1	Differential patterns of taxonomic and functional diversity for two groups of canopy
2	arthropods across spatial scales
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21	Authorship: TOC originally conceived of the study and collected samples; all authors conceived
22	of current manuscript; HJP and KUC identified spiders and ants from samples; MBM conducted
23	statistical analyses; MBM, HJP, and KUC wrote the first draft of the manuscript; all authors
24	revised the manuscript
25	
26	Biosketch: This research team aims to better understand and quantify arthropod community
27	structure and diversity at landscape scales through local experiments and broad scale
28	observational approaches. They are interested in refining our understanding of the structuring of
29	biodiversity to improve conservation efforts.

## 1 Differential patterns of taxonomic and functional diversity for two groups of canopy

- 2 arthropods across spatial scales
- 3 **Running Title:** Scale impacts canopy arthropod diversity
- 4 **Key Words:** beta diversity, functional diversity, diversity partitioning, forest canopy,
- 5 Formicidae, Araneae
- 6 Abstract
- 7 *Aim*
- 8 Arthropod diversity is often linked to variation in resource use, dispersal ability, habitat
- 9 connectivity, and climate factors that differ across spatial scales. The aim of this research was to
- 10 examine how species richness, functional diversity, and community composition of two taxa
- 11 differing in functional roles and dispersal ability are structured across spatial scales and to
- 12 identify the importance of vegetation, climate, and landscape in explaining these patterns at
- 13 different scales.
- 14 *Location*
- 15 96 trees in 24 stands of 6 deciduous forest sites in 2 ecoregions of the eastern USA (North-
- 16 Central Till Plain and Western Allegheny Plateau)
- 17 Time period
- 18 2000
- 19 Major taxa studied
- 20 Canopy dwelling ants (Hymenoptera: Formicidae) and spiders (Araneae)
- 21 *Methods*
- 22 Organisms were collected from tree canopies using insecticidal fogging. Ant and spider
- 23 taxonomic and functional beta diversity were partitioned across four hierarchical spatial scales

24 (individual tree, forest stand, site, and ecoregion). The contribution of climactic, landscape, and
25 vegetation variables was determined using model selection.

26 Results

27 Ant and spider species richness, functional diversity, and community composition differed 28 between taxa and across spatial scales. Alpha diversity (within trees) was lower than expected for 29 both taxa and types of diversity, with host tree species supporting different species of ants and spiders. While beta components of species diversity among trees and forest stands was greater 30 31 than expected for both taxa, spiders also showed significant levels of beta diversity among sites. 32 Functional beta diversity was less scale-dependent than taxonomic beta diversity. Stand-level patterns of beta diversity were significantly predicted by variation in climate and landscape 33 34 connectivity. 35 Main conclusions Effects of climate and landscape fragmentation on the diversity and community structure of both 36 taxa indicate that anthropogenic climate change and land use change will alter canopy arthropod 37 communities. Results also suggest that patterns of diversity among fragmentation metrics is 38

39 influenced by differences in dispersal ability.

## 40 **1 INTRODUCTION**

41 Understanding drivers of species diversity and distributions are primary goals of ecology 42 and can be used to inform conservation efforts. Measures of biodiversity span local (alpha 43 diversity,  $\alpha$ ) and regional (gamma diversity,  $\gamma$ ) scales. Studies of the variation, or turnover, in 44 biodiversity among sites or communities (beta diversity,  $\beta$ ), first proposed by Whittaker (1960), 45 have exploded over the last few decades (Anderson et al., 2011; Socolar et al., 2016; Mammola 46 et al., 2021). Beta diversity has been applied to the spatial scaling of biodiversity, temporal 47 change in communities for conservation monitoring, changes across latitudinal and elevational gradients, and environmental filtering of functional traits and subsequent assembly processes 48 (Crist et al., 2003; Kraft et al., 2011; Barton et al., 2013; Siefert et al., 2013; Jarzyna & Jetz, 49 50 2018).

