# Decentralized land use zoning reduces large-scale deforestation in a major agricultural frontier

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#### 87 Abstract

88 Reducing large-scale deforestation is a key objective of global efforts to mitigate climate 89 change. An important debate concerns the levels of governance at which deforestation 90 can be reduced effectively. Political economic theory and evidence suggests that 91 national governments are more likely than subnational governments in agricultural 92 frontiers to adopt restrictive forest conservation policies, due to differences in political 93 constituencies and capacity. Here we examine the validity of this claim using an impact 94 study of provincial-level land use planning in Argentina's main deforestation frontier, 95 the Dry Chaco. In 2007, Argentina's provinces were obliged to define land use zoning for 96 their native forests, but had considerable leeway in its implementation. We use data 97 from 30,126 properties in the provinces of Salta, Santiago del Estero, and Chaco, and a 98 rigorous counterfactual estimation strategy to quantify the extent to which adopted 99 zoning plans affected deforestation. We find evidence that provincial-level land use 100 zoning reduced deforestation in all three provinces, but not in all zones and periods. 101 Differences in impact are associated with differences in the location of zones and the 102 timing of planning. Our findings suggest that subnational governments can make 103 important contributions to reducing large-scale deforestation in agricultural frontiers. 104 105 Highlights 106 Subnational governments might become key players in climate change • 107 mitigation. 108 We study the impact of provincial land use zoning on deforestation in Argentina. 109 Provinces allocated zones in ways that limited potential impacts on forest loss. 110 However, provincial zoning plans also reduced forest loss over baseline. 111 • Our findings lend support to nested approaches to avoided deforestation policy. 112 113 Keywords 114 Argentina, Chaco, avoided deforestation, subnational, governance, matching 115

#### 117 **1. Introduction**

118 Reducing emissions from large-scale deforestation constitutes a priority for global 119 efforts to mitigate climate change. Tropical forest loss accounts for about 10% of 120 anthropogenic greenhouse gas emissions (Baccini et al., 2012; Harris et al., 2012). Large-121 scale forest conversion in the tropics and subtropics in the 21<sup>st</sup> century was largely the 122 result of agricultural expansion for the production of globally traded commodities such 123 as soy, beef, palm oil, and timber (Gasparri et al., 2015; Newton et al., 2013; Rudel et al., 124 2009). To conserve global forests and associated ecosystem services, multilateral, 125 bilateral, and private donors have begun to incentivize reductions in deforestation and 126 forest degradation (REDD+) with billions of dollars in funding (Agrawal et al., 2013; Silva-127 Chávez et al., 2015). However, considerable academic and political debate surrounds 128 the choice of strategies and policies that can effectively reduce deforestation at large 129 spatial scales (Angelsen, 2010; Larson et al., 2013; Patel et al., 2013).

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131 An important question for climate change mitigation scholars and policy makers 132 concerns the level of governance at which large-scale deforestation can be addressed 133 effectively (Angelsen et al., 2008). National governments have long held a privileged 134 position among the actors involved in forest-based climate change mitigation, both as 135 decision makers in international negotiations and as recipients of early funding flows. 136 However, scholars have also proposed that an exclusive focus on national governments 137 will not necessarily lead to effective and equitable avoided deforestation policy (Luttrell 138 et al., 2013; Phelps et al., 2010). Strategies to engage other levels of government in the 139 design of such policies have therefore become a major subject of inquiry, with authors 140 examining the feasibility of "jurisdictional", "multiscale" or "nested" approaches (e.g. Agrawal et al., 2011; Cattaneo, 2011; Fishbein and Lee, 2015; Pedroni et al., 2009). 141

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143 In practice, subnational governments are already actively involved in avoided 144 deforestation efforts across the globe (Ravikumar et al., 2015; Sunderlin et al., 2014), 145 and continue to position themselves as partners for forest conservation in international 146 arenas. Thirty-five subnational governments from nine countries (including Brazil, 147 Indonesia, and the USA) cooperate in the Governors' Climate and Forests Task Force 148 (GCF) to advance jurisdictional programs for reducing emissions from deforestation and 149 land use (GFC, 2016). In the New York Declaration on Forests, twenty subnational 150 governments from tropical countries publicly committed to end deforestation by 2030 151 (UN Climate Summit, 2014). And in Brazil, the two Amazon states of Acre and Mato 152 Grosso moved forward and signed jurisdictional REDD+ frameworks into law in 2010 and 153 2013 (Duchelle et al., 2014).

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155 In spite of this rising interest in the role of subnational policies to reduce deforestation,

the extent to which subnational governments are willing and able to inhibit agricultural

- 157 expansion in active deforestation frontiers remains largely unexamined. Initial
- reflections on the motivation and capacity (Lambin, 2005) of different levels of
- 159 governments lead us to assume that subnational governments are <u>less</u> likely than

160 national governments to engage in large-scale forest protection. This is because the 161 constituencies of national governments can be expected to be more urban, with higher 162 incomes and educational levels, and lower dependence on agricultural expansion, than 163 those of subnational governments in active deforestation frontiers. Such attributes are 164 generally associated with a higher willingness to pay for forest protection (Vincent et al., 165 2014), and might translate into stronger political support and motivation for national 166 governments to implement new effective forest conservation measures. National governments might also have a more diverse range of legal instruments, higher budgets, 167 168 and better enforcement resources at their disposal (Lambin et al., 2014), which might 169 convey them a higher overall capacity to implement strict forest conservation 170 instruments.

171

172 Empirical evidence exists for multiple cases in which national governments were willing 173 and able to implement effective forest protection policies at large spatial scales. For 174 instance, recent substantial downturns in large-scale deforestation in Brazil, China, and 175 Vietnam have been, in large part, ascribed to national forest conservation policies (Liu et 176 al., 2008; Meyfroidt et al., 2009; Nepstad et al., 2014). Sweeping deforestation bans, 177 such as those adopted in China and the Atlantic Forests of Brazil and Paraguay, were 178 also driven by national governments. Examples of major conservation policies outside 179 active deforestation frontiers, such as the U.S. Northwest Forest Plan (Thomas et al., 180 2006), protected area declarations in 1990 East Germany (Garrelts et al., 2005), and the 181 European Union's Natura 2000 directive (Kati et al., 2015), provide further evidence for 182 national (and supra-national) leadership in large-scale nature conservation – and for 183 opposition of subnational actors against such policies.

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185 Meanwhile, evidence on the impact of subnational policy on forest conservation in 186 active deforestation frontiers remains scarce. A recent review finds that a majority of existing rigorous studies of the impacts of decentralized forest governance examine 187 188 forest degradation, not deforestation (Miteva et al., 2012). Of the three rigorous studies 189 studying deforestation outcomes, none finds decentralization to reduce forest loss 190 (ibid.). In Indonesia, decentralization increased deforestation, especially before elections 191 (Burgess et al., 2011); in the Brazilian Amazon, federally protected areas reduced 192 deforestation, while state parks did not (Pfaff et al., 2012); and in Bolivia, better 193 municipality-level forest governance was associated with reductions in unauthorized 194 deforestation, but not total deforestation (Andersson and Gibson, 2007). These findings 195 lend support to the hypothesis that subnational governments are more likely than 196 national governments to prioritize local economic interests over the conservation of 197 ecosystems. This phenomenon has also been observed and described as "zoning 198 following the market" in the context of residential use (Pogodzinski and Sass, 1994; 199 Wallace, 1988).

200

Here we provide empirical evidence that subnational approaches to forest conservation
 can significantly reduce large-scale deforestation in active subtropical agricultural
 frontiers. We base our finding on a rigorous impact analysis of provincial-level land use

204 planning in the Argentinian Dry Chaco. In 2007, Argentina's federal government obliged

205 provinces to implement land use zoning for their remaining native forests. Using data

from 30,087 properties located in the three provinces with the highest historical rates of

forest loss (Lende, 2015), we show that the provinces implemented land use plans in ways that significantly reduced property-level deforestation in the short term. As

209 provinces had considerable leeway in the implementation of the law, we interpret these

- 210 impacts as partial evidence for the motivation and ability of provincial governments to
- 211 reduce deforestation.

