent protection of private lands on the loss of forest cover and  
157 undeveloped land. Ideally, matching emulates an experimental setup from observational data  
158 by identifying control groups of untreated (unprotected) parcels that, at the time of treatment,  
159 were as similar as possible to treated (protected) parcels in terms of observable confounders.  
160 By capturing key differences in terrain, water, accessibility, demographics, parcel size, and  
161 nearby protection (Table 1), we control for several well-known sources of selection bias that are  
162 of common concern in impact evaluations of conservation interventions. We minimize these  
163 differences with the use of pre-matching and then control for them explicitly using regression  
164 analysis. Because matching does not allow us to control for unobserved sources of bias (e.g.,  
165 individual landowner preferences or scenic appeal), we conduct sensitivity checks to assess the

166 vulnerability of our findings to the potential presence of unobserved confounders (see  
167 Supporting Information).

168  
169 We conduct two distinct analyses. First, we measure the impact of protection on the loss of  
170 forest and undeveloped land within protected parcels (hereafter, “impact analysis”). Our  
171 treatment group consists of parcels that experienced an increase in protection coverage of >  
172 80% between 1985 and 2006 ( $n=6,676, 1,120 \text{ km}^2$ ). We include only parcels protected before  
173 2006 in order to have a reasonably long time period for the observation of post-protection  
174 outcomes. Our pool of potential controls includes all parcels that remained “unprotected” until  
175 the present (defined as having < 20% of their area protected,  $n=182,982, 9,527 \text{ km}^2$ ).

176  
177 Second, we measure the impact of protection on the loss of forest and undeveloped land on  
178 nearby parcels (hereinafter, “spillover analysis”). Our treatment group consists of all  
179 unprotected parcels that experienced an increase in protection of at least 1% within a given  
180 radius (default: 200m) between 1985 and 2006 ( $n=29,296, 1,965 \text{ km}^2$ ), in which case the year of  
181 the greatest increase was defined as the treatment year. Our pool of potential controls includes  
182 all unprotected parcels that did not experience such an increase in nearby protection  
183 ( $n=144,332, 6,940 \text{ km}^2$ ). Because neighboring parcels are frequently contiguous, we conduct  
184 matching of 25% samples with 20 repetitions to reduce the likelihood of spatial autocorrelation.  
185 We present average results in the figures, and their distribution in the Supporting Information.

186  
187 We measure outcomes as the average annual change in the land cover of interest (forest or  
188 undeveloped, as percentage of parcel area). For each treatment-control pair, the time period  
189 over which land-cover change is observed begins in the year in which the treatment parcel was  
190 protected (spillover analysis) or three years after (impact analysis), and ends in 2012, the last  
191 year for which land-cover data is available. The three-year offset in the impact analysis is added  
192 to reduce the influence of a small number of parcels with large land-cover losses that co-  
193 occurred with protection. We observe such losses in the case of conservation restrictions held  
194 by local governments and placed on parcels with new subdivisions and golf courses, which  
195 imply that protection was created specifically to accompany planned development. As our data  
196 does not allow us to separate such planned development restrictions from those we aim to  
197 study here (direct acquisitions or donations independent of development), we use an offset and  
198 implement a several alternative robustness checks (Supporting Information).

199  
200 To account for possible selection bias, we use Mahalanobis nearest neighbor covariate  
201 matching (one neighbor, with replacement), and post-matching linear regressions. Regressions  
202 predict annual observed land-cover change as a function of all covariates and a continuous  
203 treatment variable (impact analysis: % of parcel protected, spillover analysis: % increase in  
204 protection within given radius). All observations are weighted by parcel area. To explore how  
205 threats and impacts vary as a function of location, we split matched samples at the 33% and  
206 67% quantiles for each covariate and estimate impacts for each subgroup separately. Our  
207 default setting uses calipers of 1 standard deviation, which retains 70-75% of treated parcels  
208 (59-66% of area) in the impact analyses and 72-75% (62-67% of area) in the spillover analyses,  
209 respectively, dropping the remainder because of the absence of comparable controls.