51 Beta diversity has been primarily focused on measures of taxonomic diversity but not functional diversity (i.e. based on the ecological roles and traits of species). To a certain extent, 52 species diversity and functional diversity are correlated, but not reliably so, as not all species are 53 54 equivalent in a given system and their relative roles may change depending on the presence of 55 and interactions with other species. Species and functional diversity are, therefore, not 56 interchangeable. Indeed, one could imagine two communities comprised of entirely different species (high species beta diversity) that carry out the same functions (low functional beta 57 58 diversity; Swenson, 2011; Swenson et al., 2012; Siefert et al., 2013). Because functional diversity quantifies ecological impacts and species interactions, studies including the functional diversity 59 of organisms provide more robust data for conservation and restoration than studies on species 60 presence and abundance alone (Cadotte et al., 2009). 61

Conservation and restoration efforts are landscape-level problems requiring an 62 63 understanding of how both species and functional diversity scales from local to regional 64 perspectives. Using taxa, such as arthropods, that are widespread, abundant, speciose, and 65 functionally diverse can help researchers test diversity hypotheses across multiple scales (Kemp 66 et al., 2017; McCreadie & Adler, 2018). Ants and spiders play several important roles in 67 ecosystem functioning through direct and indirect species interactions (Pearce & Venier, 2006; 68 Maleque et al., 2009). Specifically, ants are numerically dominant in most ecosystems, affecting 69 ecological processes through their nest building, predation, and mutualisms (Folgarait, 1998; 70 Crist, 2009); therefore, shifts in ant communities can have multifaceted impacts across trophic 71 levels. Similarly, spiders are obligate predators of other arthropods and may be more susceptible 72 to changes in landscape fragmentation, and loss of key predators is linked to trophic cascades 73 (Bonte et al., 2004; Voigt et al., 2007).

The mechanisms that might structure arthropod taxonomic and functional beta diversity 74 75 across spatial scales (e.g., region, site, and forest stand) include characteristics such as land use 76 and topography, management history, and habitat characteristics, respectively. Forest 77 management practices (e.g., logging, fire regimes, and patch size) affect arthropods across 78 multiple scales. At landscape and regional scales, forest fragmentation due to logging or land-use 79 changes lead to decreased habitat area and isolation as well as greater edge to area ratio and 80 altered biotic and abiotic conditions along edge boundaries (Gluck & Rempel, 1996; Fahrig, 81 2003; Haddad et al., 2015). At the stand level, decreased forest area and increased edge cause shifts in tree stand pattern, vegetation structure, and soil characteristics (e.g. moisture, organic 82 matter, texture; Johnston & Elliott, 1996). The effects of these changes on arthropods are many 83 84 and varied. Fragment size impacts both taxonomic and functional diversity of ants and spiders as

well as shifting community composition of these taxa (Pearce et al., 2005). Forest habitats with 85 86 greater tree diversity and structural variability and complexity give rise to a greater variety of 87 niches, and, thus, species and roles to inhabit those niches (Tews et al., 2004). Thus, vegetation 88 or habitat complexity can drive patterns of arthropod species richness through direct effects on 89 foraging success and resource availability and indirect effects on ant nest site selection via soil 90 moisture and shading (Mcnett & Rypstra, 2000; Wang et al., 2001; Lassau & Hochuli, 2004). 91 Taken together, beta diversity and composition of key arthropod communities may be 92 affected at a range of spatial scales, potentially driven by several biotic and abiotic 93 characteristics. However, many studies of arthropod diversity focus on a single spatial scale and on taxonomic diversity alone. In this study, we evaluated the role of spatial scale on the 94 95 taxonomic and functional diversity and the community composition of two arthropod groups 96 (ants and spiders). We quantified these measures across four spatial scales: trees, forest stands, sites, and ecoregions. Specifically, we asked: 1) What spatial scales explain the greatest variation 97 in taxonomic and functional beta diversity and in the community composition of these two taxa? 98 99 and 2) What are potential mechanisms of scale-associated community structuring, including 100 variation in tree species, forest area and connectivity, and seasonality in climate? We 101 hypothesized that both taxonomic and functional diversity have greater variation at finer spatial 102 scales (tree and stand levels) than at broader spatial scales (site and ecoregion). We also expected 103 that spiders, with greater dispersal abilities, would exhibit greater levels of beta diversity at 104 broader spatial scales than ants, which are more dispersal limited. Similarly, we hypothesized that the biotic and abiotic factors structuring these patterns of diversity differ for spiders and ants 105 106 given inherent differences in resource use and dispersal ability.

