# More than environmental control at local scales: do spatial processes play an important role in floristic variation in subtropical forests?

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#### **ABSTRACT**

The effect of environmental variables on the plant species distribution in the Atlantic Forest raises the interest of researchers, but few studies, mainly at local scales, analyzed the influence of spatial factors on the environment and species distribution. Linear models and spatial correlograms were used to evaluate whether environment and space may predict variations in species composition of trees and shrubs in a Subtropical Rainforest remnant. The study area was divided in 25 plots of  $400 \text{ m}^2$ , systematically distributed, where individuals with DBH  $\geq 5 \text{ cm}$  were measured. Elevation and slope were the most important predictors of the floristic variations, but space, generally neglected in researches on South American's vegetation, had a relevant influence, as a structural component, in terms of environmental variables, and as intrinsic biological component, here probably as result of constraints in species dispersion. Although the origin of great part of the variation in species composition remained unknown, which is common in studies on vegetation, results indicated, for the first time, the complex role of spatial and environmental variables in the composition of trees and shrub species in a Subtropical Rainforest of Brazil.

Keywords: Atlantic Forest, environmental gradient, linear models, species distribution, variance partitioning

#### Introduction

Studies analyzing the effects of environmental factors on community composition patterns may provide evidence on the ecological preferences of these species in different abiotic conditions (Peña-Claros et al. 2012; Sanchez et al. 2013; Thuiller 2013). Environment heterogeneity is one of the main factors influencing species composition in tropical forests. Its effects may be related to edaphic factors, light and water availability, and topographic characteristics (Jones et al. 2011; Peña-Claros et al. 2012; Baldeck et al. 2013; Eisenlohr et al. 2013; Oliveira-Filho et al. 2013; Sanchez et

al. 2013; Thuiller 2013). Topography and physicochemical soil properties have been recognized as crucial drivers of vegetation mainly at local scales (e.g., O'Brien et al. 2000; Bohlman et al. 2008; Machado et al. 2008; Souza et al. 2012; Toledo et al. 2012; Mélo et al. 2013). Topography, for example, plays a fundamental role in the description of variations in the composition of vegetation, mainly because a variation in topography may cause alterations in the physical and chemical properties of soils (John et al. 2007; Baldeck et al. 2013; Mélo et al. 2013).

Despite the relevance of environmental factors as

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indicators of species distribution along gradients (e.g., Gentry 1988; Toledo et al. 2012), a more recent ecological approach takes into account both the spatial structure of plant communities and environmental factors, especially the influence of spatial structure on variations in species composition (Lichstein et al. 2002; Legendre et al. 2009; Eisenlohr et al. 2013; Gasper et al. 2013; Mélo et al. 2013; Oliveira-Filho et al. 2013; Lewis et al. 2014). Spatial structure may be caused by contagious ecological processes such as dispersion (Legendre 1993; Cottenie 2005), but environmental factors may also have spatially structured fractions in space, because neighbor sites tend to present similar environmental characteristics (Fortin & Dale 2005). When the spatial structure of plant communities is analyzed, ecological variables such as dispersion and competition mechanisms arise as important drivers of species distribution (Cottenie 2005; Diniz-Filho et al. 2012; Baldeck et al. 2013). When spatial structure is disregarded, incorrect statements may be made on species composition patterns related to environment (Legendre & Legendre 2012). Furthermore, if space is disregarded, interpretation of numerical analyses of ecological studies may be biased and compromise the reliability of statistical tests (Peres-Neto 2006; Diniz-Filho et al. 2007; Dale & Fortin 2009). In addition, it is particularly interesting to investigate which part of variation in species composition is explained by fractions corresponding to environment, to spatially structured environment, and only to space (Peres-Neto 2006; Peres-Neto & Legendre 2010).

