

Species diversity and stand structure as drivers of canopy complexity in southern African woodlands

Abstract

1 Introduction

The characterization of tree canopy structure in wooded ecosystems constitutes a long-standing field of research that has been fundamental to interpreting, modelling, and improving understanding of ecosystem function (Watt, 1947; Whittaker and Woodwell, 1969; Horn, 1971; Maarel, 1996). Canopy structure describes the spatial distribution and density of canopy foliage, comprising the primary interface between trees, the atmosphere and sunlight. Canopy structural complexity, i.e. the spatial heterogeneity of foliage distribution through the canopy, has been positively linked to canopy productivity (Hardiman2011; Chen2012; Law2001; Baldochii2001; Morin2015). It is therefore essential to understand the drivers of variation in canopy structure to improve modelling of earth-atmosphere carbon fluxes and community assembly.

At continental scales, variation in canopy height and canopy cover, two coarse measures of canopy structure both of which have been shown to affect woody productivity and correlate with woody biomass (), can largely be explained by climate and edaphic data (GEDI). Increased resource availability allows for larger trees and more closed canopies (). At the scale of a single tree community however, where variation in climate and soil may be negligible, variation in canopy structure is thought to be affected principally by an interacting combination of tree canopy species composition (), and disturbance history (). However, empirical testing of these mechanisms thought to drive canopy structure in natural wooded ecosystems remains sparse across many biomes ().

Following established biodiversity-ecosystem function theory, the niche partitioning of canopy space, i.e. the spatial complementarity of individual tree canopies, is thought to be a key mechanism underlying positive biodiversity-productivity effects in wooded ecosystems (Pretzsch, 2014; Barry et al., 2019). Biodiversity-ecosystem function theory predicts that crown complementarity and thus canopy complexity and foliage density should increase with tree diversity in the local neighbourhood, increasing standing biomass and woody productivity, as coexisting species must occupy non-identical niche space to avoid competitive exclusion (Gadow1993).

As well as the species diversity of trees in a local neighbourhood, the spatial distribution and relative size dominance of those trees, i.e. stand structure, is also expected to affect crown structural complexity. Increased heterogeneity in tree size, whether a result of species diversity, disturbance history or some other factor, is expected to increase crown complexity and overall canopy density

as individuals of different sizes occupy different parts of the vertical canopy space (). Additionally, clustering of individuals in space is expected to increase canopy structural heterogeneity across a stand, but ultimately decrease total foliage density due to an increase in competitive interactions (). Clustering may occur as a result of disturbance history, or as a result of strong facilitation effects among individuals in a hostile environment (**Ratcliffe2017**).

While much work in the field of forest management has been done to test biotic drivers of tree canopy structure in temperate () and boreal forests (), similar work in the tropics is comparatively scarce (). In dry tropical woodlands and savannas especially, tree canopy structure and its effect on ecosystem productivity has received little attention, possibly due to the misplaced assumption that woody productivity in these ecosystems does not represent a globally significant carbon flux (), or that tree canopies in these smaller stature woodlands do not interact and compete for resources to the same degree as in large stature forests (). In recent years however, it has been shown that dry tropical woodlands represent the largest uncertainty in our estimates of the terrestrial carbon cycle (Quéré et al., 2018; Ahlstrom et al., 2015). Sitch et al. (2015) demonstrated the dominant role of the dry tropics in driving variability in the terrestrial carbon sink, and showed that the dry tropics are the fastest increasing component of the terrestrial carbon sink. Part of this uncertainty arises from our lacking a nuanced understanding of how species composition and structure affect ecosystem function in these ecosystems, which conceptually underpins the Dynamic Global Vegetation Models (DGVMs) fed into global carbon dynamics models. The pertinence of this knowledge gap has prompted further research of the biotic drivers of variation in productivity in the dry tropics, and momentum in this field of research is building, to create general theories of the ecosystem function of the dry tropics ().

Canopy structure is multi-dimensional and has previously been explained using a plethora of simple metrics that originated in forest and community ecology (). Assessments of canopy structure in the dry tropical have most often modelled tree canopies as a series of ellipses (2D) or ellipsoids (3D) based on field measurements with measuring tapes (). Measurements of this kind are time consuming and yet are an over-simplification of canopy structure (). Alternatively, canopy cover is often measured using indirect optical methods which partition sky from canopy material, i.e. with hemispherical photography or the commonly used LAI-2000, providing a 2D representation of the canopy but lacking information on vertical canopy structure. In recent years, particularly in temperate and boreal forests, LiDAR (Light Detection And Ranging) has emerged as a suitable technology for rapidly and precisely assessing canopy structure in 3D, conserving information on 3D structure of the calibre that is required to understand its complexities ().

