Spatial structure of the vertical layers in a subtropical secondary forest 57 years after clear-cutting

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Introduction
A forest is a three-dimensional ecosystem, and its structure can vary in both the vertical and horizontal directions. The horizontal distribution pattern of trees has always been a research hotspot. It is believed to be directly related to ecological processes within the forest community (e.g., seed dispersion, regeneration, death, resource utilization, and gap formation – Wiegand & Moloney 2013, Omelko et al. 2018), and strongly influences tree growth and timber yield (Del Río et al. 2016). In contrast, vertical structure has not attracted much attention. This may be because the vertical structure is difficult to measure directly and the results are difficult to apply in practice. However, several studies have indicated that vertical structure plays an important role in forest ecosystems (Hao et al. 2007, Gazol & Ibáñez 2009). It affects the forest dynamics and modifies the supply, capture, and efficiency of use of resources, and thus also inter- and intra-specific interactions (Del Río et al. 2016). It is also closely related to the ability of a forest stand to resist external non-biological interference, and the number and species of wild animals living in the forest (Kuuluvainen et al. 1996).

Vertical stratification of trees is an important phenomenon in natural forests (Ishii et al. 2000). The upstory constitutes the skeleton of the forest community. A change of canopy structure in the horizontal and vertical directions leads to the formation of a complex light and shade structure in the understory (Giesbrecht et al. 2017), determining the species composition, distribution pattern, and ecological processes in the understory. Illumination may be the most readily available resource for understory (Martens et al. 2000). When seedlings are successfully renewed and escape competition from the groundcover, light often becomes a key factor for the survival and growth of young trees in the understory (Paluch 2005). Through canopy
interference, the upstory will increase the variability and heterogeneity of understory resources, and change the internal rainfall distribution and quantity (Paluch 2005), soil properties (soil organic matter, available N, cation exchange capacity, litter thickness, soil calcium content, humidity, temperature, pH, and frost – Beatty 1984), and litter decomposition (Gazol & Ibáñez 2009). This in turn affects the productivity and diversity of understory (Giesbrecht et al. 2017). In addition, the canopy density of the upstory also affects the seed distribution of understory and upstory in the forest and the seed bank composition (Taylor et al. 2004).

The understory may also affect regeneration, seedling growth, and forest dynamics by changing the soil seed bank, invading and occupying forest gaps, which will eventually have an impact on the upstory structure and species composition (Taylor et al. 2004, Giesbrecht et al. 2017). In both tropical and temperate regions, the interaction between upstory and understory is considered to be an important factor determining the structure and dynamics of forest communities (Taylor et al. 2004). However, the relationship between upstory and understory is not fully understood (Barbier et al. 2008). Most studies of the spatial structure of forest communities consider the species comprising the community as a whole (Hui & Pommerringen 2014, Carrer et al. 2018), while others have analyzed the spatial correlation of the different life histories of single populations or several groups of the same genus (Hao et al. 2007, Zhang et al. 2010, Omelko et al. 2018). Still other studies have focused on tree groups that belong to certain synusia or growth stages, such as seedlings and small trees (Collet et al. 2017), or dominant trees (Torimaru et al. 2013, Ghalandarayeshi et al. 2017). Elsewhere, the spatial association between specific objects, such as dead trees and living organisms, have been considered (Salas et al. 2006, Xiang et al. 2013), in addition to regeneration and dead wood or earthen mounds (Janik et al. 2014). However, few studies have stratified forest communities and considered each layer as a separate part, especially at the meso-scale (Gazol & Ibáñez 2009). Each layer may have different growth or distribution pattern characteristics, providing valuable clues for investigations of the mechanisms by which species coexist.