# 212 2. Argentina's Dry Chaco and the 2007 Forest Law

Argentina's Dry Chaco is a vast semiarid plain located in the country's northwestern
region. Its subtropical forest ecosystems are characterized by rich levels of biodiversity
(Bucher and Huszar, 1999; Giménez et al., 2011) and globally significant carbon stocks

216 (Baumann et al., 2016; Gasparri et al., 2008). Throughout the late 20<sup>th</sup> and early 21<sup>st</sup>

217 centuries, the Dry Chaco witnessed some of the world's highest deforestation rates,

218 mostly due to the expansion of large-scale soy and beef production by well-capitalized

agribusinesses (Aide et al., 2013; Gasparri and Grau, 2009; Vallejos et al., 2015).

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221 In Argentina, provinces are the constitutional original owners of natural resources, and 222 entitled to manage land and forests within their territories (Article 124 of Argentina's 223 1994 National Constitution). Each province designs its own laws, directives, processes, 224 and administrative structures to define, allocate, and enforce rights to land and its use. 225 However, in order to guarantee all Argentinians the right to a healthy environment 226 across provincial borders, the constitution also allows the federal government to define 227 minimum standards for environmental protection (presupuestos mínimos, Article 41). If 228 such standards are adopted, provinces are obliged to translate them into provincial law. 229

In the wake of rapid deforestation, catastrophic floods, and resulting societal pressure
(Romero, 2012), the Argentinian federal government made use of this constitutional
provision to define minimum standards for the protection of native forests. The Law
#26.331 of 2007, hereinafter referred to as the Forest Law, obliged provinces to conduct
a land use planning process with the goal to categorize all remaining native forests into
three zones with different levels of protection (García Collazo et al., 2013; Gautreau et
al., 2014):

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- Category 1 (red): forests of high conservation value, which require permanent protection, but can be used by indigenous communities or for research.
- Category 2 (yellow): forests of medium conservation value, which can be used for sustainable resource use, tourism, gathering, or research.
- Category 3 (green): forests of low conservation value, which can be converted
   partially or completely.
- 244

245 The Forest Law defined several procedural criteria for the land use planning process,

- which the federal government proved willing to enforce. For instance, it required the
- 247 planning process to be "participatory", and prohibited the issuance of deforestation
- 248 permits until the process was concluded. Argentina's government rejected the land use
- 249 plan of at least one province (Córdoba), because it allowed extractive activities in red
- 250 zones and did not follow a participatory process (Silvetti et al., 2011). In another
- province (Salta), the Supreme Court of Justice revoked all deforestation permits and
   imposed a moratoria on deforestation in 2009, because the provincial government had
- not concluded its land use planning process (Di Paola et al., 2011).
- 254

255 However, within these procedural boundaries. Argentina's provinces appeared to enjoy 256 considerable leeway in the allocation and implementation of the three zones across 257 their jurisdiction. First, the Forest Law does not define a minimum percentage of native 258 forests in each province that needs to be protected. Second, while the Forest Law lists 259 ten socio-ecological criteria to characterize the conservation value of forests, it does not 260 stipulate how these criteria are to be translated into zones. Third, the law remains 261 ambiguous about the activities allowed under "sustainable resource use" in yellow 262 zones. Specifically, it does not define guidelines for silvopastoral systems, an expanding land use in the Dry Chaco, in which understory is removed to grow grasses underneath a 263 264 relatively sparse tree canopy (Grau et al., 2015; Mastrangelo and Gavin, 2012). Because 265 silvopastoral systems can lead to a slow transition from forest cover to pasture, this 266 regulatory absence constitutes a potential loophole for legal deforestation.

267

268 While the Forest Law did not dictate how to allocate protective zones, it offered some 269 financial incentives for stricter protection through a "National Fund for the Enrichment 270 and Conservation of Native Forests" (NFECNF), which indicated that funds would be 271 allocated as a function of the size of stricter zones in each province. Similar policies exist 272 in other deforestation frontiers, e.g. in Brazil (Ring, 2008; Sauquet et al., 2014). Although 273 the NFECNF ended up chronically underfunded, receiving less than 10% of the envisaged 274 budget between 2010 and 2015 (Ministerio de Ambiente y Desarrollo Sustentable de la 275 Nación, 2016), it might have incentivized provinces to zone larger areas of forests under 276 stricter categories. However, as deforestation risk is highly heterogeneous across the 277 Argentinian Chaco (Piquer-Rodriguez et al., 2015), provinces were also in a position to 278 allocate stricter protection to the most remote forests that were already protected by 279 virtue of their location (Joppa and Pfaff, 2010, 2009).

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281 Within two to five years after the promulgation of the Forest Law, most of Argentina's 282 provinces had concluded their land use planning processes and signed the results into 283 law. The resulting maps and regulations reveal substantial variation in the interpretation 284 of the zoning criteria. For instance, among nineteen provinces, the percentage of forest 285 allocated to the strictest category (red) varies between 2% and 80% (Gautreau et al., 286 2014). Zones frequently exhibit discontinuities at provincial boundaries, which do not 287 appear to correspond to differences in forest condition or conservation value (García 288 Collazo et al., 2013). Forests with particularly high ecological values (e.g. remnants of

289 quebracho forests) were often assigned weak protection, probably due to their vicinity 290 to valuable agricultural land (Piquer-Rodriguez et al., 2015). Provinces also differ widely 291 in the percentage of deforestation that is allowed in different zones (Gobbi, 2015). 292 Although no previous studies quantify how landscape factors influenced the allocation 293 of zoning in Argentina's native forests, this existing evidence appears to corroborate the 294 view that provincial governments had substantial leeway to shape the zoning plan in 295 ways that served their interests. 296 297 Did these subnational land use planning processes reduce deforestation in Argentina's 298 major deforestation frontier? Conventional wisdom suggests that provincial 299 governments had little incentive to adopt zoning plans that would inhibit agricultural 300 expansion. The economies of the four main provinces in the Dry Chaco – Salta, Santiago del Estero, Chaco, and Formosa – are strongly dependent on agriculture. In 2003, the 301 302 primary sector contributed 22%, 21%, 16%, and 17% respectively, to their provincial 303 gross domestic product (Ministerio de Economía y Finanzas Públicas, 2012). For 304 comparison with other soy-and-beef frontiers in the area, these percentages are about 305 twice as high as for Argentina as a whole, three times higher than for the country of 306 Brazil, which reduced deforestation considerably, and of similar magnitude as in 307 Paraguay and Bolivia, where deforestation is on the rise (Nolte et al., 2016). If provinces 308 had the option to "follow the market" by allocating stricter zones to forests not exposed 309 to the risk of deforestation, we would expect the impact of zoning on deforestation to 310 be negligible, especially in the short term. In what follows, we examine empirically

311 whether this was the case.

## 312 3. Data & Methods

## 313 *3.1. Study Area*

Our study area encompasses the Dry Chaco ecoregion of three provinces: Salta, Santiago 314 315 del Estero, and Chaco, an area of approximately 276,000 km<sup>2</sup>. In the years preceding the 316 Forest Law, these three provinces were responsible for the largest share of forest loss in 317 Argentina. Between 1996 and 2007, hand-digitized GIS datasets (Vallejos et al., 2015) 318 recorded 28,934 km<sup>2</sup> of deforestation in our study region. This corresponds to 79% of 319 the forest loss observed across Argentina's Dry Chaco over the same time period, with 320 the remainder distributed across nine provinces. All three provinces also have significant 321 remaining forest cover and are therefore key localities for avoiding future deforestation. 322 Salta, Santiago del Estero, and Chaco concluded their land use planning by 2009 with the 323 324 adoption of legally binding zoning plans. However, implementation processes varied 325 considerably. Santiago del Estero had been pressured by the federal government to 326 adopt a provincial-level zoning plan since 2004; the province adopted a provincial plan 327 in 2006, which was then translated into national categories by mid-2009 to comply with

328 the Forest Law. Chaco did not feature a provincial plan, but had signaled an intention to

- 329 conduct a planning process via a 2006 provincial decree, which suspended deforestation
- permits until a zoning plan was adopted. In Salta, in turn, the planning process was
- characterized by major controversies. In 2007, while the Forest Law was being discussed
- in the federal legislature, Salta tripled the number of issued deforestation permits as
- compared to previous years, prompting the national Supreme Court of Justice to declare
- a provincial moratorium on deforestation in 2009. Salta concluded its land use planning
- just a few months later (Gobbi, 2015).