In this study we applied terrestrial LiDAR techniques to woodland-savanna mosaics at two sites in southern Africa, with the aim of increasing our understanding of how various metrics of tree

66 canopy structural complexity are affected by tree neighbourhood diversity and stand structure.
67 Our overarching contention is that neighbourhoods of greater tree diversity and greater structural
68 diversity allow greater canopy complexity and foliage density, resulting in higher productivity, and
69 ultimately a more ‘forest-like’ community, rather than an open canopy savanna.

70 **2 Materials and methods**

71 **2.1 Study sites**

72 Measurements were conducted at two sites, the first in Bicuar National Park, southwest Angola
73 (S15.1°, E14.8°), and the second in and around Mtarure Forest Reserve, southeast Tanzania (S9.0°,
74 E39.0°) (Figure 1). At each site, 1 ha (100x100 m) plots were sited in areas of miombo woodland
75 vegetation, across a gradient of stem density. In Angola, 15 plots were sampled, while in Tanzania,
76 seven were sampled following the curtailment of fieldwork due to COVID-19 travel restrictions.
77 Fieldwork was conducted between February and April at both sites, during the peak growth pe-
78 riod of each site in order to capture the highest foliage volume in the canopy.

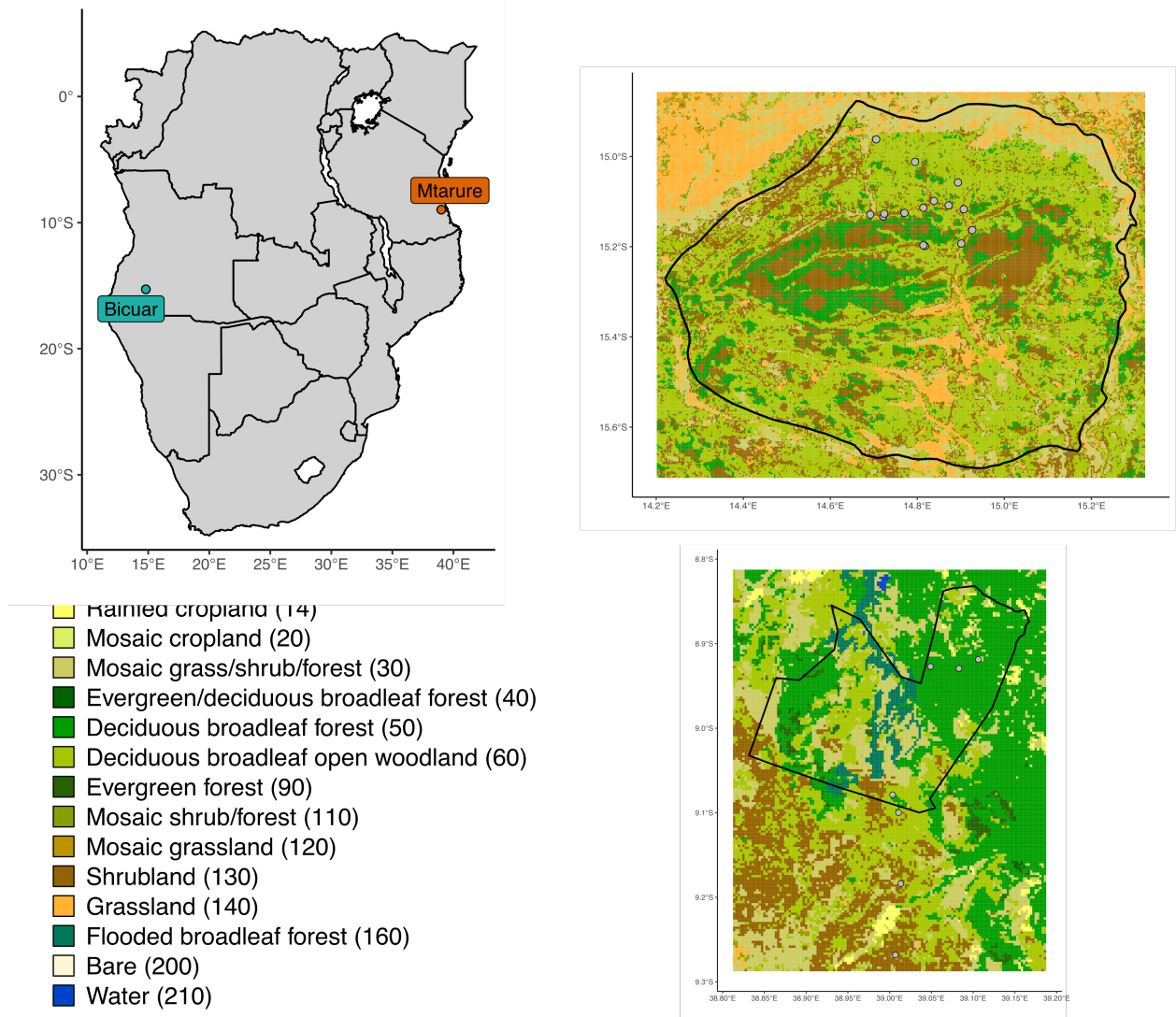


Figure 1: Location of study sites within southern Africa (a), and of 1 ha plots within each site. The blue polygons denote the boundaries of protected areas which encompass the majority of study sites, Bicular National Park in Angola (b), and Mtarure Forest Reserve in Tanzania (c). The background of each site map is a re-classified version of the GlobCover global land cover classification (**Globcover**).

