

Past cover modulates the intense and spatially structured natural regeneration of woody vegetation in a pastureland

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Abstract Vegetation natural regeneration after agricultural abandonment is changing the landscape patterns in many areas worldwide. However, the expansion rate, spatio-temporal dynamics, and the role of past vegetation cover in shaping such patterns are still barely quantified in fine and meso scales. Here, we aim to quantify the expansion rate and assess the spatio-temporal patterns and the effects of past cover on natural woody vegetation cover increase. We sampled woodland and shrubland cover from 1992 to 2016 in 30 ha in a formerly managed pastureland in Sicily, Italy. We combined field sampling, GIS tools, and spatial analysis to assess the spatial structure

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Departamento de Biología de Organismos y Sistemas – UMIB, University of Oviedo, Oviedo, Spain dynamics and test the effects of past cover amount and type and distance from forest or nearest woody patch on the proportional expansion of natural regeneration. After 24 years, woody cover increased 68%, despite the aggregated spatial structure in 1992 remaining almost unchanged in 2016. The past vegetation cover was the best predictor of woody vegetation expansion in two out of three plots. Distance to continuous forest and to the nearest woody patch, as well as cover type, was not relevant. Our study highlights the importance of fine- and meso-scale studies to reveal both the deterministic and stochastic facet of woody vegetation dynamics. Natural regeneration may strongly change landscape patterns even under constant herbivory pressure and long-term deforestation. The detection of cold and hotspots of regeneration provide an important prompt for the design of restoration programs and landscape management.

Keywords Landscape pattern · Natural regeneration · Scale · Spatial regression · Vegetation dynamics

Introduction

In the Mediterranean region, human land use has altered the cover, composition and spatial patterns of natural vegetation over millennia, in many cases favoring annual and perennial herbaceous vegetation over woody vegetation (Pausas and Millán 2019; Plieninger et al. 2014). However, due to the current trend of land abandonment in rural areas, the natural recovery of woody vegetation on deforested land through secondary succession is highly expected, although frequently influenced by concomitant changes in major disturbance regimes, such as fire and herbivore pressure (Amici et al. 2013; Falcucci et al. 2007; Massa and La Mantia 2007). A process of succession such as this has long been described in floristic and phytosociological terms (e.g., Debussche et al. 1982; Ne'eman and Izhaki 1996), but the intrinsic spatial dynamics of woody vegetation over time and the mechanisms underpinning those dynamics are still poorly understood (Allen et al. 2016; Méndez et al. 2008; Quero et al. 2011). Even basic questions, such as how intense in magnitude and how fast woody vegetation recovery can be, still remain unsatisfactorily answered for mesic woodlands (but see Abadie et al. 2018; Álvarez-Martínez et al. 2014). Studies have focused mainly on analysis in extreme environments, such as semi-arid or mountain areas (Alodos et al. 2004; Bonet and Pausas 2004; Martínez-Duro et al. 2010; Pueyo and Begueria 2007) or on herbaceous or shrub communities (Bashan and Bar-Massada 2017; Carmel and Kadmon 1999; La Mantia et al. 2008; Ne'eman and Izhaki 1996). Similarly, little attention has been devoted to the spatial structure acquired by woody vegetation during succession, whether aggregated or random (Carmel et al. 2001). This is a relevant gap in knowledge given that spatial patterns result from, and, therefore, may be informative of, cumulative ecological processes involving both responses to abiotic conditions and, especially, interspecific interactions (Carmel and Flather 2004; Garcia et al. 2011; Keitt 2003). This spatio-temporal pattern analysis has a number of applications in landscape management. For example, the identification of those areas with greater potential for passive restoration (i.e., vegetation expansion through natural regeneration) may optimize resource use for active restoration (e.g., seedling plantation), directing efforts towards areas with lower recovery potential or regeneration cold spots (Berdugo et al. 2017; Carmel et al. 2001; Holl et al. 2018).