# 336 3.2. Unit of Analysis

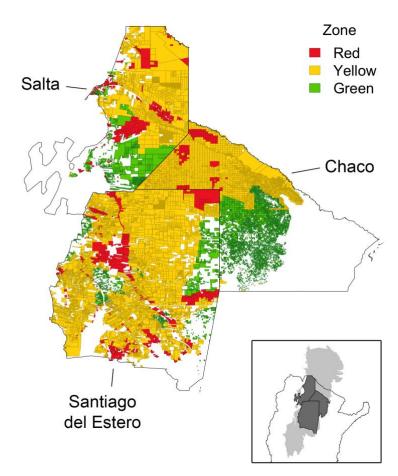
- For each province, we obtained wall-to-wall data of rural property boundaries from
  provincial agencies and universities. Our analysis is based on all rural properties that:
- 339
- 340 1. are situated within the Dry Chaco ecoregion (Olson et al., 2001),
- 341 2. contained more than 20% forest cover at baseline (2007)
- 342 3. are larger than 10ha, as the Forest Law does not apply to smaller properties,
- 343
  4. have an average slope of less than 2.5% (to exclude small patches of
  344
  345
  345
  and
- 346
  5. have a ratio of edge length to area of less than 150m<sup>-1</sup> (which excludes extremely
  347 elongated shapes that appear to be fill-in errors)
- 348
- Our final dataset contains 30,126 properties, of which 3,169 are situated in Salta, 11,196
  in Santiago del Estero, and 15,761 in the province of Chaco (Fig. 1)

# 351 3.3. Variables

- We use the official land use zoning maps of the Forest Law of each province to categorize properties into three groups (red, yellow, or green), ascribing categories as a function of the zone that covers the majority of the property's remaining forest. This categorization of properties is relatively clear-cut: changing the coverage threshold by
- 356 10% in either direction results in only a small number of properties changing groups
- 357 (0.08% 1.6%, mean: 0.45%).
- 358
- We measure deforestation as the percentage of a property's area that was deforested 359 360 within any given time period, using a hand-digitized dataset of observed annual 361 anthropogenic forest conversion in the Dry Chaco (Vallejos et al., 2015). The dataset is 362 based on visual interpretation on LandSat imagery, and infers anthropogenic forest 363 conversion to cropland and pasture from the spatial shapes of converted plots (e.g. 364 regular shapes, hedgerows, etc.), with an overall classification accuracy of 97.8%. The 365 data is also likely to record clearly timed transitions from forests to silvopasture, usually 366 caused by the mechanical removal of tree cover, but unlikely to pick up slow and 367 gradual degradation of forests caused by the mere presence of cattle. We compute
- 368 annual averages of deforestation percentages for a six-year pre-treatment period (2001-

2006), a three-year implementation period (2007-2009) and a five-year post-treatmentperiod (2010-2014).

570 period (2010-201



- 372
- Fig. 1: Schematic map of the 30,126 properties included in this analysis, colored by the
- 374 zone that covers most of the forests within each property. Inset shows study area (dark
- grey) within the Dry Chaco ecoregion (light grey) and Argentina (black outline).376
- We compile data for seven key covariates, i.e. variables that we expect to be associated with the likelihood of deforestation and the likelihood of stricter protection. These
- 379 include indicators of agricultural productivity, accessibility, neighborhood effects, and 380 property type. We use the following indicators and data sources:
- 380 property type. We use the following indicators and data sources:
- Average annual precipitation (mm/yr), a pivotal determinant of agricultural suitability in the Dry Chaco, as estimated by the Tropical Rainfall Measuring Mission (TRMM)
- Percentage of mollisols (%), an important indicator of the potential agricultural productivity of soils, estimated by the National Agricultural Technology Institute (INTA) (Volante et al., 2016)

387 388 389	3.	Cost-distance to towns (in USD), an indicator of accessibility to commodity markets, computed on the basis of existing road networks and estimated average travel costs (unpublished data, María Piquer-Rodríguez)
390	4.	Distance to rivers and wetlands (m, square root transformed), which affects
391		access to water, drainage, and was an important variable in the allocation of
392		protection, adapted from Instituto Geográfico Nacional (2012).
393	5.	Percentage of deforestation that occurred prior to the adoption of the Forest
394		Law (2001 – 2006) within a 50km radius around the center of each property, in
395		order to account for local agglomeration economies and for unobserved factors
396		influencing local variation in deforestation risk (Vallejos et al., 2015).
397	6.	Percentage forest cover at baseline (2007), as shown in the official zoning maps
398		of each province, which followed the methodology laid out in Argentina's official
399		forest inventory (Secretaría de Ambiente y Desarrollo Sustentable, 2005).
400	7.	Size of the property (ha, log transformed), to account for potential differences in
401		the economics of scale at the property level.
402		
403	We ma	ap all raster-based covariates (1-5) to properties as average values using Zonal
404	Statist	ics 0.1 in QGIS 2.18.0.

405 *3.4. Zone Allocation Models* 

To test whether our covariates influenced the allocation of zonation in the three provinces, we develop province-specific multinomial logit models that estimate the likelihood of zone allocation (green, yellow, and red) as a function of all seven covariates. In addition, we develop a pooled multinomial logit model with all covariates and interactions terms (each covariate interacted with each province) to test whether the allocation of zones differed between provinces.

# 412 3.5. Impact Estimation

413 We combine nearest-neighbor covariate matching with a differences-in-differences 414 (DID) estimator to assess the impact of provincial-level land use zonation on property-415 level deforestation. Matching is a quasi-experimental method for causal inference that 416 has witnessed a rapid uptake in the impact evaluation of land use policies over the past 417 eight years (Andam et al., 2008; Joppa and Pfaff, 2011; Nelson and Chomitz, 2011; Nolte 418 and Agrawal, 2013). Matching mimics an experimental setup by identifying groups of 419 control units that are as similar as possible to the units that received a given treatment 420 in terms of covariates, i.e. confounding factors that affect both the outcome and the 421 likelihood of treatment (Ho et al., 2007; Joppa and Pfaff, 2010). However, matching 422 eliminates bias only if confounders are observable and successfully controlled for 423 (Ferraro and Hanauer, 2014). This assumption cannot be tested directly. We therefore 424 use post-matching DID estimators to account for the possible influence of time-invariant 425 unobservable confounders. 426

427 We define treated units as properties whose forests were predominantly situated in a 428 stricter zone (red or yellow), and control units as properties whose forests were 429 predominantly situated in a less strict zone (yellow or green). This results in three 430 pairwise comparisons (red to yellow, yellow to green, red to green) for each province. 431 We conduct matching without replacement (repeated draws of control units) and 432 discard treatment units for which no suitable control parcel can be found within 1 SD of 433 each covariate (calipers). Because matching without replacement is affected by the 434 initial order of treatment units, we repeat our analysis 50 times with randomized order 435 of observations and report average results.