In recent years, the Atlantic Forest, one of the world's 35 hotspots for conservation of biodiversity (Ribeiro et al. 2011), has been the object of many studies aimed at verifying the effect of environmental variables on vegetation distribution patterns along environmental gradients (Budke et al. 2007; Ferreira-Júnior et al. 2007; Rodrigues et al. 2007; Silva et al. 2009; Gonçalves et al. 2011; Souza et al. 2012; Higuchi *et al.* 2013; Marcuzzo *et al.* 2013; Mélo *et al.* 2013). However, few studies have analyzed the influence of spatial components on the environment and distribution of vegetation (Machado et al. 2008; Mélo et al. 2013). Studies that address such influence would be appropriate, because they could improve the predictive power of ecological modeling, thereby providing a more complete assessment of the drivers of floristic variation, and because they could be useful to inspire further large-scale studies with more generalization power.

The present study assessed whether environment and space may consistently predict variation in the composition patterns of tree and shrub species in a subtropical Atlantic rainforest remnant in order to answer the following questions: (i) Are there environmental factors influencing tree and shrub species composition patterns? We expected to find a great effect of soil and terrain topography in determining floristic and structure composition, because they are very well known as the main factors conditioning changes in the composition of vegetation in the Atlantic

Rainforest at local scales. (ii) Does the spatial component have an influence on the composition patterns of tree and shrub species together with the environment? We expected to find a significant effect of space, because recent studies have indicated that inclusion of the spatial component in the analysis of vegetation patterns helps in the identification of potentially important processes. Thus, the spatial component may be important to determine the distribution of species along environmental gradients.

# Materials and methods

Study area

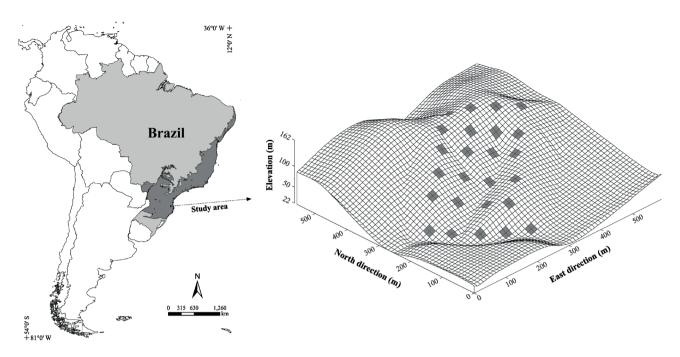
The study area is within the catchment area of the Itajaí River in Brusque, SC. It belongs to the RPPN Chácara Edith, which has a total area of 415.8 ha and an altitude ranging from 20 to 300 m asl, and it is located between  $27^{\circ}05$ ' and  $27^{\circ}06$ ' S and  $48^{\circ}51$ ' and  $48^{\circ}54$ ' W (Fig. 1).

The predominant vegetation in the study area consists of Submontane Subtropical Evergreen Broadleaf Rainforest (sensu Oliveira-Filho 2015), hereinafter referred to as subtropical rainforest, being located in the Atlantic Forest Domain. In 1923, the former owner of the reserve began logging in the study area; this activity continued until the 1930s. The logging was focused on blackleg (Ocotea catharinensis Mez) and other species of economic interest (Hoffmann et al. 2002). The reserve was explored during the beginning of the 1900s, but the vegetation has gone about 80 years without any anthropogenic action (Maçaneiro et al. 2015).

The climate of the city of Brusque, according to the Köppen classification, is Cfa – a humid mesothermal subtropical climate with hot summers (Alvares *et al.* 2014). The annual average temperature varies between 19 and 20° C, with monthly averages ranging from 15° C in the coldest month (July) to 25° C in the warmest months (January and February). The annual relative humidity ranges between 84 and 86%, with monthly averages between 80% in the driest month (December) and 88% in the wettest months (June and July) (Pandolfo *et al.* 2002). The data collected by the Brusque Station of the National Water Agency (27°06′02.16″S, 48°55′04.08″W, 25 m asl), 2.8 km away from the study area, register a total annual rainfall ranging from 460 to 3,004 mm, with an average of 1,568 mm, from 1941 to 2013.