107

#### 108 **2 METHODS**

## 109 2.1 Sampling Design, Study Sites, and Arthropod Collection

110 Samples were collected from tree canopies of southern Ohio and southeastern Indiana 111 using a hierarchically nested design with four levels. The broadest level – ecoregion – was 112 represented by the glaciated North-Central Till Plain and unglaciated Western Allegheny Plateau 113 (Figure S1). Each ecoregion varies in soil type, forest composition, and topography. The second 114 level – site – comprised 3 sites in each ecoregion for a total of 6 sites (Figure S1). The third level 115 - stand - comprised 4 forest stands within each site (24 total). Within each site, two stands were 116 located in uplands, and two stands were located in lowland topographic positions. The fourth 117 level - tree - included 8 individual trees (96 total) within a 1-ha area of each stand. Each tree was 118 sampled using canopy fogging and arthropods were collected in an array of 12 1-m<sup>2</sup> funnel arrays 119 placed under the crown of the fogged tree. Arthropods were knocked down by insecticidal fogging (0.5 L of 0.5% pyrethrin-based insecticide) for 3 minutes using a Curtis Dyna-Fogger 120 121 hoisted into the tree crown and collected by funnels with attached vials of ethanol. Insecticidal 122 fogging is not dependent on arthropod activity and lethality is non-specific (Basset et al., 1997; 123 Stork & Hammond, 1997). Sampling was completed during early (22 May-20 June 2000) and 124 late (2-25 August 2000) summer to capture seasonal variation, with early and late season samples 125 pooled for analyses. For more details on sampling methods, see Gering & Crist (2002). Ants were 126 identified to species using the Ants of Ohio (Coovert, 2005) and ant functional groups were trait-127 based (see below) rather than explicitly based on taxonomy or behavior (Andersen, 1997; Crist, 2009). Spider adults were identified to species based on the Spiders of North America (American 128 129 Arachnological Society, 2005), Spiders of Connecticut (Kaston, 1981), and a provisional list of Ohio spiders (Bradley, 2017), while juveniles were identified to family level when possible. 130

Spider families were used as an indicator of functional guild, based on foraging strategy, prey
range, habitat stratification, and circadian activity, as these traits are highly correlated with family
(Cardoso *et al.*, 2011).

We refer to these samples as canopy or arboreal arthropods because they were collected by fogging tree crowns. Although some of the common species we recorded are primarily arboreal (e.g., *Aphaenogaster mariae*, Table S3), most of the species of ants and spiders are known to move between strata and nest or overwinter in the ground. Nonetheless, Crist and Campbell (2017) recorded significant differences in the community composition of ants from canopy fogging and pitfall traps samples at the same study sites in the North-Central Till Plain.

140

141 2.2 Data Analyses

142 2.2.1 Ant functional groups

To classify ants according to functional traits, we selected 10 binary, categorical, and 143 144 continuous traits that may influence the ecological role of ants (Table S1; (Del Toro *et al.*, 2015; 145 Record et al., 2018; Mahon, 2019). We used trait definitions and data from Del Toro et al. 146 (2015), Record et al. (2018), Coovert (2005), and AntWiki (2022). Missing morphological data 147 were supplemented with measurements taken from 2-10 mounted specimens per species collected 148 during this study. Functional groups were formed from a dendrogram based on functional 149 dispersion and the Ward clustering method using the dbFD function (FD package, R; Laliberté & 150 Legendre, 2010); functional groups were delineated by setting six functional groups, where we noted a clear break of functional groups (Table S1, Figure S2). 151

153 2.2.2. General patterns and species accumulation curves

154	For all statistical analyses, early and late sampling periods were pooled and all univariate
155	analyses were conducted in R v4.0.0 (R Core Team, 2022). We constructed sample-based
156	rarefaction curves of ants and spiders by tree species to assess differences in species richness
157	among the most common host trees. To determine the effectiveness of our overall sampling
158	effort, we also conducted species rarefaction and extrapolation curves for ants and spiders using
159	the iNEXT package in R (Chao et al., 2014). We also estimated species richness by host trees and
160	overall richness using the Chao1 estimator (Chao, 1984; Colwell & Coddington, 1994).
161	
162	2.2.3 Diversity partitioning
163	To analyze taxonomic diversity, we partitioned species richness and functional group
164	richness $(q = 0)$ across hierarchical levels, with multiplicative partitioning methods based on the
165	PARTITION software developed by Crist et al. (2003) using the R package, PARTITIONR
166	(Mahon et al., 2019). Multiplicative partitions express beta diversity as the effective number of
167	distinct communities, whereas alpha diversity is in units of species richness (Anderson et al.,
168	2011). Using hierarchical diversity partitioning, beta components can be separated into nested
169	hierarchical levels (i.e. spatial scale). Hierarchical multiplicative partitioning calculates beta
170	diversity at a given level (i) by dividing the average alpha at the i+1 level, $\beta_i = \alpha_{i+1}/\alpha_i$ . We
171	tested the significance of $\alpha_1$ (within trees), $\beta_1$ (among trees), $\beta_2$ (among stands), $\beta_3$ (among
172	sites), and $\beta_4$ (between ecoregions) against null, random distributions using 1000 sample-based
173	randomizations. This type of randomization preserves intraspecific aggregation at each
174	hierarchical level, and thus tests the null hypothesis that observed patterns of species diversity at