436

437 We use three strategies to verify whether matching successfully reduced selection bias. 438 First, we compute the absolute standardized differences in covariate means between 439 treatment and control groups, averaged across seven covariates, before and after matching. Second, we test for significant differences in pre-treatment deforestation 440 441 rates (2001-2006) between matched treatment and control groups. If matching 442 successfully controlled for selection bias in zoning allocation, this "Placebo" test should 443 reveal no significant differences in pre-treatment deforestation. Third, we test whether 444 outcomes on matched treated and control units followed parallel time trends, a key 445 assumption for the validity of DID estimation. To do so, we group pre-treatment years 446 into two periods of equal length (2001-2003 vs. 2004-2006) and use a DID estimation to 447 test for differences in time trends between groups.

448

449 To estimate the impact of stricter zones on deforestation, we use a DID estimator 450 comparing deforestation rates between pre-treatment (2001-2006) and post-treatment 451 (2010-2014) years on treated properties vs. matched controls. In addition, we test for 452 the possible presence of pre-emptive clearing (higher deforestation triggered by the 453 anticipation of regulation) by estimating group differences in deforestation rates in pre-454 treatment (2001-2006) vs. planning (2007-2009) years. As properties vary considerably 455 in size and given our interest in the effect of zoning on aggregate deforestation, we 456 weigh all tests and impact estimators by property size. All our DID estimators use robust 457 standard errors (Rogers, 1993) and are specified as follows:

- 458
- 459 460

Deforestation =  $\beta_0 + \beta_1$  Treatment +  $\delta_0$  T2 +  $\delta_1$  Treatment • T2

461 Where <u>Deforestation</u> is annual average deforestation within the respective time period, 462 <u>Treatment</u> is a dummy variable for treatment units, <u>T2</u> is a dummy variable for the 463 second time period, and  $\delta_1$  is the DID coefficient of interest.

464

Spatial proximity between properties can lead to leakage of deforestation from more
regulated to less regulated properties, leading to an overestimation of impact. To
ensure our results are not impacted by this effect, we conduct a robustness check of our
analysis, using as controls only properties whose average distance from treatment units
is larger than 10km.

#### 470 3.6. Supplementary Evidence

- 471 To identify plausible mechanisms for the impact of zonation on deforestation rates, we
- 472 use evidence from documents review and key informant interviews. The first three
- authors conducted a total of 13 weeks of field research in the three provinces between
- 474 2013 and 2015, carrying out 122 interviews with government officials, researchers,
- 475 advocacy groups, producers, and commodity traders. In addition, we revise
- 476 implemented zoning plans and the corresponding legal documents.

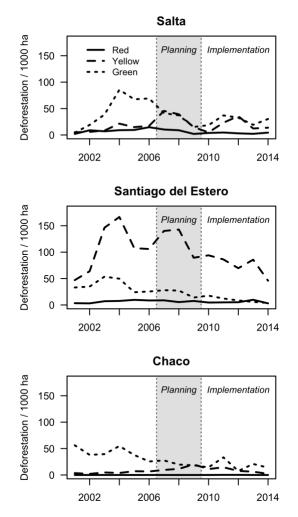
## 477 **4. Results**

## 478 4.1. Aggregate Deforestation Trends

Fig. 2 shows aggregate deforestation trends on properties zoned red, yellow, and green
in the adopted land use plans. Deforestation reached rates of > 150,000ha prior to the
adoption of the Forest Law. In Salta and Chaco, green properties constituted the bulk of
pre-treatment deforestation. Yellow properties accounted for the majority of forest loss
in Santiago del Estero. In all three provinces, deforestation rates followed downward
trends during planning and implementation of zoning.

## 485 4.2. Selection Bias in Zoning Allocation

486 We find strong evidence that Salta, Santiago del Estero and Chaco did not allocate zones 487 randomly across the forests of the Dry Chaco. On average, stricter zones tend to be 488 allocated to land that is less valuable for agricultural production, and thus less likely to 489 be deforested than less strict zones (Fig. 3). In our province-specific multinomial models 490 of zone allocation, 37 out of 42 coefficient estimates (88%) have the expected sign and 491 are estimated to be significant (Table 1). This bias in the allocation of protection is most 492 consistent in the province of Chaco. A pooled multinomial model of zone allocation with 493 province interactions confirms that provinces differ significantly in the determinants of 494 zoning allocation (Table S1). This observation appears to support our working 495 assumption that provinces had leeway in the allocation of zoning.



496

Fig. 2: Aggregate deforestation observed on properties situated in green, yellow, and
red zones by province and year. This data includes properties that had less than 20%
forest cover by 2007 (n = 40,388)

500

Differences in pre-treatment deforestation rates also confirm the presence of selection
 bias. In eight out of nine pairwise comparisons, properties located in stricter zones had

- significantly lower deforestation rates than properties located in less strict zones before
- the Forest Law was passed (Figs. 4-6, Panels C). An interesting exception is Salta, were
- 505 properties later located in red zones exhibited, on average, higher deforestation rates
- 506 before the adoption of the Forest Law than properties later located in yellow zones.
- 507 Indeed, Salta's red-zoned properties are located closer to past deforestation, on
- average, than yellow-zoned properties (Fig. 3), and many are situated in direct
- adjacency to green properties (Fig. 1). As a result, red properties might have had a
- 510 higher potential to reduce deforestation in Salta than in other provinces.

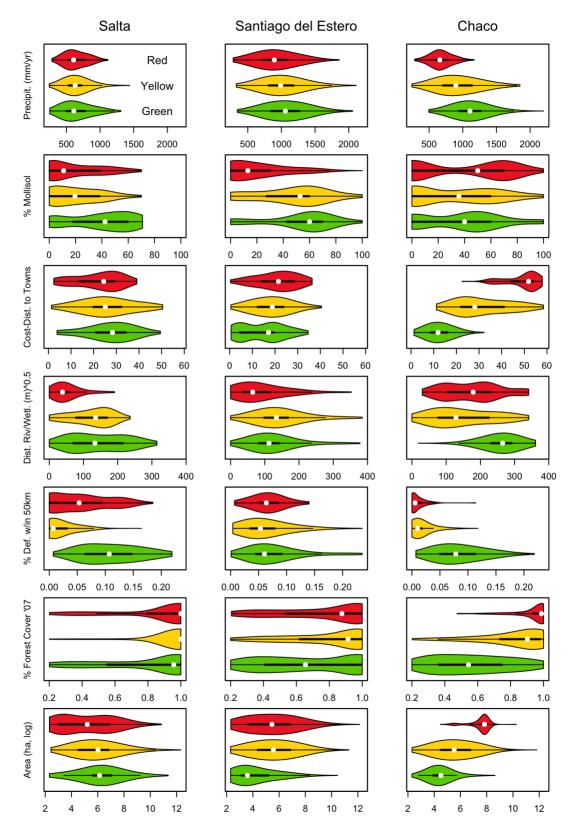




Fig. 3: Kernel density plots of covariate values of 30,126 properties in red (top), yellow 513 (middle) and green (bottom) zones, by province.