The geology of the region is formed by the Brusque Metamorphic Complex, composed of the Botuverá Formation, which is itself composed of Guabiruba Granite and Valsungana Granodiorite (Schulz-Júnior & Albuquerque 1969; Santa Catarina 1986). This formation consists of a metavolcano-sedimentary association composed of phyllites and micaceous schists intercalated with semipelites, quartzites, calc-silicate rocks, marbles, magnesian schists, and basic metavolcanic rocks (Philipp *et al.* 2004).

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**Figure 1.** Location and surface grid showing the topography of the area studied in the RPPN Chácara Edith, Brusque, Santa Catarina, with the distribution of 25 plots of 20 x 20 m used to sample the vegetation.

In the study area, an association between two soil types predominates: moderate A Tb Haplic Cambisol in undulating relief of moderately argillaceous texture and moderate A dystrophic Litholic Neosol in strongly undulating mountainous relief of moderately argillaceous texture (EMBRAPA 2013).

#### Data collection

To collect data, we used plots arranged in groups (Soares et al. 2012) with the aim of representing the largest possible variation of vegetation along the slope analyzed. We placed the plots along five transects, running from the base to the top of the slope. In total, we systematically distributed 25 plots of 20 x 20 m (400 m²), totaling a sampling area of 10,000 m². These plots were approximately 50 m apart, and there was a distance of 25 m between tracks. In each plot, we sampled all individuals, living trees, and shrubs with DBH (diameter at breast height)  $\geq$  5 cm.

We identified the collected botanical material by comparing it with digital photos from sites belonging to Flora Digital (http://ufrgs.br/floradigital) and Virtual Herbarium (http://reflora.jbrj.gov.br), with dried plant specimens deposited at the Dr Roberto Miguel Klein Herbarium of the Regional University of Blumenau (FURB), and also in consultation with taxonomic literature and FURB experts. We sent duplicates of species of the Lauraceae and Myrtaceae families to specialists.

Two composed samples of superficial soil (0-20 cm depth), weighing around 500 g each, were collected in each plot of the experimental area, according to the

recommendations of Santos et al. (2013). Each composed sample was made of various subsamples collected in each plot. Samples were stored and identified in plastic bags and were sent to the Laboratory of Soil Analysis of the Universidade do Estado de Santa Catarina (UDESC/Lages) to identify the following chemical variables: pH-H<sub>2</sub>O, SMP index, calcium (Ca), magnesium (Mg) and aluminum (Al) contents, potential acidity (H + Al), effective cationic exchange capacity (CEC), Al saturation, bases saturation, organic matter (OM), clay content, phosphorus (P), sodium (Na), potassium (K), and (CEC). The physical characteristics analyzed were granulometry, with proportions of silt, sand, and clay. The drainage class of soil was evaluated for each plot, according to EMBRAPA (2013). We measured slope in degrees and the gradient of each plot with a Bosch GLM 80 laser distance meter.

Modeling species composition with spatial and environmental variables

The influence of spatial and environmental variables on the composition of tree and shrub species was analyzed by transformation-based redundancy analyses (tb-RDA; Borcard *et al.* 2011). In the present case, we used the Hellinger transformation (see below). We chose the tb-RDA technique because it is a composition of linear models and does not use chi-squared distance, avoiding biases caused by other multivariate techniques of direct analysis of gradients, like CCA (Legendre & Gallagher 2001; Borcard *et al.* 2011; Legendre & Legendre 2012). With the intention

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to identify and remove collinear environmental variables, a principal components analysis (PCA) was performed in the PC-ORD 6.0 (McCune & Mefford 2011). After this procedure, environmental variables that remained in the analysis were slope, elevation, pH-H<sub>2</sub>O, sand and clay contents, Ca, Mg, OM, K, P, Al saturation, and bases saturation. In the matrix of presence and absence of species, the species with low frequency in plots were removed from the analysis, because they give low contribution to the community structure and may hinder the analysis of data (Causton 1988). In the present study, low-frequency species were considered the ones appearing only in up to four plots, because when the preliminary tb-RDA was performed, these species did not show a strong relation with the studied environmental variables. After this procedure, the Hellinger transformation was applied to the matrix of species occurrence, with the objective of avoiding biases caused by the Euclidean distances in RDA (Legendre & Gallagher 2001). Standard Score transformation was applied to the environmental variables, with the objective to correct differences in scale (Zar 2010). The geographical coordinates (latitude and longitude) of each plot were used to create spatial variables named MEMs (Moran's Eigenvector Maps), allowing a better understanding of the role of space in different scales (Dray et al. 2006).