The species composition of forests in temperate regions is relatively simple and, due to the particular renewal and growth characteristics of some species (e.g., “wave”, “stop-and-go”, or “continuous and fast growing” strategies – Hao et al. 2007), vertical stratification is common (Gazol & Ibáñez 2009, Torimaru et al. 2013, Ghalandarayeshi et al. 2017, Omelko et al. 2018). However, in tropical South Asia, there is an abundance of species components and a continuous distribution of species in the vertical direction can readily occur. Many climbing vines may also be attached to trees (Yuan et al. 2015), which further influences the vertical stratification of the forest community. In addition, the “slash-and-burn” agricultural management model was widely practiced in the region. This destroyed large amounts of primary vegetation, resulting in a complex composition and structure of secondary forests (Xiang et al. 2013). The mixed pine-oak forest along the Nanpan River in southwest China is a typical example; it is a secondary forest that developed after the original forest suffered clear cutting in 1960s and was then completely abandoned and left unused. After self-development, it formed a forest that mainly consists of Pinus yunnanensis var. tenuifolia and several species of Quercus (Li et al. 2017). It is currently in the early stage of succession with high species diversity, but vertical stratification is very obvious. This provides a good opportunity to study the construction of secondary forest structure after disturbance.

The purpose of this study was: (1) to identify the distribution pattern and spatial correlation of trees in the early stages of a secondary forest; and (2) to determine the spatial distribution of tree species and tree size in the vertical layers of the forest.

**Material and methods**

**Study area and field measurements**

The study site was located in Yachang Township, Leye County, Baise City, Guangxi Zhuang Autonomous Region (Fig. S1 in Supplementary material). The Nanpan River, which originates in Yunnan Province, separates Guangxi and Guizhou provinces as it passes through this region. The southern part of Guangxi Zhuang Autonomous Region has a typical karst landform, with limestone as the main rock type, but there are also some large mountains comprised of weathered rock. Historically, the majority of the township was part of the state-owned Yachang Forest Farm, which was later renamed the Guangxi Yachang National Orchid Reserve due to the depletion of forest resources, together with the existence of a large number of wild orchid communities and the natural landscape of the “Tiankeng Group”. Temperature, soil types, and rainfall in the region vary dramatically with elevation and geomorphology (Li & Wang 1981). The river valley has a relatively high temperature and long hot summers. Its average temperature in July is 28.4°C, with an extreme high temperature of 42.5°C (historical record since 1981). There is no frost and snow in winter. The average annual rainfall is 800 mm, but the distribution is extremely uneven. Summers are wet, but spring and autumn are rainless and seasonal droughts can occur. The soil type gradually transitions from cinnamon soil in the river valley to cinnamon red soil in the mountain plain and a yellow soil in the mountains. The soil is generally barren, with a high gravel content (Li et al. 2017, Yu et al. 2018).

The sample site was located on a mud hill, 20 km from the administrative center of Yachang township. The elevation was about 760 m a.s.l. and the study site was about 250 m from the river. The forest stand was divided into two vertical layers. The upper layer mainly contained *P. yunnanensis*, *Quercus variabilis* Bl. and *Albizia kalkora* (Roxb.) Prain, while the lower layer was dominated by *Vaccinium bracteatum* Thunb., *Phyllanthus emblica* L., *Wendlandia uvarifolia* Hance, *Crabidobiondron stellatum* (Pierre) W. W. Smith, *Ardisia quinquegona* Bl., and *Toona ciliata* M. Roem. Some orchids were present in the stand, including *Dendrobium officinale* Kimura & Migo, *Cleistostoma fuerstenbergianum* Kraenzl., *Vanda concolor* Bl. and *Cymbidium aloifolium* (L.) Sw. Other herbs in the stand included *Thysanolaena maxima* (Roxb.) Kunzte and *Miscanthus floridulus* (Lab.) Warb. ex Schum et Laut (Fig. S2 in Supplementary material). Because it was situated in a mountainous location far from residential areas, the secondary forest had suffered almost no external interference, although some of the *P. yunnanensis* were scarified for sap in 1990a (Li et al. 2017).

![Fig. 1 - Distribution of trees species in the sampling plots (a: plot 4; b: plot 5). Rings and crosses represent the locations of understory and upstory plants, respectively.](image-url)
Vertical structure of a secondary forest