Table 1: Coefficient estimates of multinomial logit models of zone allocation, by province. The reference zone is "green". \*\*\*
 p<0.001 \*\* p<0.01 \* p<0.05</li>

	Expected	Sa	lta	Santiago	del Estero	Ch	асо
	Sign	red	yellow	red	yellow	red	yellow
Intercept		5 ***	2.4 ***	-4 ***	-4.3 ***	-19.7 ***	-1.9 ***
Precipitation (mm/yr)	-	-0.00098 *	-0.00275 ***	-0.00099 ***	-0.00018	0.00024	-0.00088 ***
Percentage Mollisol (%)	-	-0.0322 ***	-0.0119 ***	-0.0303 ***	0.005 ***	-0.0287 ***	-0.0202 ***
Cost-Distance to Towns (\$)	+	-0.017 *	-0.039 ***	0.168 ***	0.12 ***	0.309 ***	0.145 ***
Distance to Rivers and Wetlands (m) ^ 0.5	-	-0.02638 ***	-0.00084	-0.01139 ***	0.00405 ***	-0.00973 ***	-0.01722 ***
Deforestation 2001-2006 within 50km	-	-20.1 ***	-62.5 ***	-15.4 ***	-31.6 ***	-64.6***	-53.9 ***
Initial Forest Cover (%)	+	0.82 **	2.77 ***	1.89 ***	2.25 ***	4.26 ***	3.98 ***
Property Size (log ha)	+	0.037	0.474 ***	0.697 ***	0.729 ***	1.4 ***	0.607 ***

#### 517 4.3. Mechanisms

518 Document reviews show that all provinces defined zone-specific legal limits for clear-cut 519 deforestation and silvopastoral use on individual properties. Table 2 illustrates how 520 these percentages vary between provinces. In addition, Salta allowed re-categorization 521 (downgrading) of red zones between 2008 and 2014, possibly reducing the impact of its 522 red zones within the time period of our analysis. All provinces enforced restrictions 523 through a combination of remote detection, field visits, and sanctioning. During our field 524 research, government authorities in all provinces reported increases in enforcement 525 capacity after the adoption of the Forest Law, as exemplified by increased remote 526 detection capabilities using satellite imagery, more frequent field visits, higher levels of sanctions, and an increase in numbers of sanctions issued. These changes did not occur 527 528 abruptly but continuously throughout the planning and implementation period. While 529 an in-depth analysis of enforcement patterns is beyond the scope of this paper, this 530 evidence suggests that observed impacts of zonation are likely a result of increases in 531 regulation and enforcement capacity.

532

Table 2: Percentage of property-level clear-cut deforestation permitted in each zone and province. Numbers in parentheses refer to permitted levels of silvopastoral use.

535

Zone	Salta	Santiago del Estero <sup>a</sup>	Chaco
Red	0%	Z6: 0%	0%
Yellow	0% (100%)	Z5: 0% (30%)	20% (70%)
		Z4: 10% (50%)	
		Z3:	
		<1000ha: 20% (60%)	
		>1000ha: 15% (55%)	
Green	<1000ha: 70% <sup>b</sup>	Z2:	<100ha: 90% (100%)
	>1000ha: 60% <sup>b,c</sup>	<1000ha: 70%	>101ha: 80% (100%)
		>1000ha: 60%	>201ha: 70% (70%)
		Z1:	or 60% (100%)
		<500ha: 95%	>1000ha: 70% (70%)
		>500ha: 90%	or 50% (100%)

<sup>a</sup> Z1 – Z6 refer to zones defined by Santiago del Estero's provincial zoning plan

<sup>b</sup> Percentages are stricter for properties with >5% slope (excluded in this analysis)

<sup>c</sup> At least 700 ha of deforestation is allowed for properties >1000 ha

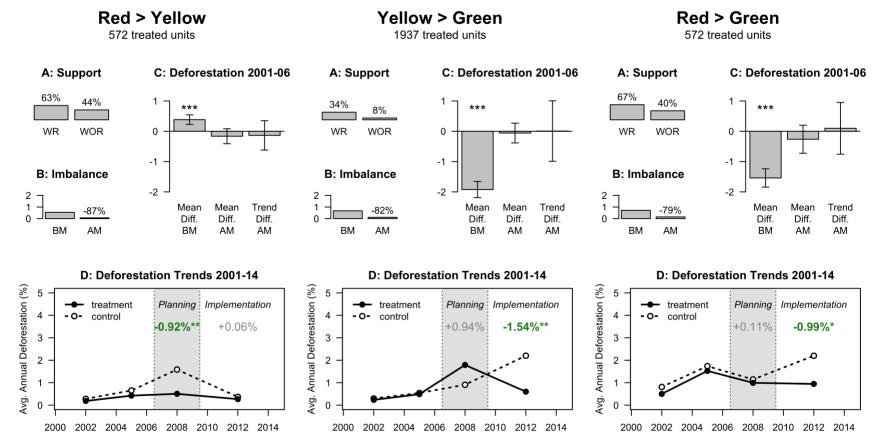
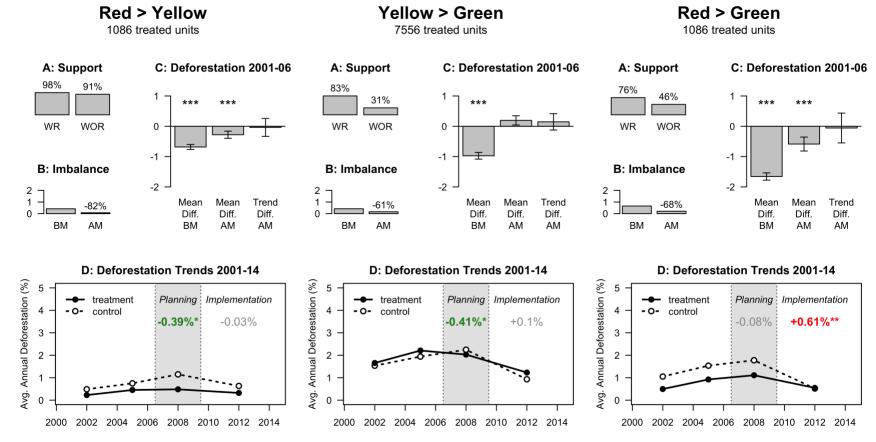


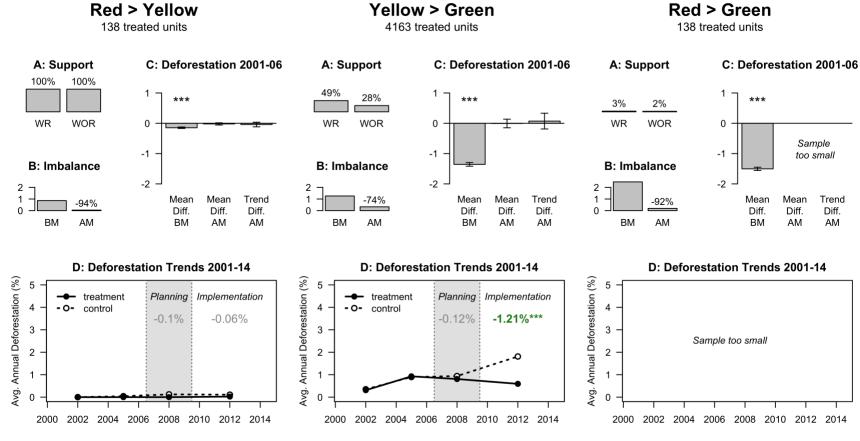


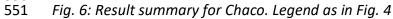
Fig 4: Result summary for Salta. A: Percentage of matched treated units if matching is with (WR) and without (WOR) replacement. B: 541 Average absolute standardized difference between groups in covariate means before (BM) and after (AM) matching. C: Differences in

- 542
- pre-treatment deforestation rates between groups before (BM) and after (AM) matching. "Trend Diff." refers to differences-in-543
- differences (DID) between 2001-03 and 2004-06 (test of parallel time trend assumption). D: Average annual deforestation by period 544
- for treatment and control groups, with reported DID estimates for the planning (2007-2009) and implementation (2010-2014) period. 545
- Significant differences (p<0.05) are highlighted as bold. \*\*\* p<0.001 \*\* p<0.01 \* p<0.05 546
- 547



548549 Fig 5: Result summary for Santiago del Estero. Legend as in Fig. 4





#### 552 4.4. Quality of Matching

553 Due to marked differences in covariate distributions of properties in different zones (cf. 554 Fig 3), not all pairwise comparisons yield sufficiently large matched control groups for a 555 confident assessment of policy impact. In three out of nine comparisons, comparable 556 control units exist for only <50% of treated units (Figs. 4-6, Panels A). In the province of 557 Chaco, were selection bias is strongest, only 3% out of 138 red properties could be 558 matched to green controls. We therefore drop this comparison from the remaining 559 analysis. For the remaining comparisons, suitable control units exist for an average of 560 71% of treated units. It is thus important to keep in mind that our impact estimates are 561 based on comparable sections of the landscape and might not apply to the remainder. 562 563 Matching without replacement produces smaller sample sizes (Figs. 4-6, Panels A),

because pools of suitable controls can become exhausted before all treatment units are
 matched. This effect is particularly strong when comparing yellow to green properties,
 as the former tend to cover larger and more heterogeneous parts of the landscape. It
 provides justification for our strategy to conduct multiple repetitions and report average
 results to ensure that findings are not specific to individual subsamples.