79 2.2 Field measurements

80 Within each 1 ha plot we identified each stem >5 cm stem diameter to species, measured stem di-
 81 ameter (diameter at breast height - 1.3 m) and recorded stem location within the plot using tap
 82 measures. Each 1 ha plot was further subdivided into nine 10 m diameter circular subplots ar-
 83 ranged in a regular grid, with a 15 m buffer from the plot edge and 35 m between subplots. For
 84 each subplot, we identified woody stems >5 cm diameter with canopy material inside the subplot.
 85 We measured the distance and direction from the subplot centre of each of these stems.
 86 Within each subplot, a variable number of scans were recorded using a Leica HDS6100 phase-shift

87 terrestrial laser scanner (TLS) (**Leica**). The number and position of scans within a subplot was
88 determined by the arrangement of canopy material in the subplot. Scan positions were arranged
89 to minimise shadows within the canopy of the subplot, and to maximise canopy penetration. The
90 number of scans per subplot ranged between one and five in both sites.

91 **2.3 Data analysis**

92 **2.3.1 Scan processing**

93 Point clouds from scans in each subplot were registered and unified using Leica Cyclone (version
94 9.1) (**Cyclone**), using cross-pattern reflective targets visible to all scans. Point clouds were vox-
95 elised to cubic voxel sizes of different sizes depending on the application of the data. For subplot
96 height profile estimation and gap fraction we used 5 cm³ voxels, and for whole plot canopy ru-
97 gosity we used 10 cm³ voxels. Voxels were classed as filled if they intersected with one or more
98 points. Variation in voxel size reflects the spatial scale of each analysis, and is bounded by the
99 beam divergence of the scanner over longer distances (). Choosing voxels that are too small can
100 result in pock-marked representations of surfaces that are especially problematic when calculat-
101 ing larger scale canopy structure metrics, such as canopy top roughness, while voxels that are
102 too large can result in an over-estimation of plant volume when estimating canopy foliage den-
103 sity at the subplot scale (Seidel, Fleck, and Leuschner, 2012; Cifuentes et al., 2014). We used a
104 noise reduction algorithm from to discard points based on mean nearest neighbour distances.
105 This effectively removed ‘ghost points’ produced by partial interceptions and also removed many
106 erroneous returns caused by airborne dust particles, which was common at our study sites. Raw
107 points clouds for each subplot had ~2.9e+08 points, ~4.5e+07 points after voxelisation, and ~2.1e+07
108 points after noise reduction.

109 Ground points were classified using the Progressive Morphological Filter (PMF) from Zhang et al.
110 (2003). Point cloud height was reclassified height based on this revised ground layer by measuring
111 the vertical distance between the nearest ground point and each point.

112 We used ray-tracing to calculate canopy cover at the subplot centre from multiple TLS scans.
113 Hemispherical images were created using the POV-ray software (). Voxels were converted to matt
114 black cubes filling the voxel volume, with a white sky box and no light source. A ‘camera’ with a
115 180° fisheye lens was placed at the subplot centre within POV-Ray, at a height of 1.8 m pointing
116 directly upwards. The images produced by POV-Ray were analysed using Hemiphot (**Steege**) to
117 estimate canopy cover as the proportion of pixels filled by canopy material.

118 2.4 Stand structure

119 For each subplot, we calculated an adapted version of the Hegyi index to estimate crowding, as
120 an alternative to stem density that works better to describe stand structure at small spatial scales
121 (Hegyi1974). We also calculated the coefficient of variation of stem diameter as a measure of the
122 heterogeneity of tree size in the neighbourhood.

123 At the plot level, we estimated the regularity of species spatial distribution using the spatial min-
124 gling index (Gadow2002). We also measured whole plot stem distribution using the winkelmass,
125 which measures the degree of clustering of stems (). Finally, we calculated stem density per plot.

126 2.5 Statistical analysis

127 Linear mixed effects models tested the effects of tree species diversity and stand structure on canopy
128 structure. Two sets of models were conducted, the first at the subplot level with random effects
129 for plot nested within site, and the second at the plot level with random effects for site only. Sep-
130 arate models were fitted for each canopy structure variable, resulting in six models at the subplot
131 level and five models at the plot level.

132 To explore variation in tree species composition among plots and sites, we conducted a Non-metric
133 Multi-dimensional Scaling (NMDS) analysis using tree species abundance in each plot. We ex-
134 cluded species with only one individual across all plots.