Abiotic factors such as climate, topography and soil may influence vegetation expansion and induce nonrandom spatial structures even on a small scale (Gallego Fernández et al. 2004; García et al. 2014; Pueyo and Begueria 2007), although abiotic influence on some tolerant species may be less significant (Bacaro et al. 2008; Getzin et al. 2008; Quero et al. 2011). In such cases, intensity and type of past land use, the amount and distribution of past vegetation cover, and biotic mechanisms become the main interrelated drivers of vegetation expansion, defining spatial resilience in a determined site (Allen et al. 2016; Carmel and Flather 2004; Holl et al. 2018). Vegetative growth (e.g., canopy expansion and clonal reproduction), and facilitation (e.g., when a small plant enables the recruitment of a larger one) are inherently dependent upon past cover (Bakker et al. 2004; Garcia and Obeso 2003). In turn, the process of seed dispersal in woody plants, which in many ecosystems such as tropical forests or in the Mediterranean is highly dependent upon animals (i.e., zoochory), may shape the spatial structure of woody recovery in different ways. On the one hand, most seeds are deposited at short distances from their source, thus, most recruitment and vegetation expansion is expected near source (González-Varo et al. 2017; Martínez and García 2015). On the other hand, depending on the degree of redundancy or complementarity of the dispersal vectors, dispersed seeds may eventually reach greater distances but might be conditioned by the existing woody cover in different levels (Bueno et al. 2013; Garcia et al. 2011; Jordano 2017). Notwithstanding the above, the spatial template generated by seed dispersal is often subject to subsequent modification by recruitment losses due to small-scale disturbances, such as seed predation and herbivory, or large-scale disturbances, such as fire (Baeza et al. 2007; Carmel and Kadmon 1999; Gómez-Aparicio 2008). Cover type (i.e., shrubland or woodland), therefore, may affect seed dispersal and recruitment. Shrubs may attract seed dispersers and seed predators, and may compete with or facilitate other plants differently than higher trees, the same being true for patches or isolated individuals (Andivia et al. 2017; Lasky and Keitt 2012; Olff et al. 1999). Consequently, studies that quantify expansion rates and spatial patterns of woody vegetation over time and relate them to pre-existing vegetation cover are necessary in order to reveal the footprints that the ecological processes driving secondary succession leave behind (Chuang et al. 2018; García et al. 2014; Holl et al. 2018). Furthermore, these studies may

contribute to disentangle the role of spatial effects (e.g., spatial autocorrelation and random dispersal) affecting vegetation expansion and dynamics, and can enhance the effectiveness of management plans of protected areas and the recovery of degraded sites (Hu et al. 2012; Méndez et al. 2008; Quero et al. 2011).

Here we combined current field sampling, GISbased interpretation of sequential, high-resolution images (satellite and aerial photographs) and spatially explicit analysis to evaluate magnitude, spatial patterns and potential drivers of expansion (i.e., increase) relating to woody vegetation cover over 24 years in a Mediterranean woodland pasture. We sought to answer the following questions: How fast, in terms of the magnitude of expansion over time, does woody vegetation recover in pastureland? How dynamic is the spatial structure along such expansion? Are current cover and expansion spatially correlated with the amount of past woody vegetation cover? And to what extent might distance to the continuous forest and to the nearest woody patch affect cover expansion? Our study presents an appropriate system in order to seek answers to those questions as it combines relatively homogeneous abiotic conditions with an absence of fire, providing a clearer quantification of woody vegetation recovery potential and interpretation of vegetation dynamics due to major biotic processes, such as zoochory and herbivory.

Methods

Study site

The study was conducted at Alpe Cucco (37° 52′ and 13° 24′), a site located in the center of the "Bosco della Ficuzza, Rocca Busambra, Bosco del Cappelliere e Gorgo del Drago" nature reserve (hereafter FBCD), which is a 7397 ha protected area composing the last, large remnant of forest in western Sicily (Badalamenti et al. 2017; Gianguzzi and La Mantia 2004). The site is a 160 ha woodland pasture comprising seasonal pastures intermingled with woody vegetation, surrounded by forest patches (Fig. 1). The local climate is mesic Mediterranean, with an average rainfall of 850 mm concentrated mostly in autumn and winter, and mean annual temperatures of 14.3 °C, with average temperatures of 9.4 °C in the coldest months

(January and February) increasing to an average of 23.5 °C in the hottest months (July and August).