569

570 Matching successfully reduces bias in covariates between treatment and control groups. 571 Across pairwise comparisons, covariate imbalance, measured as the absolute 572 standardized difference in means between treatment and control groups, declines by 573 61-94% post-matching, with an average of 82% (Figs. 4-6, Panels B). Post-matching 574 placebo tests confirm that matching successfully reduces, if not eliminates, selection 575 bias in a majority of cases. While properties in different zones exhibit highly significant 576 differences (p<0.001) in pre-treatment deforestation, these differences become non-577 significant (p>0.01) after matching in Salta and Chaco (Figs. 4-6, Panels C). In Santiago 578 del Estero, significant differences in pre-treatment deforestation remain after matching. 579 This observation points towards the presence of unobserved confounders and provides 580 justification for the adoption of a DID design. The suitability of DID is further supported 581 by test results for parallel time trends: None rejects the hypothesis that matched 582 control and treatment properties followed similar deforestation trajectories before the 583 Forest Law was adopted (Figs. 4-6, Panels C).

# 584 4.5. Impact Estimates: Planning Period

In two out of three provinces, deforestation trends diverged between treatment and
control groups during the implementation process of the Forest Law (2007-2009). In
Salta, DID estimates suggest that properties that were later assigned red zonation
reduced deforestation in 2007-2009 vis-à-vis properties that were later assigned yellow
zonation (Fig. 4, Red > Yellow, Panel D). This appears to be due to a sudden rise in
deforestation on yellow properties in Salta, which is also noticeable in the comparison
of yellow to green properties (Fig. 4, Yellow > Green, Panel D). In contrast, deforestation

- rates in red and yellow zones in Santiago del Estero appear to have declined vis-à-vis
- 593 less strict zones during the implementation process of the Forest Law (Fig. 5, Panel D).
- 594 This might reflect impacts of Santiago del Estero's early provincial-level zoning process, 595 which concluded in 2006.

# 596 4.6. Impact Estimates: Post-Treatment Period

In Salta and Chaco, our estimates suggest that yellow zones significantly reduced deforestation if compared to green zones. In Salta, we find that red zones also significantly reduced deforestation as compared to green zones. Deforestation rates on matched green properties increased noticeably after the adoption of zoning plans (Fig. 4 and 6, Panels D). No significant impact estimates of red vs. yellow zones are found in either province, mostly because matched red properties and their yellow controls experienced low rates of deforestation both before and after the planning period.

In Santiago del Estero, we do not find significant effects of stricter zonation on
 deforestation after 2009 (Fig. 5). On the contrary, our DID estimates suggest that yellow
 zones <u>increased</u> deforestation if compared to green zones after the conclusion of the
 Forest Law planning process. This is true in spite of a noticeable over-time decrease in
 deforestation on yellow properties, as deforestation on green controls decreased even

610 further.

# 611 4.7. Robustness

612 Excluding properties situated nearby treated units (< 10 km) from the pool of potential 613 controls reduces the number of treatment unit with comparable controls by 1% to 40% 614 across runs (average: -19%). This does not affect overall findings in Santiago del Estero 615 or Chaco (Figs. S2 and S3). However, significance estimates change in Salta (Fig. S1), the 616 province most affected by this reduction in the number of comparable treatment units 617 (by an average of -24%, to values as low as 12%): After exclusion of neighbors, red zones 618 in Salta are not estimated to have reduced deforestation if compared to green zones. 619 However, diverging time trends in pre-treatment deforestation indicate that after 620 exclusion of neighbors, treatment and control groups in Salta might not be sufficiently

621 similar to estimate impacts robustly, at least when comparing red to green properties.

# 622 5. Discussion

Did provincial governments in Argentina's Dry Chaco implement land use plans that
inhibited agricultural expansion and reduced deforestation? Our empirical answer is
cautiously affirmative. On one hand, we find evidence that provincial governments
"followed the market" in the allocation of zoning. In all three provinces, stricter zones
were significantly more likely to be allocated to lands of lower agricultural value for beef
and soy expansion, and thus faced lower deforestation risk than less protective zones.

- This phenomenon caps the maximum inhibitory effect these zones can have. It also
  might put forests of high ecological values at risk, if these are situated on land valuable
- 631 for agriculture, such as remnants of <u>quebracho</u> forests (Piquer-Rodriguez et al., 2015).
- 632 We also find that provinces differed significantly in how landscape attributes influenced
- 633 the allocation of zoning, supporting our initial assumption that provinces had some
- 634 leeway in deciding where and how to allocate zones within their jurisdictions.
- 635

636 On the other hand, we find that land use plans adopted by Salta, Santiago del Estero, 637 and Chaco effectively reduced deforestation over counterfactual scenarios, at least in 638 some time periods. These restrictions were effective immediately, with measurable 639 impacts within years after the approval of the land use plans. Across provinces, the 640 inhibitory effect of yellow zones appears particularly consistent. Yellow zones are frequently adjacent to (and thus more comparable to) green zones, where deforestation 641 pressures are higher and restrictions are lower; as a result, yellow zones face higher 642 643 counterfactual deforestation pressure, and can thus have higher potential impact. It is 644 noteworthy that this impact was observed in spite of the significant share of 645 silvopastoral land use that each province permits in yellow zones (Table 2). In contrast, 646 red zones are often located away from deforestation pressure, which attenuates their 647 potential short-term impacts on forest loss. This finding is consistent with other studies 648 that find stricter protection to be more likely allocated to remote areas (Nelson and 649 Chomitz, 2011; Pfaff et al., 2014; but see Nolte et al., 2013b). In Salta, where many red 650 properties are located close to green properties, estimated impacts on deforestation are 651 higher, but might also be more vulnerable to the confounding effects of leakage. 652 Inhibitory effects of red zones might become more important in the future as 653 deforestation frontiers advance.

654

655 Our analysis reveals important dynamics of policy implementation over time. In Salta, we find deforestation on yellow properties to increase during land use planning before 656 657 treatment was assigned. As noted in Section 3.1, Salta had shown little initiative to 658 adopt land use plans prior to the adoption of the Forest Law, and its agencies sped up 659 deforestation permits in anticipation of regulation. Elevated deforestation on yellow 660 properties thus appears to be pre-emptive clearing, i.e. land use conversion to prevent stricter regulation (Seghezzo et al., 2011). In Santiago del Estero, we find stricter zones 661 to have reduced deforestation after the provincial land use plan was adopted (2006), 662 663 but not after its translation into the Forest Law (2009), when red zones are estimated to 664 have increased deforestation if compared to green zones. This counterintuitive finding 665 appears to be driven by a rapid decline in deforestation on green properties that we do 666 not observe in other provinces. Provincial differences in deforestation trajectories and 667 the impact of zoning might also have been influenced by factors that we could not 668 explicitly consider in this analysis, including 1) the dynamics of silvopastoral land use, 669 whose responses to zoning might differ from that of clear-cut deforestation (available 670 data conflates both types of deforestation), and 2) differences in the presence of 671 indigenous and other forest users and their success in resisting commercial agricultural

- expansion, for which comparable data was not available at the time of writing. Future
- 673 research could elucidate these more complex relationships.