135 3 Results

136 3.1 Vertical canopy complexity

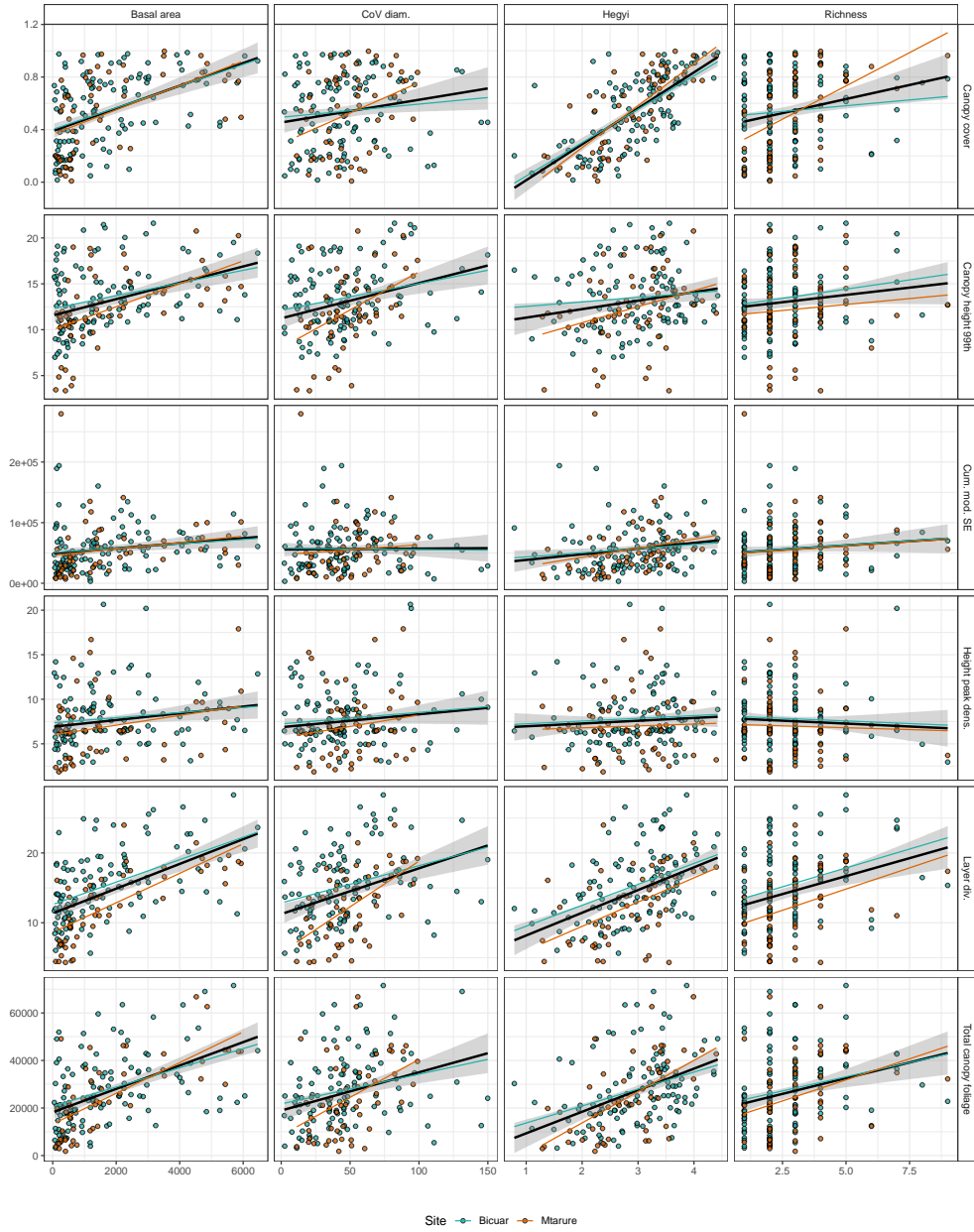


Figure 2: Bivariate relationships between subplot canopy structure metrics (y axis) and diversity and stand structure metrics (x axis). Points and linear model lines of best fit are coloured by site. The black line of best fit is a linear model including both sites. See for comparison of linear model fits by site.

137 Bivariate plots show that species diversity, measured by species richness of the tree neighbourhood
 138 around each 10 m diameter subplot, appears to have weak but positive effects on canopy layer di-
 139 versity and total canopy cover (Figure 2). Stand structural metrics have much stronger positive

140 effects on canopy structure, except for uniformity of foliage distribution and height of peak den-
 141 sity. The two sites had very similar bivariate relationships in all cases.