each level are similar to those expected by randomized aggregation of samples at each level of 175 176 the hierarchical design; alternatively, the null hypothesis is rejected if observed patterns are 177 significantly different from those expected from null distributions, supporting non-random 178 hierarchical species assemblages that are structured by ecological processes (Crist *et al.*, 2003). 179 Partitioning of alpha, beta diversity in this manner therefore accounts for issues of spatial 180 pseudoreplication that can arise in local-to-regional comparisons of species richness (Gering & 181 Crist, 2002). We used a two-tail probability of p = 0.05 (p = 0.025 for each tail of the null 182 distribution) to determine whether observed diversity patterns were higher or lower than expected via randomizations. To compare relative deviations of the null distributions from the expected 183 values across hierarchical levels and endpoints, we calculated standard effect sizes (SES; also 184 185 termed beta and alpha deviations) from the mean and standard deviation of the null distributions for each hierarchical level (SES =  $(I_{obs} - I_{exp})/\sigma$ ) (Gotelli & McCabe, 2002; Kraft *et al.*, 2011). 186 To analyze variation in community composition across hierarchical levels, we conducted 187 analyses for multivariate location using PRIMER-E and PERMANOVA+ v6 (Anderson, 2001). 188 We used permutational multivariate analysis of variance (PERMANOVA) to partition the 189 190 variation in community composition across hierarchical levels. PERMANOVAs were conducted 191 using 9999 permutations of the data, with a nested design. Since multivariate analyses were conducted on square-root transformed abundance data with a Bray-Curtis dissimilarity, variance 192 components of PERMANOVA can be interpreted as percent dissimilarity (Anderson, 2001). 193 194

195 2.2.4 Landscape and environmental variable analyses

To address potential mechanisms of diversity structuring, we compiled and evaluated 196 197 landscape, climactic, and vegetation variables for forest stands and surrounding areas. Landscape 198 variables were habitat fragmentation measures (estimated using ESRI arcGIS Dekstop v10.6.1 199 (ESRI, 2019), Fragstats v4.2.1 (McGarigal et al., 2012)) and land use data (USDA NASS Crop 200 data layer for 2008, the closest year with most reliable data for both Indiana and Ohio (USDA 201 National Agricultural Statistics Service, 2022)). Land cover/land use composition and 202 configuration was collected for of 1.0, 3.0, 6.5, and 10.0 km buffers surrounding each stand to 203 accommodate the potential dispersal distances of both spiders and ants, and to minimize overlap 204 between stand buffers (Thomas et al., 2003). Initial model selection (see methods below) was 205 conducted to determine which buffer size explained the most variation for both ants and spiders; 206 data at the 6.5 km buffer were used for the remainder of the statistical analyses. Each landscape 207 was analyzed using a "no-sampling" strategy and an 8-cell neighborhood rule for the following class level metrics as specified in Fragstats: CLUMPY, PLAND, GYRATE\_MN, 208 GYRATE\_AM, and PARA (McGarigal et al., 2012). CLUMPY (fragmentation index) is an 209 210 index of fragmentation of deciduous forest within the measured landscape where -1 is highly 211 fragmented forest and 1 is a complete, unfragmented forest; the CLUMPY index is not 212 confounded by changes in forest area. PLAND is the percentage of deciduous forest within the 213 landscape. GYRATE MN is the mean distance (m) to forest edge from patch centroid. 214 GYRATE\_AM (patch connectedness), is the area-weighted distance (m) to forest edge from 215 patch centroid, or patch connectedness. PARA (patch edge:area ratio) is the mean edge:area ratio 216 for all deciduous forest patches. We obtained 19 bioclimatic variables from WorldClim v2.0 217 (Fick & Hijmans, 2017); due to the relatively course resolution ( $\sim 1 \text{ km}^2$ ) of the bioclimatic 218 variables, 2 of our stands (within Brookville) had identical climatic data, but all other stands and

sites varied. We included a stand-level vegetation measure of tree species richness (dbh  $\ge 10$  cm) by recording all tree species present within the same 1-ha stands where trees were fogged (Crist *unpublished data*).