## 674 6. Conclusion

675 Subnational governments are increasingly important players in forest-based climate 676 change mitigation, but their ability and willingness to reduce large-scale deforestation is 677 rarely the subject of systematic empirical inquiry. For the Argentinian Dry Chaco, one of 678 the planet's most active deforestation frontiers, we show that the three provinces with 679 the highest historical levels of forest loss implemented zoning plans that reduced 680 deforestation in a counterfactual scenario, at least in locations of the landscape that 681 allow causal inference. While we also document a systematic bias in the allocation of 682 protection to remote and unthreatened locations, this bias does not appear to void the 683 forest conservation impacts of the adopted land use plans.

684

685 Can our findings serve as an indicator of the willingness and ability of subnational 686 governments to reduce large-scale deforestation? As constitutional decision-making 687 powers to land and forests rest with Argentina's provinces, we expect land use policies 688 and resulting impacts in the Dry Chaco to be influenced by provincial priorities. 689 Differences in the definition, allocation and implementation of zones in the Argentina's 690 Dry Chaco corroborate our assumption that Salta, Santiago del Estero, and Chaco had 691 considerable freedom to define and allocate zones in ways that served their interests. 692 However, we also observe instances in which Argentina's federal government exercised 693 power to ensure that provinces implemented land use zoning. We also cannot discard 694 the possibility that the federal government influenced the allocation of zones towards 695 stricter protection in ways we do not observe. Neither do we explicitly consider the role 696 of indigenous and local forest users in resisting the expansion of commercial agriculture 697 in the region. We therefore propose to interpret our results as what they are: rigorous 698 empirical evidence that large-scale deforestation in major agricultural frontiers can be 699 slowed down by subnational policy within a national framework that prescribes 700 processes, but not outcomes.

701

702 Reducing global emissions from deforestation and forest degradation will require the 703 combined efforts of actors at multiple governance levels, from the local to the 704 international. Within international arenas, national governments will remain pivotal 705 decision makers and drivers of forest conservation policies. Nonetheless, the 706 engagement of subnational, local, and private actors in forest conservation warrants 707 more attention. Our observation that provinces with high historical deforestation rates 708 can effectively reduce forest loss if prompted to do so lends support to the proponents 709 of jurisdictional, multilevel and nested approaches to avoided deforestation policy. We 710 encourage further rigorous empirical research on the impacts of multilevel forest 711 conservation efforts in the world's deforestation frontiers to build a stronger evidence 712 base for effective and equitable climate governance.

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## **Supplementary Material**

Table S1: Coefficient estimates of pooled multinomial logit models of zone allocation with province interactions. The reference zone is "green". The reference province is Chaco. \*\*\* p<0.001 \*\* p<0.01 \* p<0.05

		red			yellow	
		Interactions w/ Salta	Interactions w/ Santiago d. Est.		Interactions w/ Salta	Interactions w/ Santiago d. Est.
Intercept / Province Dummy	-19.7 ***	24.7 ***	15.7 ***	-1.9 * * *	4.3 ***	-2.4 ***
Precipitation (mm/yr)	0.0002	-0.0012 *	-0.0012 *	-0.0009 ***	-0.0019 ***	0.0007 ***
Percentage Mollisol (%)	-0.0287 ***	-0.0035	-0.0015	-0.0202 ***	0.0083*	0.0252 ***
Cost-Distance to Towns (\$)	0.309 ***	-0.327 ***	-0.141 ***	0.145 ***	-0.184 ***	-0.025 ***
Distance to Rivers and Wetlands (m) ^ 0.5	-0.0097 ***	-0.0167 ***	-0.0017	-0.0172 ***	0.0164 ***	0.0213 ***
Deforestation 2001-2006 within 50km	-64.6***	44.4 ***	49.1 ***	-53.9***	-8.6***	22.3 ***
Initial Forest Cover (%)	4.3***	-3.4 ***	-2.4 ***	4 ***	-1.2 * * *	-1.7 ***
Property Size (log ha)	1.4 ***	-1.36 * * *	-0.7 ***	0.61 ***	-0.13*	0.12 **

Table S2: Results of difference-in-difference estimations in Salta. Predicted variables are annual average deforestation rates observed on matched treatment and control properties in the time periods of interest. T1 and T2 indicate the first and second time period in each comparison, respectively. The reference time period is T1. All values are averaged across 50 repetitions (see main manuscript, section 3.5). Significance: \*\*\* p<0.001 \* p<0.01 \* p<0.05

	Time Trend		Planning		Implementati	on
	T1: 2001-03		T1: 2001-06	,	T1: 2001-06	
	T2: 2004-06		T2: 2007-09		T2: 2010-14	
Red > Yellow						
Intercept	0.2815	*	0.4678	***	0.4678	***
T2	0.3728		1.1172	***	-0.0992	
Treatment	-0.095		-0.1626		-0.1626	
Treatment * T2	-0.1351		-0.917	**	0.0621	
R-Squared	0.008		0.031		0.003	
Observations	997		997		997	
Yellow > Green						
Intercept	0.2968		0.4193	*	0.4193	*
T2	0.2449		0.4871		1.7837	***
Treatment	-0.0616		-0.0578		-0.0578	
Treatment * T2	0.0076		0.9422		-1.5447	**
R-Squared	0.006		0.031		0.096	
Observations	609		609		609	
Red > Green						
Intercept	0.8105	***	1.2739	***	1.2739	***
T2	0.9269	**	-0.1279		0.9248	*
Treatment	-0.311		-0.2618		-0.2618	
Treatment * T2	0.0984		0.1093		-0.9921	*
R-Squared	0.023		0.002		0.028	
Observations	905		905		905	

	Time Trend		Planning		Implementation	
	T1: 2001-03		T1: 2001-06		T1: 2001-06	
	T2: 2004-06		T2: 2007-09		T2: 2010-14	
Red > Yellow						
Intercept	0.486	***	0.6178	***	0.6178	***
T2	0.2637	*	0.5314	***	0.0151	
Treatment	-0.2597	**	-0.2781	***	-0.2781	***
Treatment * T2	-0.0368		-0.3887	*	-0.0337	
R-Squared	0.013		0.022		0.013	
Observations	3933		3933		3933	
Yellow > Green						
Intercept	1.5338	***	1.7369	* * *	1.7369	***
Т2	0.4064	***	0.5091	***	-0.8063	***
Treatment	0.122		0.1953		0.1953	
Treatment * T2	0.1467		-0.4109	*	0.1054	
R-Squared	0.007		0.003		0.026	
Observations	9235		9235		9235	
Red > Green						
Intercept	1.052	***	1.2941	***	1.2941	***
T2	0.4841	*	0.4804	*	-0.7738	***
Treatment	-0.557	***	-0.5856	***	-0.5856	***
Treatment * T2	-0.0572		-0.0799		0.6121	**
R-Squared	0.023		0.023		0.033	
Observations	1993		1993		1993	

Table S3: Results of difference-in-difference estimations in Santiago del Estero. Legend as in Table S2.