142 Linear mixed effects models showed that species richness of the subplot neighbourhood had vari-
 143 able effects across different measures of canopy structure, but the effect sizes were not signifi-
 144 cant (slope standard errors not overlapping zero) for any model (Figure 3). On the other hand,
 145 stand structural metrics, the Hegyi index, Coefficient of Variation of stem diameter and total stem
 146 basal area, had a much greater effect on canopy structure variables. The Hegyi index had a strong
 147 positive significant effect on canopy cover, while basal area had positive significant effects on to-
 148 tal foliage density, layer diversity, and canopy height, with strong non-significant effects on non-
 149 uniformity of vertical foliage distribution and the height of peak foliage density.

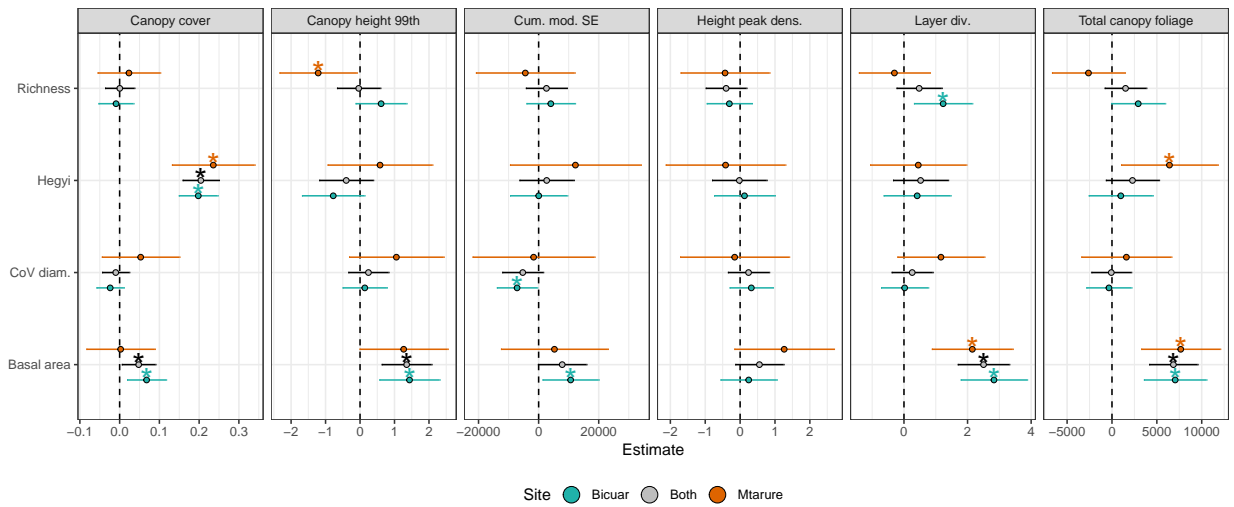


Figure 3: Standardized fixed effect slopes for each model of a canopy structure metric. Slope estimates are ± 1 standard error. Slope estimates where the interval (standard error) does not overlap zero are considered to be significant effects. Points are colour

150 The model selection process showed that basal area was the best predictor of all canopy structural
 151 metrics, but the best model for layer diversity also included species richness and the best models
 152 for total canopy foliage and canopy cover both included the Hegyi index.

Response	Richness	Hegy	CoV diam.	Basal area
Layer div.	✓			✓
Total canopy foliage		✓		✓
Height peak dens.				✓
Canopy height 99th				✓
Cum. mod. SE				✓
Canopy cover		✓		✓

Table 1: Explanatory variables included in the best model for each canopy structure variable.

153 3.2 Canopy rugosity

154 Similar to the subplot analyses, at the whole-plot scale, tree species diversity, measured here by
 155 the shannon index, tended to have weak positive effects on canopy complexity metrics, while stand
 156 structural diversity metrics had stronger positive effects (Figure 5). Strong positive relationships
 157 of basal area on canopy complexity are driven mostly by two plots with particularly low basal area
 158 in Mtarure. These plots are sparse thorny savanna, dominated by *Senegalia* spp. (Figure 4). In-
 159 deed, linear models using only Bicular plots show divergent relationships. These two plots also
 160 have particularly low canopy cover, canopy height, and canopy top roughness, despite having simi-
 161 lar tree species richness and spatial distribution of trees (winkelmass) as other plots.