During this step, we used the software "spacemakeR" in R environment (R Core Team 2013), according to the recommendation of Borcard et al. (2011), and MEMs were created by a connectivity matrix obtained through a Delunay triangulation (Borcard et al. 2011). Eighteen MEMs were created, considered as predictors of the species composition in a model generated by a tb-RDA, which regressed the MEMs on the matrix of species. Selection of significant MEMs for the model was obtained by the "forward" method (Blanchet et al. 2008). Next, we selected the predictive environmental variables by means of another tb-RDA. Finally, the last tb-RDA was processed in PC-ORD 6.0 (McCune & Mefford 2011) using species matrixes (Hellinger-transformed), with the selected environmental (standardized) and spatial (MEMs) variables, in order to investigate the effect of these variables on the species composition along the analyzed environmental gradient. This tb-RDA will henceforth be referred to as "final tb-RDA."

We observed an "arch effect" in the graph of the first two axes of the final tb-RDA. Despite some criticisms, we agree with Vieira *et al.* (2015), for whom "arch-like curvature is an important, inherent property of successive-replacement data, which must be considered in any discussion or analysis (...)." Although we decided to maintain the tb-RDA, this requires caution with respect to the association of explanatory variables to plots in a linear framework in case such an association would be performed for each axis individually. As Hill & Gauch (1980) argue, "for the axes to be separately interpretable, they need to be independent, not merely uncorrelated." Thus, in addition to the abovementioned caution, we calculated the contribution of each

explanatory variable to the whole final tb-RDA rather than restrict our analysis to a subset of axes. This was done by means of partial tb-RDAs followed by permutation-based ANOVA.

The presence of spatial autocorrelations was verified by Mantel and Moran correlograms (Fortin & Dale 2005) using PASSaGE 2.0 (Rosenberg & Anderson 2011). We obtained the correlograms to provide additional information on the spatial effect on plant communities, thus unveiling, for example, at which distance the spatial autocorrelation disappears. The variables used for preparing Mantel correlograms were the occurrence of the 20 and 10 most frequent species as a whole. The variables used for preparing Moran correlograms were species richness and the occurrence of each of the 10 most frequent species, separately.

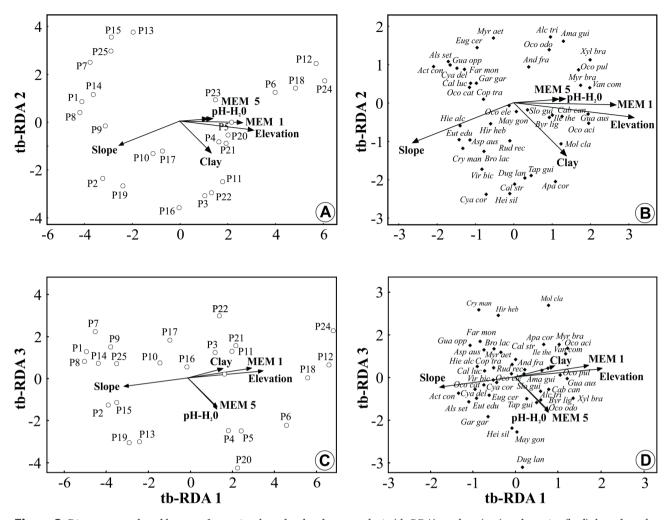
### Variance partitioning

Having the previously selected spatial and environmental variables at our disposal, we performed the variance partitioning to know how "pure" environment [a], spatially structured environment [b], "pure" space [c], and undetermined variables [d] could explain variations in the species composition of our database. We evaluated whether fraction [a] influenced species composition, thus controlling type I error caused by fraction [b] (Peres-Neto 2006; Peres-Neto & Legendre 2010). In this step, we ran a permutation-based ANOVA to test the significance of fractions [a] and [c] (Legendre & Legendre 2012). The R-codes used followed Eisenlohr (2014).