210  
211 Our default analyses are based on parcels, as they constitute the key decision unit in private  
212 land protection. However, state-wide parcel boundaries were only available for post-treatment  
213 time periods (2010-18). Because we include parcel area in both sample definition and matching,  
214 protected parcels are less likely to be matched correctly to controls that might have originally  
215 been the same size and were subsequently sub-divided and developed. For this reason, the use  
216 of post-treatment parcel boundaries might lead to an underestimation of impact. We therefore  
217 supplement parcel-based analyses with corresponding pixel-based analyses that are not  
218 vulnerable to this type of bias.

219  
220 We also conduct extensive robustness checks with alternative model specifications (see  
221 Supporting Information).

## 222 Results

223 Across all model runs, we find protected parcels to have significantly lower levels of forest loss  
224 and development than they would have experienced in the absence of protection (Figure 2).  
225 Our main results suggest that protection avoided about half of forest loss ( $-55\% \pm 30.8\%$ ) and  
226 about four fifths of development ( $-83\% \pm 27\%$ ).

227  
228 Differences in impact are mostly driven by differences in pressure levels along rural-urban and  
229 income gradients rather than by differences in observed outcomes on protected parcels. For  
230 instance, protected parcels in high-income locations and parcels close to cities were exposed to  
231 significantly higher levels of development pressure ( $0.063 \pm 0.015$  and  $0.078 \pm 0.020$  % loss /  
232 year, respectively) than parcels in low-income locations and parcels further away from cities  
233 ( $0.006 \pm 0.009$  and  $0.008 \pm 0.008$ , respectively). In contrast, the observed rates of development  
234 are remarkably similar (Figure 3).

235  
236 We find significant impacts for both protection instruments (fee or conservation restrictions)  
237 and conservation actors (local, state, or NGO), but not for all combinations (Figure 4). For  
238 instance, impact of conservation restrictions on forest loss are only significant at the 10% level  
239 ( $p=0.08$ ), and weak for local ( $p=0.80$ ) and NGO-held restrictions ( $p=0.87$ ). Not all subgroup  
240 impact estimates are robust (see Supporting Information).

241  
242 Unprotected parcels that experienced an increase in nearby protection between 1985-2006 did  
243 not exhibit higher levels of forest loss or development than their pressure estimate (Figure 2).  
244 Sizes of estimated spillover effects are near zero for forest loss ( $p=0.50$ ); for development, they  
245 point in the opposite direction ( $p=0.14$ ). We did not find significant spillover effects (at the  $\leq$   
246 0.05 level) for any locations along the rural-urban gradient (Figure 3). Our robustness checks  
247 confirm that spillovers of protection are either positive or absent (see Supporting Information).



## 248 Discussion

249 Our results suggest that voluntary, permanent land protection reduced forest loss and  
250 development in Massachusetts between 1985 and 2012. We find significant and robust  
251 evidence of reductions in both types of land-cover change on protected parcels. In addition, we  
252 find no evidence that protection increased nearby land-cover change. Taken together, these  
253 findings suggest that voluntary, permanent land protection delivered tangible conservation  
254 results in Massachusetts.

255  
256 On average, percentage reductions in land-cover change were higher for development than for  
257 forest loss. This is consistent with expectations. In Massachusetts, most VPLP transactions  
258 extinguish development rights, while forest conversion is not always regulated by conservation  
259 restrictions and might even be desired to improve conservation outcomes (e.g., through the  
260 creation of early successional habitat). We also note that absolute land-cover change on  
261 protected parcels in Massachusetts varies very little as a function of pressure. Protection thus  
262 appears to deliver consistently low loss of land-cover change along the state's rural-urban  
263 gradients – which differs notably from findings from other world regions: in the Brazilian  
264 Amazon, for instance, forest loss inside protected areas has been found to be much higher  
265 where counterfactual pressure is high (e.g., Nolte et al. 2013b), likely as a result of imperfect  
266 enforcement (Robinson et al. 2010).

267  
268 Our findings have at least two important policy implications. First, we show that differences in  
269 the impact of VPLP transactions in Massachusetts are largely driven by differences in pressure,  
270 not by differences in observed outcomes. This finding illustrates the caveats of relying on  
271 observed outcomes as estimates of “success” and underscores the need for a rigorous  
272 quantification of pressure. Data and methods now exist to develop counterfactual pressure  
273 estimates at decision-relevant spatial scales (parcels). If combined with empirically grounded  
274 and spatially disaggregated data on conservation costs (Armsworth 2014), such estimates can  
275 help conservation decision makers target investments to where they are likely to generate the  
276 highest conservation returns (Newburn et al. 2006). Their systematic inclusion in the allocation  
277 and evaluation of conservation interventions could enhance the targeting of a range of VPLP  
278 occurring in Massachusetts today, such as forest-based carbon credits, natural resources  
279 damage compensation programs, direct public land acquisitions, and environmental  
280 philanthropy.