Vegetation in Alpe Cucco corresponds to the Meso-Mediterranean type, based on a classification of altitudinal belts (Rivas-Martínez 2008). Surrounding forest patches are mostly composed of holm oak (Quercus ilex), downy oak (Quercus pubescens) and, to a lesser extent, maple (Acer campestre) and manna ash (Fraxinus ornus) (Badalamenti et al. 2017; Gianguzzi and La Mantia 2004). The woodland pasture is dominated by seasonal grasses and forbs (see Bianchetto et al. 2015 for a species list) with isolated shrubs and trees, and varied-sized patches of woody vegetation comprising mostly six fleshyfruited species (Pyrus amygdaliformis, Crataegus monogyna, Crataegus laciniata, Rubus ulmifolius, Rosa canina and Prunus spinosa) and, to a lesser degree, one dry-fruited shrub (*Calicotome infesta*), the two oaks and maple (Gianguzzi and La Mantia 2004). The site was subject to a reforestation program during the decade 1970-1980; individuals of Fraxinus angustifolia and Pinus halepensis, two species previously absent in the pasture, were planted in scattered patches and isolated individuals in a part of the Alpe Cucco area (Gianguzzi and La Mantia 2004). Extensive cattle (cow) raising has a long history in Alpe Cucco, and is still present nowadays, whereas an active pastureland management program including irrigation, plowing, and seeding was developed during the period 1960–1990. (A.S.F.D 1959; Bianchetto et al. 2015).

Sampling design

In September 2016, we established three 500×200 m plots, each divided into 160 subplots of 25×25 m (sampling unit) covering a gradient from continuous forest to woodland pasture (Fig. 1). The minimum and maximum distances between plot edges were 300 m and 510 m (700 m from the centroid). All plots extended over relatively homogeneous flat areas (average slope 15%) with northern exposure. Plot 1 has an average altitude of 975 m $(\pm 17 \text{ SD})$, plot 2—1004 m (± 36) and plot 3— 901 m (\pm 9), with maximum altitudinal difference between subplots of 190 m (880 to 1070 m a.s.l.) All plots are classified as having deep soils (> 100 cm), sub-alkaline, clay-dominated vertic haploxeralfs (Raimondi 1983) and share mostly the same plant species.



Fig. 1 a Location of the Ficuzza Reserve in Sicily (black spot). b Google Earth 2016 image showing the Ficuzza forest remnant and location of Alpe Cucco inside the reserve (white square). c Google Earth 2016 image with the three 500×200 m study

For example, the two most abundant tree species, *P. amygdaliformis* and *C. monogyna* have established individuals scattered across 76% and 55% of the subplots (Bianchetto et al. 2015; Bueno unpub. data). Consequently, we assume that our plots are rather uniform regarding key environmental features such as precipitation, exposition, soil and temperature.

Vegetation mapping and classification

To measure the spatial pattern of woody vegetation cover over the years, we contrasted aerial photographies from the year 1992 (scale 1:29,000, 1 pixel = 1 m) and satellite images from 2016 (Google Earth, RGB bands, 1 pixel = 0.13 m, rescaled to 1 m), in a time lapse of 24 years, incorporated into a Geographical Information System (GIS) platform belonging to our study system (see Online Appendix S1 for details of woody cover extraction procedure). Although older images were available, the 1992 images were the

plots established in the Alpe Cucco woodland pasture, covering the gradient from continuous forest to pastureland. **d** Example of the 160 (625 m²) subplot grids that was delimited inside each plot over the 2016 vegetation cover (light gray area)

oldest following cessation of pasture management, representing a conservative date to address vegetation recovery through secondary succession without further large-scale, human-provoked disturbance.