# Results

Eigenvalues of the first three tb-RDA axes explained 32.6% of the variance (axis 1=20.3%; axis 2=7%; axis 3=5.3%) and appeared highly significant (Monte Carlo, P<0.001). Environmental and spatial variables that showed the highest correlations with axis 1 were elevation, slope, clay content, and MEM 1 (spatial variable); the variables most correlated with axis 2 were slope and clay content, and those most correlated with axis 3 were pH-H<sub>2</sub>O and MEM 5 (Fig. 2, Tab. 1).

In the ordination diagram of plots generated by the final tb-RDA (Fig. 2A, C), we verified that the first three axes discriminated a strong gradient between elevation, MEM 1, and slope. These axes segregated three groups of plots: the first group, located at the base of the slope with the greatest slope; the second group, located in the middle of the slope where the elevation and slope were intermediate; and the third group, on the top of the slope, with high elevation and a small slope. Other shorter gradients discriminated by the first three axes of the final tb-RDA were related to clay content, soil pH-H $_2$ O, and MEM 5. The first and second axes of this final tb-RDA (Fig. 2A) revealed that clay contents in



**Figure 2.** Diagrams produced by transformation-based redundancy analysis (tb-RDA) to plots (a, c) and species (b, d), based on the presence and absence of 42 species in 25 plots sampled in a subtropical rainforest in southern Brazil.

**Table 1.** Contribution of each predictor to the overall floristic patterns yielded by the final tb-RDA. R<sup>2</sup> adjusted, F and p were obtained after partitioning analysis. Significant results are in bold.

Predictor	R <sup>2</sup> adjusted	F (ANOVA)	P (ANOVA)
Slope	4.41%	2.0985	0.013
Elevation	3.51%	1.8738	0.015
MEM 5	2.93%	1.7297	0.022
Clay	2.30%	1.5730	0.037
рН	2.16%	1.5375	0.041
MEM 1	0.53%	1.1314	0.306

soil are strongly related to plots located in the middle of the slope. Plots located at the base of the hillside showed smaller clay content in soil. The first and third axes of the final tb-RDA (Fig. 2C) indicated that some plots located in the middle of the slope presented high pH-H<sub>2</sub>O values in soil and were related to MEM 5. Furthermore, this relation was inversely correlated with plots located in the base of the hillside, whose pH-H<sub>2</sub>O values were the smallest.

Because of the "arch effect," which creates dependence among successive axes, caution is needed to make associations between plots and explanatory variables for each axis separately. Here, a more complete view of the hierarchy among descriptors — i.e., the effect of such descriptors on the overall species distribution patterns — indicates that slope was the most important variable, followed by altitude and MEM 5 (Tab. 1). The other spatial

variable, MEM 1, was not significant (Tab. 1).