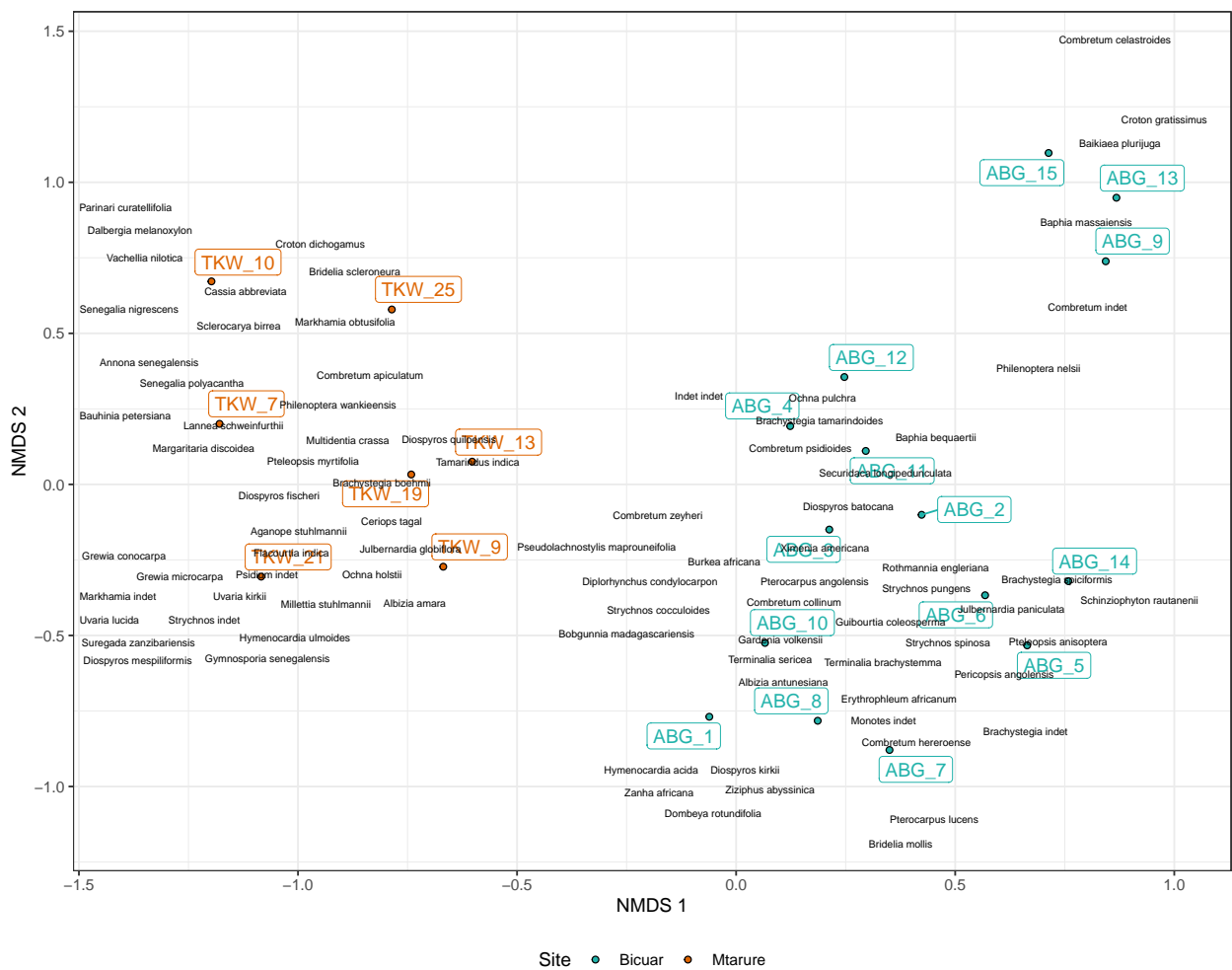


Figure 4: The first two axes of a Non-metric Multi-Dimensional Scaling (NMDS) analysis of tree species diversity in each plot. Species scores are labelled as black text, while plot scores are labelled as coloured points. Plots can be split into three principal groups: 1) ABG_9, ABG_13 and ABG_15, dominated by *Baikiaea plurijuga*; 2) the other Bicar plots plus TKW_9, TKW_13, TKW_19 and TKW_21, dominated by *Julbernardia* spp., *Brachystegia* spp. and *Ochna* spp.; 3) TKW_7, TKW_10 and TKW_25, dominated by *Senegalia* spp. and *Vachellia* spp..