We sought to differentiate three types of woody cover in aerial views: woodland patches, shrubland patches, and isolated trees, shrubs and small nucleation patches of trees or shrubs (hereafter ITSN). In the 2016 extracted cover, woodland patches $(> 50 \text{ m}^2)$ were classified by contrasting with the original high-resolution image as well as by the heterogeneous dark gray texture which produced conspicuous black shadows (field checked as > 2 min height) (see Online Appendix S1 and S2 for details). The species composing this vegetation type were Quercus pubescens, Quercus ilex, Acer campestris, Pyrus amygdaliformis, Crataegus monogyna, Crataegus laciniata, together with Fraxinus angustifolia and Pinus halepensis, two species naturally absent from the study area and that dominate the canopy cover wherever planted. Shrubland patches were defined as with an area of $> 50 \text{ m}^2$, homogeneous texture with a lighter-gray pattern compared to woodlands and which produced no conspicuous shadow (< 2 m tall), with or without trees. The species composing this cover type were Rubus ulmifolius, Prunus spinosa, Rosa canina and Calicotome infesta together with juveniles of P. amygdaliformis and C. monogyna. The remaining woody cover with a surface of $< 50 \text{ m}^2$ was classified as ITSN. All the woody vegetation cover extracted from the images was added as a new layer (polygons) to the GIS platform (Online Appendix S2). As the digital pixel value did not enable us to differentiate clearly between woodland and shrubland boundaries which occurred within the same polygon, we divided the 2016 and 1992 woody cover using photointerpretation, separating the different cover types into distinct polygons (Online Appendix S2). We had no reasons to correct the woody cover extraction output (i.e., no woody cover signed over the grassland or vice versa), but in order to validate cover-type classification and its boundaries, all polygons were numbered and checked in the field, where minor differences were corrected (accuracy > 98%); the subplot woody cover surface (m^2) of the different vegetation types was subsequently obtained (Online Appendix S2). Continuous forest, mainly comprising Quercus ilex and Quercus pubescens, was defined as a single woodland patch which contained the dense continuous woodland remnant surrounding Alpe Cucco. Distance predictors were calculated from the centroid of each subplot up to the nearest border of the continuous forest and to the nearest woodland patch (Online Appendix S2). According to the oldest aerial image available (1955), most plot surface was already deforested at that time (Online Appendix S3). As we were particularly interested in the expansion of woody cover due to natural regeneration, we differentiated the cover attributable to natural processes from that attributable to human-made plantation (see Pueyo and Begueria 2007 for a similar procedure; Appendix S2). Woody cover extraction, classification and figures were done using software QGIS (QGIS Development Team 2016).

Spatial structure of the woody vegetation cover over time

We were interested in characterize the degree of stability of the spatial structure of woody vegetation cover from 1992 to 2016. For this purpose we used the Spatial Analysis by Distance Index-SADIE (see Perry et al. 1999). SADIE uses a transportation algorithm to estimate the distance to regularity (D), that represents the minimum distances that the values of each spatially located variable (in our case woody cover in each subplot) need to move to reach a regular spatial distribution across all subplots, based on the concept of donors and receivers (Perry et al. 1999). The ratio between D and its average values, estimated through randomization, generate the global aggregation index (Ia) as well as its significance values at 95% confidence intervals (p). An index around 1 indicate a randomly distributed woody cover, lower than 1 indicate regular distribution and higher than 1 represent an aggregated or patchy pattern. The Ia is composed by specific subplot clustering values (v), showing the magnitude of the contribution of each sampling unit to the overall cluster values and indicating if the cover value of a subplot is member of an above average patch (vi) or if a member of below average gap (v_i) (Perry and Dixon 2002; Perry et al. 1999). We calculated the SADIE aggregation indexes of the 1992 and 2016 woody vegetation cover separately (considering only the natural vegetation in 2016).

SADIE methodology also enabled us to quantify the spatial association index (*Xp*) of woody cover, derived from the mean values of 1992 and 2016 woody cover in the same subplot (i.e., correlation between the *vi* and *vj* indexes of both periods) (Perry and Dixon 2002). The index range from -1 to 1 and is significant if p < 0.05, where a positive association means that the vegetation cover spatial distribution remained stable from 1992 to 2016 (both if a woody patch or a grassland gap), while a negative value means a mismatch of cover values along the 24 years (e.g., a grassland occupied by woody vegetation).

The next step was to exclude the planted cover to obtain values for proportional expansion of the natural regeneration cover from 1992 to 2016 (hereafter PE), calculated with the formula (2016 natural regeneration cover—1992 cover)/(625—1992 cover). We selected such relative variable instead of the amount

of increase to avoid the geometrical constriction of the quadratic relationship between available space and maximum cover increase. Thus, our PE represents the magnitude of cover expansion within each subplot in each 1 m^2 unit, enabling the comparison of plots with different past covers independently from the absolute quantity of past cover within the subplot,. We then performed another SADIE analysis to obtain the spatial structure (Ia, vi and vj) of PE. For this SADIE analysis, we excluded those subplots with full woody vegetation cover in 1992 and, consequently, zero expansion due to lack of space; there then remained 155 subplots in plot 1, 160 in plot 2, and 152 in plot 3. SADIE indexes are continuous variables (Perry and Dixon 2002), thus we incorporated the vi and vi indexes of 1992 and 2016 cover, and those of PE into bi-dimensional contour maps to obtain a more comprehensive view of the spatial pattern structure and to visualize coldspots (i.e., areas with zero or low cover increase) and hotspots (i.e., areas with high cover increase). Sadie aggregation indexes were obtained with SadieShell 2.0 and the spatial association was calculated with N_AShell 1.0 (Perry et al. 1999). Contour maps were prepared using the plugin contour of the software QGIS (QGIS Development Team 2016).