We established two rectangular standard plots (plot 4: 100 m × 60 m, up-slope, 106° 19’ 20” N, 24° 51’ 15.95” E; plot 5: 80 m × 70 m, mid-slope, 106° 14’ 9.5” N, 24° 23’ 28.9” E) in 2016-2018. The two sample sites were approximately 120 m apart, and had similar vegetation components. First, we divided each area into several subplots (20 × 20 m each) and determined the location of each living tree, standing dead trees, and fallen dead wood using a total station instrument (Southern Mapping Company, Johannesburg, South Africa – precision is 2 mm + 2 ppm). We measured all stems with at least 1 cm DBH. Then, we measured the tree height, DBH, and crown width of all standing trees. We also recorded species name and marked each living individual with a unique reference number. For fallen trees, the grade of decay (i-v), length of branches, and size of both ends were measured and recorded (Fig. 1). The diversity of herbs was determined using a systematic sampling method. We surveyed 26 and 27 tree species, 1745 and 1319 living trees, and 265 and 168 dead trees in plots 4 and 5, respectively. However, we only used the data of living trees which include all species for analysis.

Data analysis

Classification of stratification

Tree height is a particularly important index of vertical structure and can be used to describe the stand (Del Río et al. 2016). Vertical stratification in this study was determined according to the height of the lower layer; a box graph of the tree height of the two forests showed that the height of the lower layer was slightly more than 6 m (red dashed line in Fig. 2 – see also Fig. S2 in Supplementary material). Most lower layer plants were shrub species, accounting for 55.58 and 53.44% of the total number of stand plants and 84.61 and 89.47% of the overall populations in plots 4 and 5, respectively (Tab. S1). The understory also included some newly settled tree species that were still small (e.g., T. ciliata, and Toxicodendron vernicifluum [Stokes] F. A. Barkl.). Four tree species Q. variabilis, P. yunnanensis, A. kalkora, and C. glauca occupied both the upper and lower layers, but were dominant in the upper layer (Fig. 2). The C. glauca population only had 21 and 19 individuals in plots 4 and 5, respectively.

Spatial pattern analysis

A pair correction function (PCF), g(r), based on the paired point distance was used for spatial pattern analysis (eqn. 1). The g(r) is an improved model of the k-function (eqn. 3), which not only eliminates the statistical cumulative error of the k-function, but also makes the distribution pattern easier to understand with a change of scale (Hao et al. 2007, Wiegand & Moloney 2013, Carrer et al. 2018, Muveengwi et al. 2018). It includes univariate distribution g1(r) and bivariate distribution g2(r) models. To ensure the correct use of the two models, it is necessary to know whether the sample site has a first-order effect caused by habitat heterogeneity (HP – Piao et al. 2013, Wiegand & Moloney 2013). We therefore analyzed the distribution pattern of large trees (DBH ≥ 20 cm) on a relatively large scale (to limit the edge effect, r = 0-30 m, should be less than 1/2 the length of the smallest plot boundary) using g1(r) and L(r) based on the null model of complete spatial randomness (CSR). L(r) (Lest) eliminates instability due to variance of the k-function, and is not very sensitive to changes in the distribution pattern at a small scale, being better able to reflect changes on a large scale (eqn. 3). In contrast, g1(r) is very sensitive to changes in the distribution pattern on a small scale (Wiegand & Moloney 2013). The results showed that there was no habitat heterogeneity in either forest (Fig. 3). To explore the biological interactions among trees, the spatial distribution of the community,

Fig. 2 - Tree height distribution in the vertical layers of the forest in both plots (plot 4, a; plot 5, b). The boundary of the vertical layers was close to 6 m. The understory mainly consisted of shrubs and a few tree species, while the upstory was dominated by several tree species. In plot 4, the proportion of the number and population for understory and upstory were 55.58%:44.42% and 84.61%:15.39%, respectively, and they were 53.44%:46.56% and 89.47%:10.53% in plot 5.

Fig. 3 - A pair correction function (PCF) and Lest heterogeneity test based on the null model of complete spatial randomness (CSR), indicating no significant heterogeneity in either forest stand. The red dashed lines are theoretical values, and the grey lines denote the 95% confidence limits of a Monte Carlo (MC) simulation. The black solid lines represent observed values and n is the number of large trees (DBH ≥ 20 cm) used to test heterogeneity.
upper layer, and lower on a slightly smaller scale \((r = 20\text{ m})\) was analyzed based on the null model of CSR \(g_0(r)\). We used \(g_0(r)\) to analyze the spatial correlation between the upper and lower layers, and the spatial correlation among their main populations \((n \geq 40)\) based on a null model of the independence of components (eqn. 1, eqn. 2, eqn. 3).

\[
g(r) = \frac{K(r)}{2\pi r} \\
K(r) = \frac{1}{N} \sum_{i=1}^{N} \sum_{j=1}^{N} w(d) \\
\hat{L}(r) = \frac{K(r)}{\frac{1}{N} \sum_{i=1}^{N} \sum_{j=1}^{N} w(d)} - r
\]

where \(N\) is the number of trees, \(r\) is the radius, \(w(d)\) is the edge-correction weighting, and \(A\) is the area of sample.