281  
282 Second, previous concerns about negative local spillover effects of protection may require  
283 greater scrutiny. Our evidence suggests that spillover effects are largely negligible, and, if they  
284 exist, more likely positive than negative. A possible explanation for this phenomenon might be  
285 that protection increases informal, voluntary conservation by surrounding landowners.  
286 However, our finding contrasts with those of earlier studies in the U.S., which find protection to  
287 generate negative local spillovers in other locations in the U.S. (McDonald et al. 2007; Zipp et al.  
288 2017). We also do not include in our analysis possible long-range spillovers effects that tend to  
289 be more difficult to identify empirically. More exploration of the possible mechanisms that  
290 drive positive spillovers is thus an important area of future empirical research.

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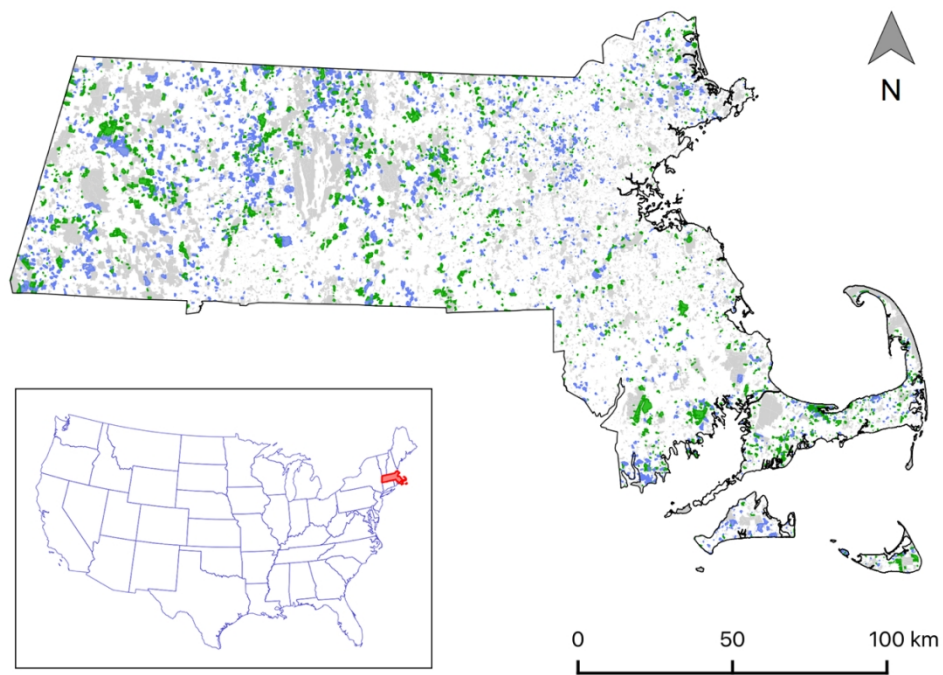


Figure 1: Protection of land between 1985 and 2006 in Massachusetts in the form of fee title acquisitions (green) and conservation restrictions (blue). Grey areas were protected either before 1985 or after 2006 and thus excluded from the analysis. Inset shows the location of Massachusetts within the U.S.