Factors affecting the expansion of woody vegetation cover

We sought to assess whether the quantity, identity and spatial configuration of past woody vegetation cover affected its further expansion over the 24 years period of study. Thus, we used multiple regression models, including the values of PE at the subplot level as a response variable and the amount of woody vegetation cover, distance to the continuous forest and distance to the nearest woodland in 1992 as continuous predictor variables. We also incorporated, as a qualitative predictor, the cover type in the 1992 subplot, classified as 0 if there was no cover, 1 for ITSN, 2 for shrubland, 3 for mixed cover (2 or more types) and 4 for woodlands). Cover type 1, 2 and 4 were defined if reaching > 70%, otherwise was mixed. We excluded the subplots with full cover in 1992 and those with negative PE values (i.e. ,entreatment of woody vegetation cover) from the analysis, resulting in 145 subplots in plot 1, 152 in plot 2 and 152 in plot 3. PE was square-root transformed to achieve normality and reduce heteroscedasticity. Before running regressions we tested for multicollinearity among predictors using the variance inflation factor (mean = 2.73, highest = 4.11) and the three continuous predictors (amount of woody vegetation cover, distance to the continuous forest and distance to the nearest woodland in 1992) were standardized prior to the analysis. The model was calibrated and validated by correlating the measured and predicted PE values across all plots, with p < 0.001 in all cases (Online Appendix S5).

As sampling included the use of spatially contiguous subplots, and vegetation expansion may present a contagious pattern (e.g., Bakker et al. 2004), we tested for the presence of spatial autocorrelation (Dormann et al. 2007; Keitt et al. 2002). Initially, we performed a global (all plots) and independent (each plot) nonspatial ordinary least squares (OLS) multiple regressions. Once all OLS models presented strong spatial autocorrelation (Moran's I p < 0.001; Online aApendix S5), and this issue can cause strong bias in the results (Dormann et al. 2007; Kissling and Carl 2008) we discarded OLS results and tested both Spatial AutoRegressive Lag and Error Models (SARlag or SARerror). These regression analysis takes into account spatial autocorrelation in data through a weight matrix that incorporates the neighborhood values of that location (SARlag) or of the errors (SARerr) (Kissling and Carl 2008; Ver Hoef et al. 2018). In our case, the weight matrix was row standardized and based on different distance classes of neighbor points (subplot centroids), starting at 36 m, which included all the immediate surrounding neighbors (borders and vertex first order) and the consequent orders of neighbors at each 36 m distance class. Due to our sampling design (spatially discontinuous sets of equally spaced grids), the differences in the weight matrix among plots (Online Appendix S5) and the objective to test for meso-scale differences in vegetation expansion, we ran independent SAR models for the different plots. The best model from SARlag and SARerr was selected based on lowest AICc, highest log-likelihood values, minimum residual spatial autocorrelation and model fit (pseudo- R^2). All the spatial tests and regressions were performed with GEODA 1.8 (Anselin et al. 2006).

Results

From 1992 to 2016, the overall woody vegetation cover in the three plots increased by 6822 ha (90.1%), rising from 7568 to 14,390 ha. Most of the increase was due to natural regeneration (5159 ha), including 2849 ha of woodland, 2164 ha of shrubland and 0146 ha of ITSN (Fig. 2, Online Appendix S4). However, proportionally (i.e., increase relative to initial cover), shrubland almost doubled in surface (94.9%), followed by woodland (67.1%) and ITSN (14.1%). Planted cover accounted for the remaining 24.4% (1663 ha). We found no correlation between 2016 planted cover and woody vegetation cover in 1992 (Pearson r = -0.01, p = 0.69), suggesting that the development of planted cover occurred almost exclusively throughout the 24-year period. The three plots differed in the average proportional expansion of natural regeneration woody cover (ANOVA F = 45.94, p < 0.001) with contrasting variations in each cover-type contribution to expansion (Fig. 2).