### Spatial distribution of species and sizes

Tree species and size differentiation are also closely related to scale (Hui & Pomerening 2014, Chalandyayashi et al. 2017). In this study, we used Stoyan’s generalized model \(k_i(r)\) of the mark correlation function (MCF) based on a point process, \(X_i\) to analyze changes in tree species and DBH with changes in the observation scale (eqn. 4). The \(k_i(r)\) is very flexible and can measure the spatial correlation of the mark of the \(i^{th}\) and \(j^{th}\) neighbors at a distance \(r\). The mark used could be one of several classification variables (tree species in this study) or numeric variables (DBH in this study). An arbitrary \(f\) is often used to test the \(k_i(r)\) function. It has two parameters \((m_i, m_j)\), which are the corresponding possible marks. When the mark is DBH, \(f(m_i, m_j) = m_i \times m_j\), and when tree species are the mark, \(f(m_i, m_j) = 1(m_i = m_j)\). In eqn. 4, \(E_i\) is the conditional expected value of the \(i^{th}\) and \(j^{th}\) trees at distance \(r\) in the point process \(X_i\) and \(M(i)\) and \(M(j)\) are the marks attached to the \(i^{th}\) and \(j^{th}\) trees. In the denominator, \(M\) and \(M'\) are random markers independently drawn from the edge distribution of the mark, and \(E\) is the common expected value.

When the mark is attached to the point process \(X_i\), it is independent and has the same distribution, \(k_i(r) = 1\) (eqn. 4):

\[
k_i(r) = \frac{E[f(M(i), M(j))]}{E[f(M', M')]} \quad (4)
\]

In the above analysis, the Monte Carlo (MC) method was used to randomly simulate the observation values of all observation scales 199 times, and the five maximum and five minimum values were obtained as the upper and lower 95% confidence intervals, enabling the significance of the deviation of the observed value from the theoretical value to be determined. For the spatial pattern, the \(g_0(r)\) theoretical value of a Poisson’s homogeneity test is 1 and the \(g_0(r)\) theoretical value is \(\lambda\). An observed value outside the upper confidence limit indicates an aggregated distribution or positive association. An observed value falling outside the lower confidence limit indicates a regular distribution or negative association. An observed value within the envelope between the confidence intervals indicated a random distribution and random association (Mugnogvi et al. 2018). For the MCF, when calculating tree species, \(k_m(r) = 1\) means that tree species are independent from each other at the scale \(r\), i.e., the probability that the neighbor \(j\) of any tree \(i\) is of the same species or a different species is equal. If \(k_m(r) > 1\) there is a positive correlation, i.e., conspecific aggregation or heterogeneity repulsion, and if \(k_m(r) < 1\) there is a negative correlation, i.e., heterogeneity aggregation or conspecific repulsion. For tree DBH, \(k_m(r) > 1\) indicates that the tree sizes at the scale \(r\) are independent from each other, and \(k_m(r) < 1\) indicates that the DBH of the two trees is positively correlated. If \(k_m(r) > 1\), the DBH of the two trees is considered to be negatively correlated, i.e., the DBH product of the two trees at the scale \(r\) is less than the mean DBH of all trees (Mugnogvi et al. 2018). The circle radius \(r\) simulated by the summary function was 20 m. For statistic convenience, when analyzing the spatial correlation between the main population of the upper layer and the main population of the lower layer, we set the distance \(r\) to 1 m, while the distance for other analyses was 0.001 m. Data analysis and graphical processing were conducted in the “ggplot2” and “spatstat” packages of R ver. 3.2.5 (R Development Core Team, Vienna, Austria).