222 We accounted for collinearity among landscape and environmental variables by removing 223 those that were highly correlated (Pearson  $r \ge 0.80$ ). In total, 10 variables were included: 4 224 landscape variables, 5 climatic variables, and 1 vegetation variable (Table S2). Prior to all 225 analyses, variables were standardized to z-scores to aid in model fitting and inference. To 226 determine climatic, landscape, and vegetative influence on patterns of taxonomic and functional 227 richness, we used linear regression models (Im function, stats package, R; (R Core Team, 2022) 228 with response variables of mean alpha ( $\alpha_1$ , within trees) and beta ( $\beta_1$ , among trees) diversity for 229 taxonomic and functional diversity at each stand. Model selection allowed for the additive term 230 of all environmental predictors. To identify best models, we used the lowest Akaike's Information Criterion with bias-correction (AICc). For best models, we calculated AICc weights 231 (w) and appropriate  $\mathbb{R}^2$ . We tested for spatial autocorrelation using Moran's I (Moran.I function, 232 ape package, (Paradis & Schliep, 2019), which indicated no spatial autocorrelation present in the 233 residuals of our univariate models. 234 235 Similarly, we used DISTLM and distance-based redundancy analyses (dbRDA) in

PRIMER and PERMANOVA+ (McArdle & Anderson, 2001) to assess influences of our environmental variables on community composition at the stand level. We performed stepwise model selection based on AIC values to examine the relationship between the explanatory variables and community composition. We only included variables with significance of p < 0.10in preliminary marginal tests to reduce likelihood of overfitting multivariate analyses.

Significance values and variance explained by selected predictors were found using bootstrap
tests based on 9999 iterations. Multivariate analyses were conducted on square-root transformed
abundance data with Bray-Curtis dissimilarity.

244

#### 245 **3 RESULTS**

246 We collected 3,053 individual ants representing 23 species with 2 singleton species (Table 247 S3). Estimated Chao1 richness was  $24 \pm 2$  species, with rarefaction curves indicating ant richness 248 had plateaued (Figure S3). We collected 5,221 individual spiders representing 23 families, 67 249 genera, and 97 species. Of these, 83.7% were juveniles, we identified the remaining 925 adults to morphospecies (Table S3). Of these species, 15 were represented by a single specimen. Dominant 250 251 spider families were Araneidae (23%), Linyphiidae (17%), Salticidae (16%), Anyphaenidae 252 (12%), and Theridiidae (10%). Estimated Chao1 richness was  $108 \pm 6$  species, with rarefaction 253 indicating spider richness began to plateau and likely would have been reached by 150 tree 254 samples (Figure S3).

## 255 *3.1 Hierarchical diversity partitioning*

Spider species alpha and beta diversity were higher than ant species diversity, ant functional diversity, and spider functional diversity (Table 1). Multiplicative beta diversity based on species richness (q = 0) exhibited similar patterns between taxa, with beta decreasing as scale increased (Figure 1). For species richness, the observed  $\beta_{tree}$  components were 2x the  $\alpha_{tree}$  of 4.8 species of ants per tree and 2.8x of 6.4 species of spiders per tree, respectively. The  $\beta_{stand}$  values were 1.6x the combined of 9.5 species of ants per stand, and 2.4x the 17.9 species of spiders per stand. The  $\beta_{site}$  components were 1.3x the 15.7 species of ants per site and 1.7x the 42.9 spider

263 species per site, and  $\beta_{ecoregion}$  were <1.3x the mean richness of ants and spiders per ecoregion 264 (Table 1). The functional group diversity components mirrored those of species richness, except 265 that the  $\alpha_{tree}$  and  $\beta_{tree}$  levels comprised most of variation in total functional diversity of ants with additional variation explained by  $\beta_{stand}$  for spiders. Randomization tests of species richness 266 267 indicated  $\alpha_{\text{tree}}$  was significantly lower than expected across taxa (p < 0.001),  $\beta_{\text{tree}}$  was higher than 268 expected across taxa (p < 0.001),  $\beta_{\text{stand}}$  was higher than expected for ant species and spiders (p < 0.001) 0.001),  $\beta_{\text{site}}$  was higher for spider species (p = 0.005), and  $\beta_{\text{region}}$  was not different across taxa (p > 0.005). 269 270 0.025, Table 1). These patterns emerged, despite having similar diversity across scales for diversity for ant species, ant functional groups, and spider functional groups (Figure 1, Table 1). 271 At local scales (within and among trees), for both taxonomic and functional endpoints, the 272 deviations from observed diversity were greater for ants than spiders, but the opposite pattern 273 274 emerged at broader scales (stand, site, and region; Figure 1), indicating differential hierarchical 275 patterns of species diversity between ants and spiders.