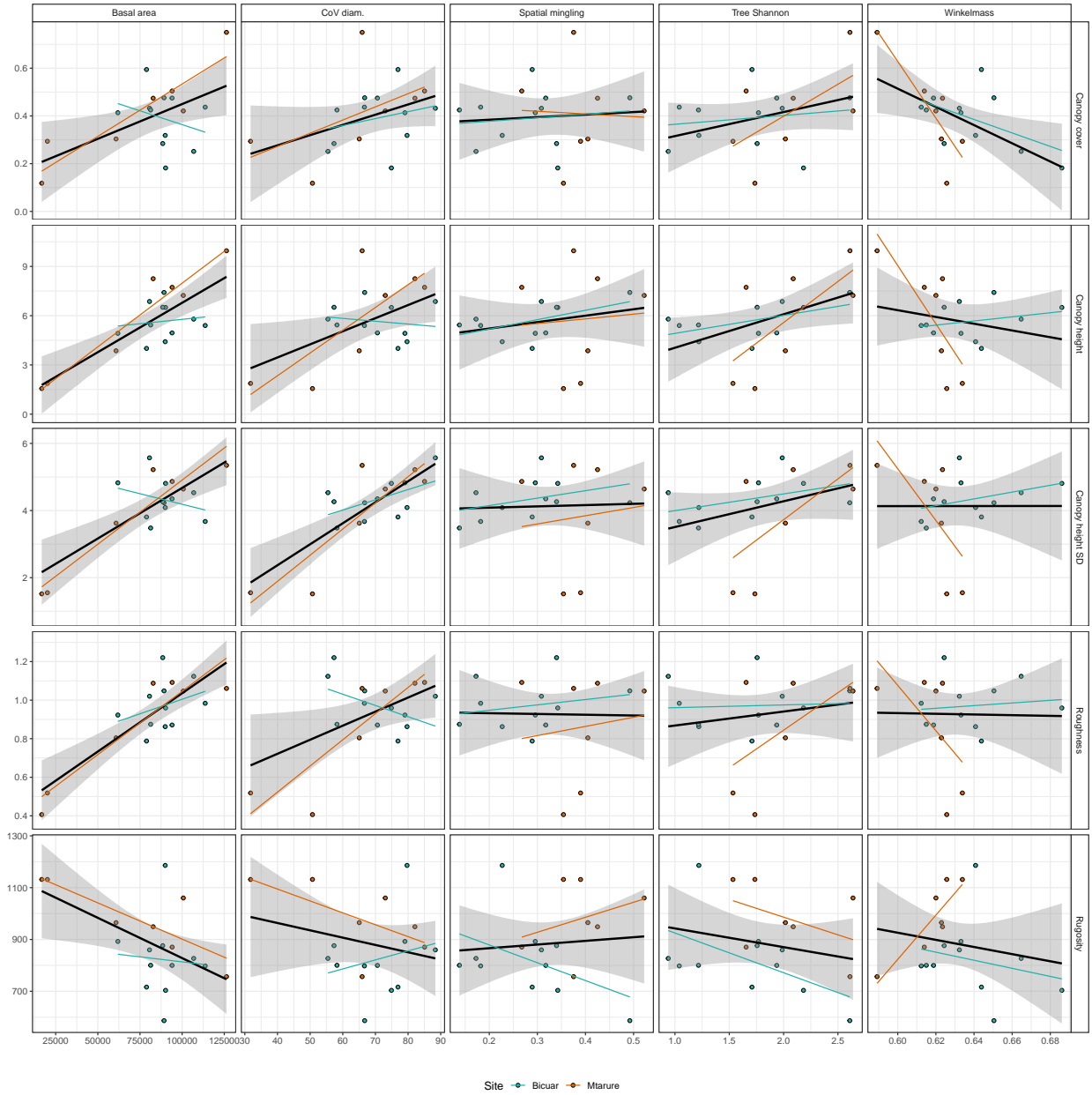


Figure 5: Bivariate relationships between diversity and stand structure metrics (x axis) and whole-plot canopy structure metrics (y axis). Points and linear model lines of best fit are coloured by site. The thick black line of best fit is a linear model including both sites.

162 Maximum canopy height at the plot-level appears to be positively affected by both tree species di-
 163 versity and total stem basal area. Basal area has a positive effect on canopy top roughness. Varia-
 164 tion in canopy height was increased both by variation in stem diameter and total stem basal area.
 165 The only significant fixed effect on canopy cover at the plot level was the winkelmass, which mea-
 166 sures the spatial clustering of trees.