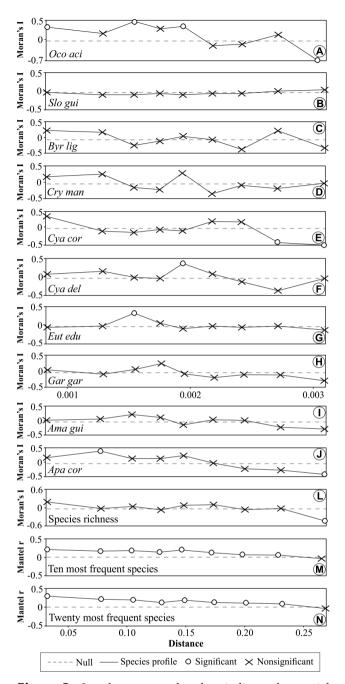
In the diagram of species ordination (Fig. 2B, D), the three first axes revealed a strong association between spatial and environmental variables with the species composition. We verified some segregation patterns between species and plots in the final tb-RDA, probably related to the topographic position in the hillside. The first group of species was associated with plots belonging to less elevated plots with a higher slope (e.g., Actinostemum concolor, Alsophila setosa, Calyptranthes lucida, Cyathea delgadii, and Euterpe edulis), with this environment defined as the base of the hillside. Additionally, a group of species was verified as strongly associated with the high-elevation and low-slope plots (e.g., Amaioua guianensis, Aparisthmium cordatum, Guappira opposita, Myrcia brasiliensis, Ocotea acyphilla, Ocotea pulchra, and Xylopia brasiliensis), in other words, to the topof-the-slope environment. On the other hand, some species did not show evident association with elevation or slope (e.g., Andira fraxinifolia, Brosimum lactescens, Byrsonima ligustrifolia, Copaifera trapezifolia, Cyathea corcovadensis, Maytenus gonoclada, Ocotea elegans, and Sloanea guianensis), because they presented uniform distribution along the entire analyzed environmental gradient. Besides, MEM patterns in the ordination diagrams coincide with the environmental variables elevation and pH-H2O, indicating that the spatial component has an influence on both species distribution and the environment.

Correlograms indicated that, when considered as a whole, species occurrence presented strong spatial autocorrelation, which disappeared only at larger distances (Fig. 3; see correlograms for the 20 and 10 most frequent species). However, species richness was much less structured at space, and the same was verified for most of the species occurrence, individually (Fig. 3). Thus, the spatial autocorrelation stood out only at the multivariate level.

Partitioning of variance indicated that fractions [a] "pure" environment (F = 1.906; P = 0.005) and [c] "pure" space (F = 1.436; P = 0.03) were significant (Fig. 4). The "pure" environment stood out among fractions, but fraction [d] indicated that most of the floristic variance (76%) remained unexplained (Fig. 3), indicating the existence of other predictive factors that were not evaluated in this study.

# **Discussion**

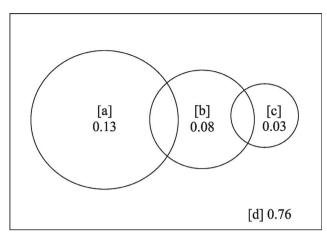
Our results show that only a few of the studied environmental variables have an influence on the distribution patterns of tree and shrub species in a gradient of the subtropical rainforest in southern Brazil. They also show that space, which is generally neglected in investigations on meridional South American vegetation, plays a relevant role, even if smaller than environmental variables. This smaller contribution of space to species composition could be due to the typical lowest proportion of wind-dispersed species in tropical rainforests (e.g., Howe & Smallwood



**Figure 3.** Correlograms produced to indicate the spatial autocorrelation of species occurrence, individually (a–j) and for larger datasets (m–n), and the spatial autocorrelation of the species richness (l).

1982); thus, it is possible that in more open physiognomies, such as savannas or dry forests, this contribution would be higher. Nevertheless, if we consider the influence of space shared with environmental variables (fraction [b]), the role of spatial components reaches almost half of the known drivers of floristic variation.

Considering the environmental influence on species distribution, differences between the analyzed species composition are mainly related to variations in the elevation



**Figure 4.** Variance partitioning among the fractions that explain the species distribution in a subtropical rainforest in southern Brazil. [a] only environmental; [b] environmental + space; [c] only space; [d] residuals.

and slope of the hillside, which were the most important environmental variables segregating plots and species in the analyzed hillside. Studies on environmental gradients performed on local scales have demonstrated that the distribution of vegetation is more strongly associated with topography and soil (Souza et al. 2003; Rodrigues et al. 2007; Souza et al. 2012; Guerra et al. 2013). Ferreira-Júnior et al. (2012) mentioned that changes in topography provide alterations in the physical and chemical properties of soil, which may influence the distribution patterns of vegetation. However, significant relations between variables associated with chemical and physical properties of soil with variations of vegetation along the analyzed gradient were not verified. The proposition of Klein (1980) was confirmed, verifying that, along the hillside in the region of Vale de Itajaí in Santa Catarina, it was possible to observe a clear differentiation in the composition of vegetation. According to that author, this observation was more evident when the slope was higher. This pattern is probably related to variations in topography, which, in turn, may reflect the high environmental heterogeneity observed in the subtropical rainforest of Southern Brazil (Klein 1980, 1984; Leite 2002; Oliveira-Filho et al. 2013).