Spatial structure of the woody vegetation cover over years

The SADIE *Ia* index indicated that global spatial structure of woody vegetation cover was significantly aggregated and remained stable over time (from 1992 to 2016) (Table 1). Furthermore, the significant Xp index indicated a strong spatial correlation between vi and vj values in the same subplot over time, denoting a conservative, small-scale spatial trend in vegetation dynamics (i.e., few subplots members of a gap became



Fig. 2 Average (\pm SE) subplot values of proportional expansion (PE) of isolated trees, shrubs, and small nucleation (ITSN), shrubland, and woodland patches in plots 1, 2 and 3 from 1992 to 2016

Table 1 Summary of SADIE analysis

Plot	Ia		Хр	Ia	
	1992	2016	1992 × 2016	PE	
1	5.69***	5.89***	0.89***	4.25***	
2	4.42***	4.38***	0.65***	3.91***	
3	5.72***	5.19***	0.79***	2.76**	

Index of aggregation (*Ia*) and spatial association index (*Xp*) of the 1992 woody cover and 2016 natural regeneration woody cover, and the index of aggregation of the proportional expansion (PE) of natural regeneration woody cover in the three study plots (***p < 0.001, **p = 0.003)

members of a patch or vice versa) (Table 1). The PE aggregation index also showed positive and significant values across plots, suggesting marked patchiness in the distribution of woody vegetation cover expansion within each plot, with no clear sign of broad, gradient-like spatial structure from the continuous forest towards open areas (Table 1; Fig. 3).

Factors affecting the expansion of woody vegetation cover

The SARerr model showed the best fit in comparison with SARlag models in all plots when assessing the effects of past cover, distance to continuous forest, distance to nearest woody patch and cover type on the proportional expansion of the natural regeneration of woody vegetation (Online Appendix S5) and subsequent results refers to this model. Woody cover in 1992 was the strongest predictor of proportional expansion (PE) across all plots, with a positive and significant effect in plots 1 and 2 (Table 2; Fig. 4). Distance to the continuous forest showed no significant effect on PE in any plot (Table 2). However, PE tended to increase at shorter distances from the nearest woody patch, as illustrated by the negative coefficient values of this predictor across plots (p = 0.07 in plot 1; p < 0.001 in plot 2). PE was also independent of the type of woody vegetation cover in all plots (Table 2).

Discussion

In our study, we quantified the intensity of woody cover expansion and revealed, on the one hand, overall maintenance of the aggregated spatial pattern of **Fig. 3** Contour maps showing the SADIE aggregation indexes of the 1992 woody cover, 2016 natural regeneration woody cover, and the natural regeneration proportional expansion (PE) for the three study plots. The legend shows the specific subplot aggregation index *vi* and *vj* value ranges



Table 2 Summary of theresults of the SARerr modelin the three study plots withthe pseudo- R^2 showingmodel variance explanationof the proportionalexpansion of naturalregeneration woody cover

	Predictor	Coefficient	Standard error	z	р
Plot 1	Constant	0.337	0.031	10.898	< 0.001
$R^2 = 0.61$	Past woody cover	0.124	0.028	4.479	< 0.001
	Distance continuous	- 0.004	0.030	- 0.139	0.889
	Distance woody patch	- 0.051	0.028	- 1.825	0.068
	Cover type	0.019	0.016	1.142	0.254
	Lambda	0.345	0.117	2.947	0.003
Plot 2	Constant	0.612	0.067	9.132	< 0.001
$R^2 = 0.63$	Past woody cover	0.077	0.019	4.138	< 0.001
	Distance continuous	- 0.037	0.048	-0.780	0.436
	Distance woody patch	- 0.068	0.020	- 3.376	0.001
	Cover type	-0.008	0.016	- 0.516	0.606
	Lambda	0.800	0.082	9.708	< 0.001
Plot 3	Constant	0.485	0.053	9.080	< 0.001
$R^2 = 0.24$	Past woody cover	0.054	0.039	1.376	0.169
	Distance continuous	-0.008	0.037	- 0.209	0.835
	Distance woody patch	- 0.010	0.032	- 0.314	0.753
	Cover type	0.018	0.023	0.761	0.447
	Lambda	0.445	0.109	4.101	< 0.001



Fig. 4 The relationship between 1992 past woody cover (m^2) and proportional expansion of natural regeneration woody cover in plots 1, 2, and 3 expressed in a simple linear regression. Each dot represents a different subplot within the plot

woody vegetation cover over 24 years, even under strong expansion. On the other hand, we detected quantitative (i.e., cover expansion) and qualitative (i.e., cover type composing the expansion) differences even in closely located plots. The increase in natural woody vegetation found in our study would seem to be relatively high, considering the current structural complexity of varied-size woodland and shrubland patches, and the deforestation, anthropic use and herbivory pressure, at least over the last 80 years inside the plots (Bianchetto et al. 2015).