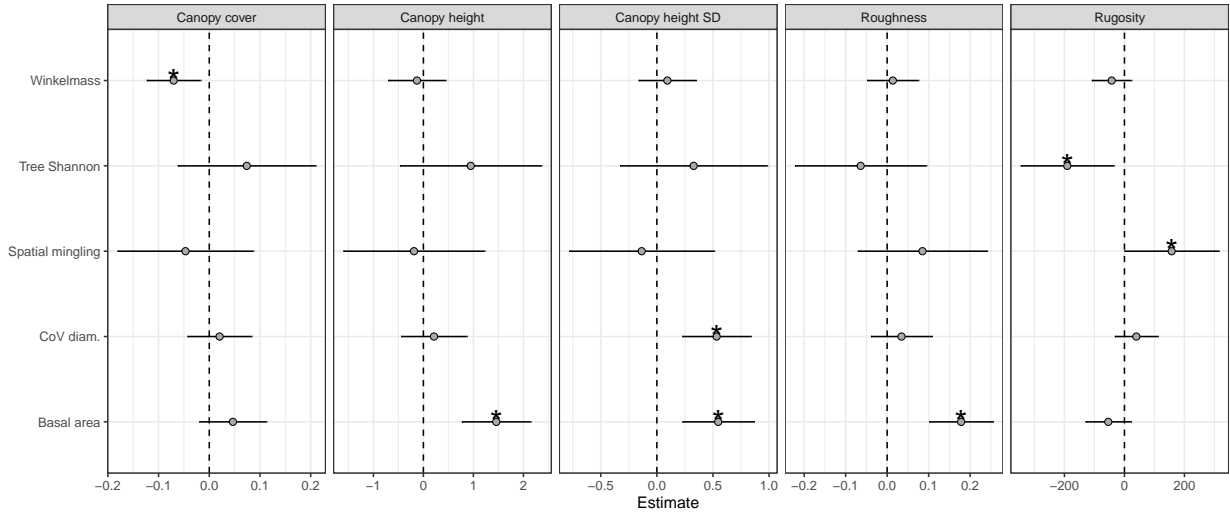


Figure 6: Standardized fixed effect slopes for whole-plot canopy rugosity. Slope estimates are ± 1 standard error. Slope estimates where the interval (standard error) does not overlap zero are considered to be significant effects.

3.3 Comparing subplot and plot measures of canopy structure

Plot-level and subplot-level canopy structure metrics were highly correlated in many cases (Figure 6). Plot canopy height especially, tended to be strongly positively correlated with subplot canopy complexity. Additionally, as canopy top roughness increases, many subplot canopy complexity and density metrics increase. In the majority of cases, both sites had similar correlations of subplot and plot measures of canopy structure, with notable exceptions for plot roughness vs. layer diversity, plot roughness vs. canopy cover, and plot canopy height vs. canopy cover.

Variance of plot canopy height and plot roughness was larger in Mtarure than Bicular. The increase in variance was caused by two particularly sparse thorny savanna plots in Mtarure, which had very low canopy height and roughness.

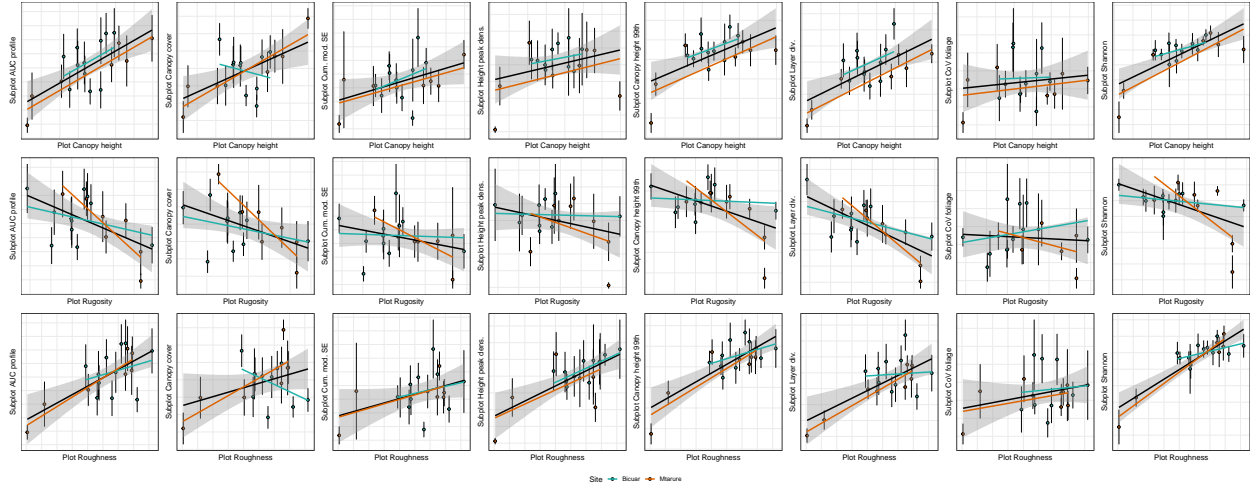


Figure 7: Bivariate plots of canopy structural metrics at the subplot and plot-level. Each point represents the mean values of a single plot. Points and linear model fits are coloured according to site. The black linear model combines both sites. Error bars on points are the standard deviation of subplot metrics.

4 Discussion

We investigated the effects of tree species diversity and structural diversity on several metrics of canopy structure that are hypothesised to affect plot productivity. Species diversity appeared to have weak positive effects on canopy complexity at both the subplot and plot scales, while stand structural diversity had greater effects.

With the whole plot effect of tree species diversity on canopy height, what we might be seeing is actually an increased canopy height allowing greater species diversity, rather than the other way round.

While there are reasons to think that stand structure would influence species diversity directly, we did not find evidence for this in our study. Correlations between CoV stem diameter and richness were weak **check**.

Canopy structure at the plot level was less well predicted by stand structure and species diversity than subplot level canopy structure.

Jucker2015 found that increased species diversity led to greater canopy packing in European forests, with trees in mixed forests having generally larger crowns. Our result that species richness correlated with greater canopy cover and canopy complexity supports this, though mixed models suggest that the role of richness is heavily tempered by stand structure and the spatial distribution of individuals. In disturbed woodlands, disturbance appears to be the primary determinant of

canopy complexity, while species richness plays a supporting role.

5 Conclusion

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