Elevation is an environmental variable that indirectly influences a series of other environmental factors intimately related to the distribution patterns of vegetation in the Atlantic Forest (Oliveira-Filho & Fontes 2000; Meireles et al. 2008; Eisenlohr et al. 2013; Oliveira-Filho et al. 2013; Sanchez et al. 2013). In this study, elevation was one of the greatest predictors of the composition of vegetation. In a similar way, Gasper et al. (2013), analyzing distribution patterns of fern species in Santa Catarina, found a significant correlation between elevation and these patterns. These authors verified that variation in elevation conditioned changes in the climatic variables and that these variables related to the distribution patterns of fern species. The

strong relation between elevation and distribution of vegetation is probably related to the topographic position along the hillside. In the base of the hillside, soil conditions and topography, such as microclimatic properties, are different when compared to the middle and top of the slope (Veloso & Klein 1959; Klein 1980, 1984; Ferreira-Júnior *et al.* 2007; Jones *et al.* 2011; Ferreira-Júnior *et al.* 2012; Marangon *et al.* 2013).

Floristic differences in the function of the topographic position in the slope highlight the influence of environmental factors on the distribution of species in the studied area. Plots belonging to less elevated positions of the slope and with a greater slope presented a very characteristic set of species. Similarly, we verified associations of species in plots located in higher positions and with a smaller slope. In the subtropical rainforest of the Vale do Itajaí, SC, these floristic patterns were also observed by Klein (1980; 1984). This author concluded that, in the base of the slopes of the region, groups of species highly adapted to soil conditions and topography characterized the vegetation, and that it was possible to meet different floristic associations while moving to the top of the hillsides. Differences between these floristic associations in the function of topographic conditions are possibly related to variations in elevation and soil topography, which could play an important role in the physical and chemical properties of soil and, consequently, in species distribution (e.g., Ferreira-Júnior et al. 2007; 2012).

Abiotic factors, together with dispersion and biotic interactions, are often suggested to explain species distribution (Boulangeat et al. 2012). Thus, besides environmental factors, another variable set that can have ecological importance in the distribution patterns of species in environmental gradients of an Atlantic rainforest is space (Eisenlohr et al. 2013; Gasper et al. 2013; Mélo et al. 2013; Oliveira-Filho et al. 2013). In the present study, the heterogeneity of the environment (here represented by edaphic and topographic variations) was not the only process explaining variations in the species composition. The spatial component, for example, revealed a significant fraction of influence on the floristic patterns, indicating that variations in the species composition may be due to biotic and stochastic processes (Hubbell 2001; Diniz-Filho et al. 2012; Peña-Claros et al. 2012). A fraction related to spatially structured environment ( $R^2 = 8\%$ ) may also be used to justify the distance-decay floristic similarity (Diniz-Filho et al. 2012). Legendre et al. (2009) and Diniz-Filho et al. (2012) suggest that a spatially structured environment has an influence on the spatial distribution of species, because as distance between environments increases, the higher their difference in terms of species composition. In smaller spatial scales, a spatially structured environment may favor limitations in the dispersion of species, and neighboring environments tend to be similar (Karst et al. 2005). This fact is very clear in the studied hillside, because the strongest predictive variables (elevation and slope) were spatially structured along the gradient, as also observed by Legendre et al. (2009).

Our results also demonstrated that spatial structures could be more sophisticated than a mere analysis of synthetic variables (e.g., species richness) would suggest. In fact, the spatial structure became clearer when the species occurrence was considered in a multivariate framework, i.e., by means of Mantel correlograms. Because species often co-occur in a non-random way — the observed co-occurrence in most natural communities is usually less than expected by chance (Gotelli & McCabe 2002) — a more complete view of the community behavior along gradients requires deeper analysis. This interpretation also agrees with the fact that the variable MEM 1 (coarser scale) was not significant in the overall tb-RDA, but MEM 5 (finer scale) was significant.