## Results

### Spatial patterns in the vertical layers

In plot 4, the clumped pattern of all populations extended for 14 m and then became a random distribution at larger scales (Fig. 4a). The distribution pattern of the upstory also gradually shifted from an aggregated distribution at larger scales to random distribution as the observation scale increased (Fig. 4b). Except for the random pattern at \(r = 18-20\text{ m}\), the lower layer was aggregated at residual scales \((r = 0.17\text{ m} - 4.0\text{ m})\). In addition, the upper and lower layers were randomly correlated at all scales (Fig. 4d). The trees in plot 5 were slightly aggregated at the small scales \((r = 0-0.10\text{ m})\), but the distribution became random at a large scale (Fig. 4e). The upper layer was close to a random distribution at small scales \((r = 0-4\text{ m})\), but at larger scales \((r = 4-20\text{ m})\)

![Fig. 4 - Spatial distribution and spatial correlation of the upstory and understory in both plots. The grey background in each figure denotes the 95% confidence limits of an MC simulation. The black solid line represents the observed value and the red dashed lines are theoretical values.](image-url)
there was a low level of aggregation (Fig. 4f). The aggregation intensity of the understory was significantly higher than that of the upper layer (Fig. 4g), but the upper and lower layers were randomly correlated at all scales (\( r = 0-20 \) m – Fig. 4h).

Spatial correlation of the main populations in the vertical layers

The populations of the main species of the upper layer of the two stands were mostly randomly correlated, and there were no positive correlations. The upper layer of plot 4 was all randomly correlated (Fig. 5a), but in plot 5 there was a weak negative correlation (17.0%) at some scales (\( r = 0-2, 5-7, 12-14, \) and 15-17 m – Fig. 5d). Although plot 5 was mostly random at small scales, there was a negative correlation (8.0-33.0%) between species at medium and large scales (Fig. 5f).

Spatial distributions of tree sizes

In plot 4, small trees were aggregated together at small to medium scales (\( r = 0-14 \) m), while trees with different DBH values were located independently of each other at the large scale (\( r = 14-20 \) m – Fig. 7a)....
upper layer was similar, except that the trees with different DBH values were located independently of each other when r = 12 m (Fig. 7b). In contrast, the observed values in the lower layer were closer to the 95% confidence limits of a MC simulation. The black solid line represents the observed value and the red dashed lines are theoretical values.

**Discussion**

**Response of the distribution pattern to stratification**

An aggregated distribution of species is a common pattern in nature, especially in the early stages of succession (Salas et al. 2006, Li et al. 2009, Wang et al. 2018). The forests investigated in this study originated from land where a virgin P. yunnanensis forest was fully cut. Site conditions after clear cutting tend to be homogeneous and only light-dependent species (e.g., *P. emblica*, *V. bracteatum*, *C. stellatum*, and *W. uvarifolia*) appeared in the gap with the improvement of habitat conditions, and some of them also become mature. Their limited height and production of succulent fruit further limited the propagation distance of the seeds (Tab. S1 in Supplementary material), further strengthening the aggregation of the understory (Zhang et al. 2010, Torimaru et al. 2013). Habitat heterogeneity on a small scale could not be excluded by the current Poisson heterogeneity process model and it may affect species composition and distribution (Beatty 1984, Torimaru et al. 2013). Moreover, seasonal drought in the study area may be another reason for some populations to gather together (Li & Wang 1981). In the same subtropical region, secondary forest in the Dinghu Mountains in Guangdong Province and the Yingzuijie Huitong National Forest Reserve in Hunan Province also had similar distribution patterns in the early stages of succession (Li et al. 2009, Xiang et al. 2013). Species richness and diversity may also be related to the overall distribution pattern of the two forests (Li et al. 2009, Hui & Pommelering 2014).