Woody vegetation natural regeneration

It has been demonstrated that livestock herbivory can slow down or even hamper woody vegetation recovery (Carmel and Kadmon 1999; La Mantia et al. 2013; Laskurain et al. 2013). No precise information on cattle density was available for our site, nor regarding the increasing population of fallow deer (Dama dama) (Bianchetto et al. 2015), thus limiting our interpretations. However, judging by the observed recovery rate, no significant negative influence of herbivory in cover expansion seems to be occurring, at least over the past 24 years. A similar outcome was also observed in other Mediterranean areas (e.g., Bashan and Bar-Massada 2017). Besides the magnitude of expansion, the quality of vegetation recovery in our study system also differed from other studies in the Mediterranean region. Bonet and Pausas (2004), for example, found an average woody vegetation cover increase of 26% after 60 years; however, less than 20% of that cover was composed of endozoochoric species, while La Mantia et al. (2008) found an increase of 57%, comprising mostly one dry-fruited species. On our site, the 68% natural regeneration woody cover increase was composed mainly of a mix of six zoochoric species (Pyrus amygdaliformis, Crataegus laciniata Crataegus monogyna, Rubus ulmifolius,

Prunus spinosa, Rosa canina) (R.S.Bueno unpub. data).

Spatial structure of the woody vegetation cover over years

Our SADIE analysis demonstrated that the woody vegetation cover kept its highly spatially aggregated pattern on both a landscape (plot) and a local (subplot) scale throughout the 24 years, even when faced with considerable expansion (Table 1). The maintenance of aggregated patterns is the expected result of expansion from previous cover (Bakker et al. 2004). In fact, past woody cover positively influenced vegetation expansion in plot 1 and 2 (Table 2; Fig. 4). In addition to the effects of past cover, herbivory intensity may be one of the factors not measured here that can influence vegetation spatial configuration, although a range of responses has been found so far (Adler et al. 2001; Bashan and Bar-Massada 2017). Seifan and Kadmon (2006), for example, found that cattle herbivory changed the spatial pattern of a scrub community by reducing the aggregation level (clumpiness), although the cover increase was not affected by different levels of herbivory intensity. In our study, we found that the aggregated pattern remained unchanged, while cover expansion varied among plots sharing the same herbivory pressure (Fig. 2).

Influence of past cover, cover type and distance to remnant vegetation

Grass cover and canopy closure (i.e., a proxy of woody cover) were deemed to be the most relevant predictors in tropical forest recovery (Holl et al. 2018), although differential influences of past cover have been found across Mediterranean-type ecosystems (Carmel and Flather 2004). On our site, we found a differential contribution of past cover at plot scale, once this predictor was found to be non-significant in explaining expansion in plot 3 (Table 2). This plot showed some large patches resulting from the direct colonization of open ground, far from any previous woody vegetation cover, and this would explain the lack of effect of past cover in the regression model, as well as the lowest SADIE aggregation index. Another fact potentially affecting the lack of effect of past cover on expansion is that some types of early woody cover can themselves impede further development of ecological succession (i.e., arrested succession) (Acácio et al. 2007). On our site, for example, a number of *R. ulmifolius* patches remained almost unchanged over the 24 years, and a few large, old trees had virtually no vegetation increase in their surroundings. In the first case, the absence of cover increase in bramble patches may derive from a strong competitive effect (Fotelli et al. 2001). In the case of isolated large trees, the lack of structure for facilitating protection against herbivory, such as basal branching and protective spiny leaves at the ground level (Garcia and Obeso 2003) could explain the above pattern.