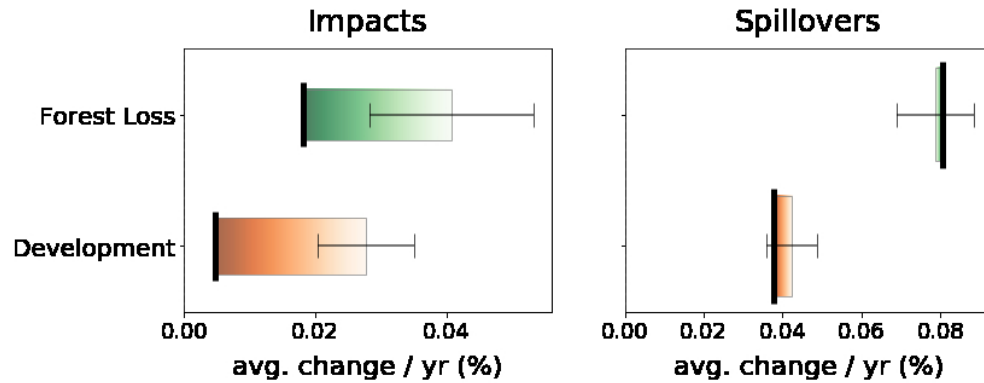


Figure 2: Estimated effects of voluntary, permanent land protection in Massachusetts (1985-2006) on forest loss and development (1985-2012) on protected ("Impacts") and nearby parcels ("Spillovers"). In this figure and all following, 1) all values refer to average annual land-cover change (as percentage of parcel area), 2) magnitudes of estimated effects are represented by rectangle widths, with bold black lines indicating predicted rates of land-cover change on treated parcels, the opposite end indicating predicted rates of land-cover change in the absence of treatment (counterfactual), and shading added to indicate direction of effect, 3) error bars indicate 95% confidence intervals around the effect estimate, 4) bar height is proportional to the total area of parcels in the matched treatment group; bar area is therefore proportional to total area of avoided (or increased) land cover change.