Furthermore, Lewis et al. (2014) pointed out that biotic processes like dispersion, facilitation, and competition between species may be attributed to the spatial component. Different species have different dispersion mechanisms that may provide different spatial distribution patterns, which were confirmed by the individual Moran correlograms found in this study. Dispersion by animals, for example, provides greater spatial heterogeneity, which may end in accumulation of seeds in preferred locations or in paths preferred by animals (Schupp et al. 2002). On the other hand, dispersion by wind provides a more uniform distribution pattern in terms of seed dispersion. Furthermore, the density of individuals of one species also has an influence on the spatial distribution pattern and, consequently, on recruiting and mortality rates of individuals (Crawley 1986).

We found a very high percentage of unexplained variation (76%). This result is common in studies considering spatial and environmental variables as predictive of the distribution of species (Souza et al. 2003; Machado et al. 2008; Jones et al. 2011; Peña-Claros et al. 2012; Baldeck et al. 2013; Gasper et al. 2013). It suggests that some variables that were not considered in the model would probably increase the explanation of variation patterns in the vegetation of the studied area. Our results agree with Baldeck et al. (2013) in relation to the hypothesis that variables such as luminosity, microclimate, and soil humidity may be potentially predictive of variations in the vegetation of tropical forests. Legendre et al. (2009) pointed out that non-spatially structured biological processes may also explain part of the residual fraction in ecological models. Besides environmental variables, other factors bound to niche conditions may have an influence on the distribution of species in tropical forests (see Baldeck et al. 2013). For example, competition is an important factor in the colonization process of species (Gotelli 1999), mainly because some species have shown a preference for determined niche conditions. Stochastic factors such as dispersion, which are bound to neutral processes, are also relevant (Hubbell 2001; Cottenie 2005), because in the Atlantic Forest there are complex interactions between plants and dispersers. Furthermore, dispersion syndromes and dispersion agents vary with elevation (Almeida-Neto et al. 2008).

In the present study, the partition of variance indicated a great percentage of unknown fractions in the explanation of composition patterns, but it also revealed the statistical significance of environmental and spatial factors as promoters of species distribution. The fraction related to "pure" environment (13%) highlights that niche processes could act on floristic patterns, because we found environmental variables strongly related to species composition. On the other hand, the low explanation of the fraction "pure" space (3%) could be understood by considering the high spatial structuration of the environmental variables along the hillside. Spatial factors resulting from limitations of seedling dispersion (Peña-Claros et al. 2012) or from neutral processes (Diniz-Filho et al. 2012) show lower influence on the variation of distribution patterns. However, even with a low level of explanation, the contribution of the "pure" spatial component was significant, indicating that factors related to spatial structure have an influence on species composition patterns. Therefore, studies on floristic patterns of the Atlantic Forest must take into consideration the spatial component. Gilbert & Lechowicz (2004) highlight that measurement of dispersion effects may be difficult due to the spatial structure and variability found in ecosystems. Thus, these authors recommend the use of more complex models that consider, for example, the effect of interactions between dispersion and competition, because they can offer greater potential in explaining these results. Environmental factors related to elevation, topography, and soil are important predictive variables for the distribution of tree and shrub species in hillsides of the subtropical rainforest in southern Brazil. In this study, environmental factors related to physical and chemical characteristics of soils did not present a strong relation with distribution patterns of vegetation.

The present work indicates a statistically significant role of environmental and spatial variations in the composition of tree and shrub species in a hillside of the subtropical rainforest in Brazil, as well as the complex role of both sets of predictors. Considering that this study focused on a local scale, other works with the same objective must focus on other locations of the subtropical rainforest. It is important for these other works to investigate association patterns between species, to put these patterns in relation with environmental factors, and to test the influence of other predictive variables like competition, dispersion, luminosity, microclimate, and soil humidity.

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