Contrary to our expectations, distance from the continuous forest proved non-significant in predicting vegetation expansion. Remnant forests usually act as a source of propagules for the colonization of open areas. Therefore, given the overall distance constraints of seed dispersal, a decreasing rate of vegetation expansion when moving away from the forest is expected (García et al. 2014). Our sampling design, with plots sharing similar configuration of continuous forest (one side forest), should be effective to check for distance effects by forcing a very large distance gradient in sampling points without lateral influence. However, on our study site, the continuous forest comprised mostly oaks (Q. pubescens and Q. ilex), while the majority of the mid-successional woody plants species composing the bulk of cover increase were located mostly in pastureland. For example, one of the most important species in terms of cover expansion (P. amygdaliformis) occurs at very low density inside the forest and is dispersed almost exclusively by mammals (Fedriani et al. 2010); therefore, even the longest distances within plots (ca. 1400 m) can fall within the mammals seed dispersal potential (González-Varo et al. 2017). Complementarily, a recent study in Mediterranean has verified no differences in seed arrival in old fields of species dispersed by both birds and mammals up to 90 m from the seed source (La Mantia et al. 2019). In addition, higher fruit availability in woodland pasture compared to continuous forest coupled with vegetation structural complexity (i.e., presence of different perches and woody patches) may influence the behavior of those birds acting as seed dispersers, fostering movement away from the forest and towards fruit-rich environments (La Mantia and Bueno 2016; Lasky and Keitt 2012). However, later in the process of secondary succession, an increase in the influence of distance to continuous forest in cover species composition is expected, when the mid-successional species composing current expansion facilitate oak establishment in woodland pasture (Alias et al. 2010; Amici et al. 2013).

Distance to woody patch, in turn, was found to negatively affect the expansion of woody vegetation; however, this pattern was only significant in plot 2. A negative influence was expected established that proximity to even small patches of woody vegetation is a proxy for higher seed rain and consequent recruitment (Martínez and García 2015; Navarro-González et al. 2013). The initial configuration of woody vegetation in plot 2, with many scattered, small patches, would have strengthened this negative relationship. The lack of relationship in the other two plots may derive, firstly, from a random distribution of mammal dispersed species (e.g., P. amygdaliformis) at this spatial scale. Secondly, the average distance range between patches may be easily covered by birds, resulting in a cover-cover directed seed dispersal (i.e., from a perch to another perch), with a consequent gap in recruitment between close patches (Garcia et al. 2011; Pausas et al. 2006). Refining the scale of the study using smaller subplots should provide additional information, although a species-specific differential response of each plant is also expected.

Woody cover expansion was independent of the type of patch composing the past cover, suggesting that, in our system, small patches of isolated trees and shrubs have the same potential to promote or prevent expansion as larger patches of shrubland and woodland. Moreover, we also observed subplots free of woody vegetation in 1992 occupied by woodland patches in 2016. For example, most oak-dominated woodland patches showed smaller expansion than mixed shrubland patches, probably an outcome of selective grazing and browsing (Carmel and Kadmon 1999; Garcia and Obeso 2003). On our site, palatable, late successional species (e.g., Quercus sp.) were almost absent from the woody cover under expansion, and mid-successional and herbivore-defended species are actually those developing woodland. Autoregressive parameters (lambda) obtained in our models, however, indicate that other factors not measured in our analysis are influencing cover expansion, a similar outcome was found also for species distribution (Hu et al. 2012). This strong neutral effect corroborates the importance of taking autocorrelation into consideration in spatial analysis (Ver Hoef et al. 2018).

Large-scale approaches are necessary in order to detect landscape and global patterns (Sluiter and de Jong 2007). However, considering the similarities (e.g., the aggregated spatial structure) and the differences (e.g., variation in the dominant cover type composing expansion) in closely located plots which share the same historical contingency, pool of species, grazing pressure and environmental conditions, the patterns found in our study highlight the complementarity of smaller scale investigations when interpreting vegetation dynamics and recovery (Hu et al. 2012; Quero et al. 2011). As an example, in an applied perspective, this complementarity may refine the calculation of ecosystem services, such as carbon stock, it may improve the evaluation of scale-dependent species richness patterns and, consequently, help design more effective restoration programs (Chuang et al. 2018; Méndez et al. 2008; Novara et al. 2017). Finally, we demonstrate that in favorable conditions, especially in the presence of remnant woody cover and in the absence of fire, even a long history of deforestation, land use and the presence of livestock do not crush the potential for intense woody vegetation expansion, even though previous cover tends to "sequester" this potential for recovery.

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Author contributions RDSB, TLM and MG conceived the study. RDSB and DG developed the sampling design, methodology and the analytical framework, and interpreted results. RDSB collected and analyzed the data and wrote the first manuscript. DG revised the first manuscript and all authors revised and improved the final version.

Compliance with ethical standards

Conflict of interest The authors declare that they have no conflict of interest.

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