The distribution pattern of the upper forest layer at each scale was significantly closer to a random distribution than that of the lower layer (Fig. 4b, Fig. 4f vs Fig. 4c, Fig. 4g). This phenomenon is widespread in many forests in tropical, subtropical, and temperate regions (Kuuluvainen et al. 1996, Salas et al. 2006, Li et al. 2009, Li et al. 2012, Omelko et al. 2018). This may be the result of the greater competitive pressure between plants in upper layers than in lower ones (Picard et al. 2009). Inter/intraspecific competition for habitat resources (e.g., light, water, and nutrition) leads to self-thinning (Hao et al. 2007, Piao et al. 2013), destroying the pattern originally generated by the restrictive seed distribution and resulting in a tendency toward a non-aggregated distribution (Torimaru et al. 2013, Kuninaga et al. 2015). This phenomenon is widespread in many forests in tropical, subtropical, and temperate regions (Kuuluvainen et al. 1996, Salas et al. 2006, Li et al. 2009, Li et al. 2012, Omelko et al. 2018). This may be the result of the greater competitive pressure between plants in upper layers than in lower ones (Picard et al. 2009). Inter/intraspecific competition for habitat resources (e.g., light, water, and nutrition) leads to self-thinning (Hao et al. 2007, Piao et al. 2013), destroying the pattern originally generated by the restrictive seed distribution and resulting in a tendency toward a non-aggregated distribution (Torimaru et al. 2013, Kuninaga et al. 2015). This may be an important reason for their random association. In contrast, if two populations compete for space in the canopy, they are expected to be negatively correlated (Petritan et al. 2014). The spatial correlation between the upper and lower layers and their main populations was very weak (Fig. 4d, Fig. 4h; Fig. 5c, Fig. 5f). This was very similar to a reported relationship between the upper and lower layers of a secondary forest formed by cutting a Cunninghamia lanceolata plantation in the same period (year 1950). The lower layer was completely random at observation scales up to 0-17 m (Xiang et al. 2013). As mentioned above, the upper and lower layer species differ in light dependence, which may determine the spatial association between large and small trees. Similar phenomena exist between some other natural forest populations (for example, the...
Quercus petraea and Fagus sylvatica mixed forest in Europe – Petritan et al. (2014). The random correlation between the upper and lower layers at small scales implies that the lower layer will exist for a long time (Xiang et al. 2013).

At all observation scales, the main populations in the understory in both forests were mostly randomly correlated, i.e., there was almost no competition or dependency (Fig. 5b, Fig. 5e). This may be related to the degree of intraspecific aggregation. Intraspecific aggregation implies a significant reduction in the probability of interspecific separation. In an old temperate forest, Zhang et al. (2010) found that six congeneric maples with different diameters had random spatial correlations. In vegetation community succession, species interactions promote random spatial patterns. This has been confirmed by actual survey data (Zhang et al. 1999) and software simulations (Uria-Diez 2005). However, the random association of the understory found in this study may be a transient phenomenon, because many understory species only occur in the early stages of succession and are eventually replaced by more shade-tolerant species. Nonetheless, it is well-known that vertical stratification of forests helps to reduce competition and promote coexistence (Nishimura et al. 2002).

Response of species spatial distribution to stratification

Conspecies tend to aggregate in nature. This is generally considered to be the combined result of seed dispersion restrictions, habitat flittering, habitat heterogeneity and ecological preferences (Li et al. 2009, Wang et al. 2018). Our study showed that conspecific aggregation not only occurred at the community level; the vertical layers in the early stage of succession also had aggregation characteristics, and their intensity decreased with an increase in observation scale (Fig. 6a-f). This trend is consistent with the results of previous studies (Hui & Pomerening 2014, Pomerening & Uria-Diez 2017, Wang et al. 2018). Although the total number of trees in the upper and lower layers was similar, the degree of aggregation in the upper layer was less than that in the lower layer (Fig. 6b, Fig. 6e vs. Fig. 6c, Fig. 6f). Pomerening & Uria-Diez (2017) also found that species segregation increased with an increase in tree diameter in stands that had suffered different degrees of disturbance and were located in different zones, i.e., the probability of another tree species surrounding the tree was greater for a large tree than for a small tree. Other studies have also shown that tree size (DBH/U, dominance) and the degree of species mixture are positively correlated in natural communities (Li et al. 2012, Li et al. 2017) and their dominant populations (Wang et al. 2018). Some studies consider this phenomenon to be a negative density effect (Piao et al. 2013, Wang et al. 2018). Other studies explain it in terms of niche complementarity. The number of large trees in a community is limited, and higher species richness promotes tree size differentiation. Differences in attributes of different species can result in the available growth space of a woodland being more effectively filled (Hui & Pomerening 2014, Pomerening & Uria-Diez 2017, Wang et al. 2018). These effects may have existed simultaneously in the plots studied here, but the niche complementarity is more important.