Intercept         0.0005         0.0213         0.0213           T2         0.0415         0.1007         0.0893           Treatment         -0.0005         -0.0213         -0.0213           Treatment * T2         -0.0415         -0.1002         -0.064           R-Squared         0.008         0.016         0.024           Observations         551         551         551           Yellow > Green		Time Trend		Planning		Implementati	on
Red > Yellow         Intercept       0.0005       0.0213       0.0213         T2       0.0415       0.1007       0.0893         Treatment       -0.0005       -0.0213       -0.0213         Treatment * T2       -0.0415       -0.1002       -0.064         R-Squared       0.008       0.016       0.024         Observations       551       551       551         Yellow > Green       0.5458       ***       0.6236       ***       0.6236         Yellow > Green       0.5458       ***       0.3155       **       1.187         Treatment       -0.0421       -0.006       -0.006       -0.006         Treatment       -0.0421       -0.006       -0.006         Treatment * T2       0.0723       -0.1249       -1.2115         R-Squared       0.015       0.003       0.037		T1: 2001-03		T1: 2001-06	5	T1: 2001-06	
T20.04150.10070.0893Treatment-0.0005-0.0213-0.0213Treatment * T2-0.0415-0.1002-0.064R-Squared0.0080.0160.024Observations551551551Yellow > GreenIntercept0.3507 ***0.6236 ***0.5458***0.3155 **1.187Treatment-0.0421-0.006-0.006Treatment * T20.0723-0.1249-1.2115R-Squared0.0150.0030.037		T2: 2004-06		T2: 2007-09	)	T2: 2010-14	
T20.04150.10070.0893Treatment-0.0005-0.0213-0.0213Treatment * T2-0.0415-0.1002-0.064R-Squared0.0080.0160.024Observations551551551Yellow > GreenIntercept0.3507 ***0.6236 ***0.5458***0.3155 **1.187Treatment-0.0421-0.006-0.006Treatment * T20.0723-0.1249-1.2115R-Squared0.0150.0030.037	Red > Yellow						
Treatment       -0.0005       -0.0213       -0.0213         Treatment * T2       -0.0415       -0.1002       -0.064         R-Squared       0.008       0.016       0.024         Observations       551       551       551         Yellow > Green	ntercept	0.0005		0.0213		0.0213	
Treatment * T2       -0.0415       -0.1002       -0.064         R-Squared       0.008       0.016       0.024         Observations       551       551       551         Yellow > Green       0.3507       ***       0.6236       ***       0.6236         Treatment       0.5458       ***       0.3155       **       1.187         Treatment       -0.0421       -0.006       -0.006         Treatment * T2       0.0723       -0.1249       -1.2115         R-Squared       0.015       0.003       0.037	Г2	0.0415		0.1007		0.0893	
R-Squared       0.008       0.016       0.024         Observations       551       551       551         Yellow > Green	Freatment	-0.0005		-0.0213		-0.0213	
Observations         551         551         551           Yellow > Green	Treatment * T2	-0.0415		-0.1002		-0.064	
Yellow > GreenIntercept0.3507 ***0.6236 ***0.6236T20.5458 ***0.3155 **1.187Treatment-0.0421-0.006-0.006Treatment * T20.0723-0.1249-1.2115R-Squared0.0150.0030.037	R-Squared	0.008		0.016		0.024	
Intercept0.3507***0.6236***0.6236T20.5458***0.3155**1.187Treatment-0.0421-0.006-0.006Treatment * T20.0723-0.1249-1.2115R-Squared0.0150.0030.037	Observations	551		551		551	
T2       0.5458       ***       0.3155       **       1.187         Treatment       -0.0421       -0.006       -0.006         Treatment * T2       0.0723       -0.1249       -1.2115         R-Squared       0.015       0.003       0.037	Yellow > Green						
Tz         0.3438         0.3133         1.187           Treatment         -0.0421         -0.006         -0.006           Treatment * T2         0.0723         -0.1249         -1.2115           R-Squared         0.015         0.003         0.037	ntercept	0.3507	***	0.6236	***	0.6236	***
Treatment * T20.0723-0.1249-1.2115R-Squared0.0150.0030.037	Г2	0.5458	***	0.3155	**	1.187	***
R-Squared 0.015 0.003 0.037	Freatment	-0.0421		-0.006		-0.006	
1	Freatment * T2	0.0723		-0.1249		-1.2115	***
	R-Squared	0.015		0.003		0.037	
Observations 4580 4580 4580	Observations	4580		4580		4580	

Table S4: Results of difference-in-difference estimations in Chaco. Legend as in Table S2.

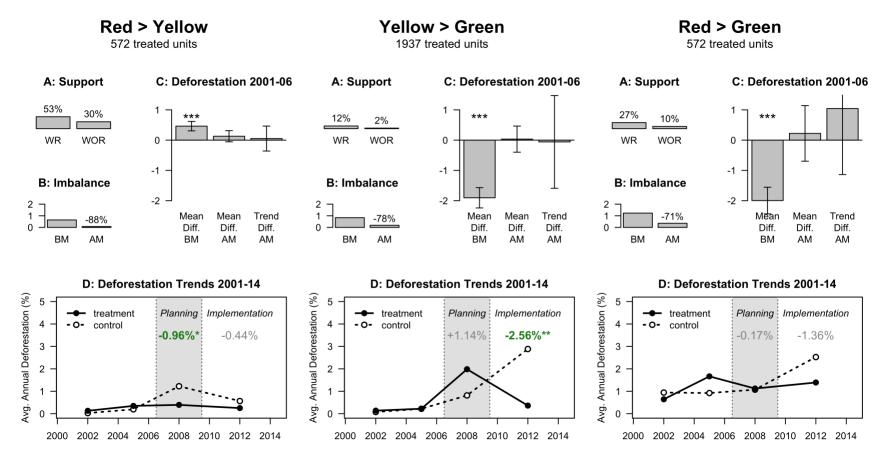


Fig. S1: Results summary for Salta, after exclusion of controls located nearby treatment units (<10km). Legend as in Fig. 4

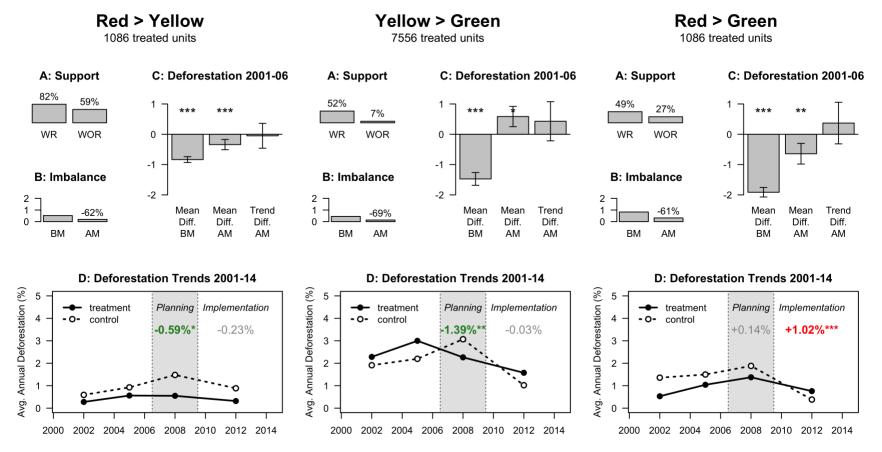


Fig. S2: Results summary for Santiago del Estero, after exclusion of controls located nearby treatment units (<10km). Legend as in Fig. 4

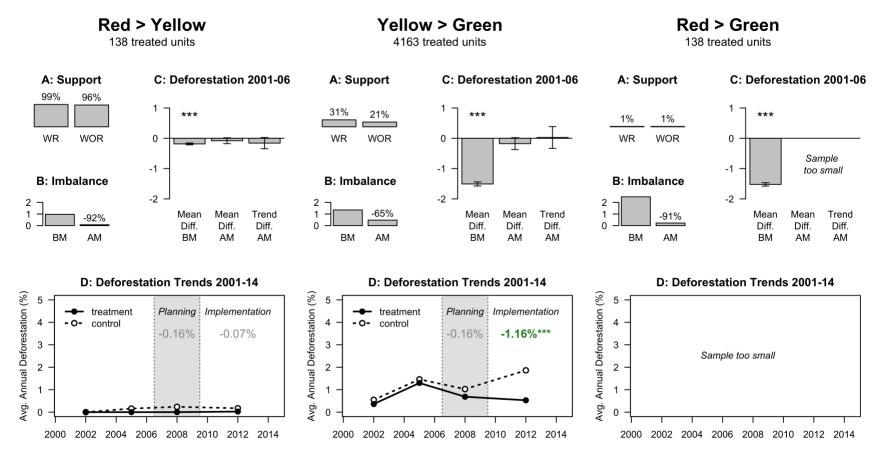


Fig. S3: Results summary for Chaco, after exclusion of controls located nearby treatment units (<10km). Legend as in Fig. 4