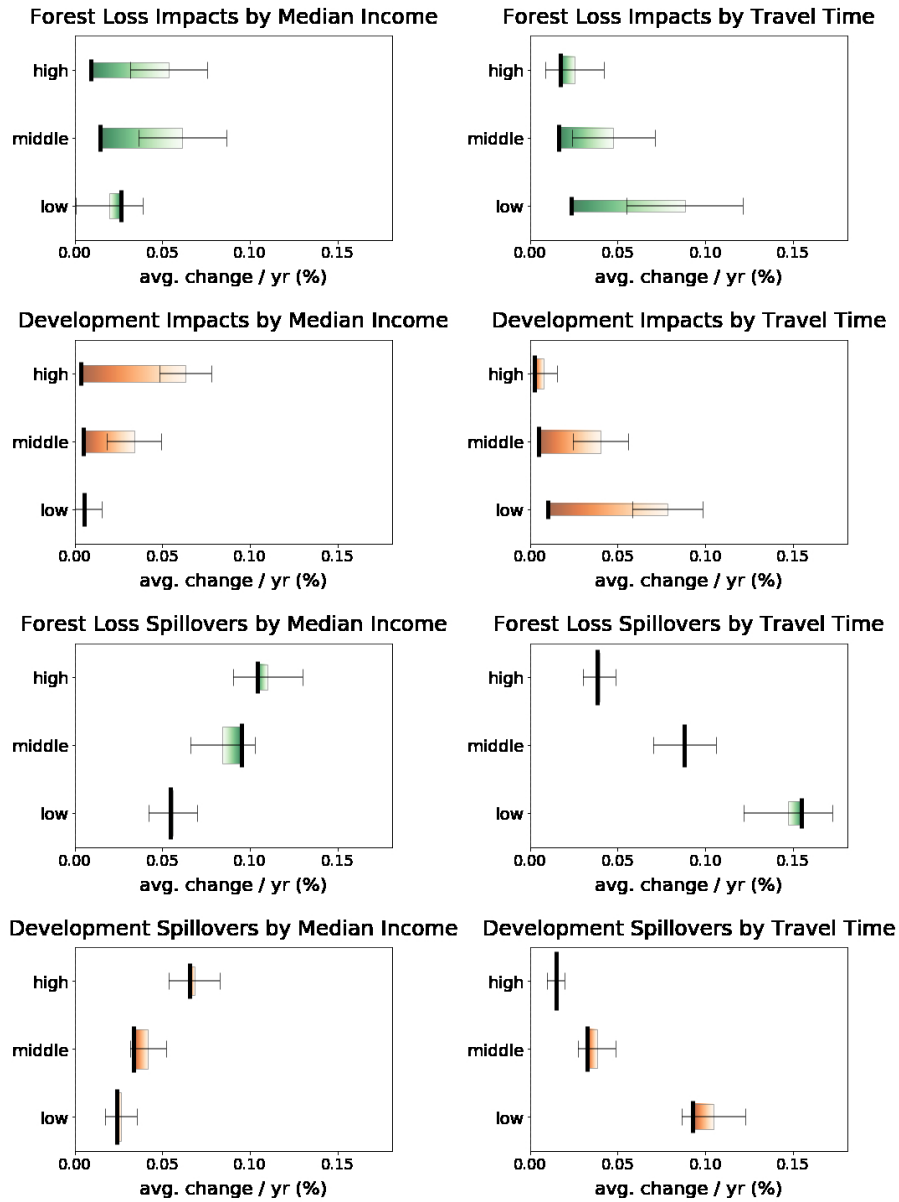


Figure 3: Estimated effects for selected subgroups of protected parcels. Subgroups are formed by splitting matched treatment groups at the 33% and 67% quantiles for the respective covariate. Legend as in Figure 2.



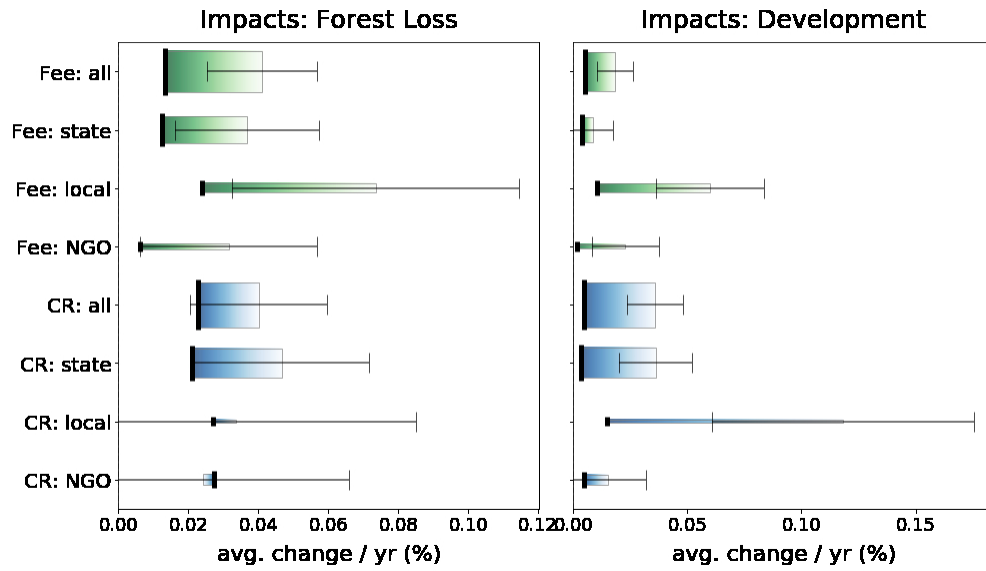


Figure 4: Estimated effects for protected parcels held in fee (green) or as conservation restrictions (CR, blue), for all actors and disaggregated by actors (state, local, and non-governmental). Legend as in Figure 2

*Table 1: Covariates used in matching, with justification. IHS: inverse hyperbolic sine transformation. Data source and further details can be found in the Supporting Information and Table S1.*

*\* variables that can be considered time-invariant within the study period,*

*\*\* time-variant variable for which no earlier data source was available.*

<b>Covariate</b>	<b>Unit</b>	<b>Year</b>	<b>Justification for Selection</b>
Slope	degree	2017*	Key driver of agricultural potential and suitability for development
Wetland coverage	% of parcel area	2018*	Creates both physical and legal obstacles to conversion
Proximity to coastal waters	% ocean area within 2.5km radius	2009*	Increases attractiveness to development and thus the cost of protection
River and lake frontage	meters (IHS)	2017*	Increases attractiveness to development and thus the cost of protection
Travel time to major cities	minutes (IHS)	2007*	Key driver of accessibility to markets, workplaces, and amenities
Median income, (block group)	USD	1990	Affects local development pressure and land prices
Population density (block group)	km <sup>-2</sup> (IHS)	1990	Affects local development pressure
Parcel size	hectares (log)	2010-2018**	Affects economies of scale and transaction costs of protection
Coverage of land cover of interest	% of parcel area	Year of protection (1985-2006)	Caps the quantity of forest or undeveloped land that can be lost
Nearby protection	% protected area within given radius (default: 200m)	Year of protection (1985-2006)	Accounts for local spillover effects, which can be positive or negative