The understorey patterns of the upper layer results in patchy light conditions in the forest, and light-dependent species become established in locations with less shade, forming an aggregated distribution (Suzuki et al. 2008). Some populations (e.g., V. bracteatum and C. stellatum) contain a number of dead small trees (Tab. S1 in Supplementary material), which provides opportunities for the arrival and survival of other species. This also results in the pattern of surviving dominant trees having a regular distribution, with more dissimilar adjacent trees (Suzuki et al. 2008, Pomerening & Uria-Diez 2017, Wang et al. 2017). Other shade-tolerating clumped populations (e.g., A. quinquegona) and newly emerging tree species (e.g., T. ciliata, Diospyros kaki silvestris) do not display growth decay or death, and their appearance is more consistent with the habitat. Therefore, we believe that the ecological effects of forest communities are related to the specific populations and developmental stages.

Response of the spatial distribution of tree size to stratification

Asymmetric competition results in size differentiation and death of adjacent trees. Therefore, tree size is often used to assess the intensity of competition between adjacent trees and to predict changes in the community distribution pattern (Suzuki et al. 2008, Potvin & Dutilleul 2009, Wang et al. 2017). In this study, the product of the paired diameters within the small range of the upper layer was far less than the expected value (Fig. 7b, Fig. 7e), indicating that individual sizes varied greatly and small trees clumped together. Smaller individuals may grow more slowly than larger trees, which may accelerate the differentiation of tree sizes (Ford 1975). In contrast, a size difference between individuals in the lower layer was not obvious (Fig. 7c, Fig. 7f). This layer was composed of multiple shrubs and small tree species (Fig. 2a, Fig. 3b), indicating that competition between adjacent trees was not strong. Differences in size differentiation of the upper and lower layers also indicated that the upper layer experienced a greater competitive pressure than the lower layer, which was consistent with the distribution pattern and the size differentiation process reported in an artificial forest. Size differentiation becomes more obvious as the succession time increases (Ford 1975). There was almost no difference in the size differentiation of the upper layers at small observation scales between both plots (Fig. 7b, Fig. 7e), but their understoreys differed from each other (Fig. 7c, Fig. 7f). The size differentiation was more obvious in the stand with a higher density (Tab. S1 in Supplementary material). It is likely that the average individual in the denser plot occupied more space and had a relatively strong demand for environmental resources. Competition will lead to size differentiation (Li et al. 2012). This is consistent with conspecies having more clumped distribution (Fig. 6c). In monoculture plantations, it is generally believed that competition facilitates the differentiation of tree size in crowded communities (Ford 1975, Weiner et al. 2001). However, in a natural forest with multiple tree species, the interaction between adjacent trees is far more complicated than in artificial plantations (Potvin & Dutilleul 2009). In addition to density factors, forest regeneration and the death of crushed trees are also factors related to size differentiation (Wang et al. 2018). Differences in the tolerance of adversity, life-history and functional traits between species may also be important for the differences in size differentiation between two forests. This was considered as a possibility given that the tree species composition of the lower layer in the two plots was not identical (Fig. 2a, Fig. 2b).

List of abbreviations

The following abbreviations have been used throughout the manuscript:
- MCF: mark correlation function
- DBH: diameter at breast height
- PEF: pair correction function
- HP: habitat heterogeneity
- CSR: complete spatial randomness
- MC: Monte Carlo

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References


Supplementary Material

Fig. S1 - The location of our study sites, Yachang Township, Leye County, Baise City, Guangxi Zhuang Autonomous Region.

Fig. S2 - The secondary pine-oak forests subjected to clear-cutting 57 years ago currently displays two layers: the understory species mainly consists of evergreen broad-leaved shrubs, while the upstory mainly contains several deciduous broad-leaved species and P. yunnanensis.

Tab S1 - The parameters of main populations occurred in our sampled plots (the number of living trees ≥ 40) and their ecological characteristics.

Link: Li_2975@suppl001.pdf