276

#### 277 3.2 Community composition

We tested for multivariate differences in ant and spider species composition across hierarchical levels based on variation in Bray-Curtis dissimilarities (Table 2). For both taxa, a greater proportion of the variation in dissimilarities was explained at site and ecoregion levels than with univariate diversity partitions, but components still decreased with increasing spatial scale (Table 2). Spiders also showed greater residual variability among trees (56%) compared to ants (40%). Similarly, spider species composition showed greater variance among stands within sites than ant species composition (Table 2). Ant assemblages showed significant variability

across all spatial scales; however, this was not the case for spiders, as the variance component for
ecoregion was not different from zero (Table 2).

For both ants and spiders, functional composition was less variable than taxonomic 287 composition across scales (Table 2). Ant and spider functional assemblages were significantly 288 289 variable at the stand and site levels, but the variance component for ecoregion was not different 290 from zero (Table 2). While spider functional composition showed reduced variance components 291 with increasing spatial scale, ant functional composition showed little differentiation among stand, site, and ecoregion scales, but variance was highest among sites (Table 2), indicating 292 293 stronger variability across sites within ecoregion than among stands within sites. Nonetheless, 294 both ant and spider functional composition was most variable among trees within stands (Table 295 2), suggesting functional differentiation of these communities among tree species.

296

## 297 *3.3 Patterns of diversity and community composition*

Tree identity influenced spider and ant species richness. For ants, white oak (*Quercus alba*), red oak (*Q. rubra*), American beech (*Fagus grandifolia*), sugar maple (*Acer saccharum*), and hickory (*Carya* spp.) hosted similar levels of richness, while tulip poplar (*Liriodendron tulipifera*) maintained the lowest richness (Figure 2B). Ant richness began to plateau on most tree species (but see hickory, Figure 2A). For spiders, white oak, sugar maple, and American beech had greater richness than tulip poplar, red oak, and hickory, as indicated by rarefaction (Figure 2B). Unlike for ants, most tree species did not begin to plateau for spiders, except for red oak.

305 Average alpha ( $\alpha_1$ , within trees) and beta ( $\beta_1$ , among trees within stand) diversity for 306 taxonomic and functional endpoints were driven by a combination of landscape and climatic

variables measured at the stand level. Temperature evenness (isothermality) and precipitation 307 308 seasonality were important predictors of alpha diversity of spider functional groups, ant species, 309 and ant functional groups as well as beta diversity of spider species, spider functional groups, and 310 ant species (Table 3). Specifically, precipitation seasonality was negatively associated with alpha 311 diversity, but positively associated with beta diversity (Table 3). This suggests that increased 312 variability in precipitation may simultaneously reduce average alpha within trees while increasing 313 differentiation of arthropod communities among tree species (beta). Stand-level richness of tree 314 species was positively associated with beta diversity of ant species, but negatively associated 315 with beta diversity of spider functional groups (Table 3). Finally, patch connectedness, size, and 316 shape were important predictors of alpha diversity of ant and spider functional groups and of beta 317 diversity of ant functional groups (Table 3).

Similar to univariate patterns of alpha and beta diversity, variation in multivariate 318 319 community composition using dbRDA was driven by a similar combination of landscape and 320 climatic drivers. Variation in ant community composition was best explained by precipitation variability ( $R^2 = 16.3\%$ ) (Figure 3A). Variation in ant functional group composition was best 321 explained by isothermality ( $\mathbf{R}^2 = 40\%$ ), patch connectedness (14.2%), and precipitation 322 seasonality (11.1%) (Figure 3B). Variation in spider community composition was best explained 323 by isothermality ( $R^2 = 17.6\%$ ), maximum temperature (15.6%), and habitat fragmentation 324 325 (10.4%) (Figure 3C). Variation in spider functional guild composition was best explained by patch connectedness ( $R^2 = 25.2\%$ ), isothermality (16%), stand tree richness (9.6%), and habitat 326 327 fragmentation (8.9%) (Figure 3D).

#### 329 4 DISCUSSION

#### 330 4.1 Spatial structure of diversity

331 Our results indicate differential patterns of beta diversity and community structure 332 between canopy-dwelling ants and spiders and that taxonomic diversity is more variable at larger 333 spatial scales than functional diversity. In support of our hypothesis that spider diversity would be 334 more affected by broader spatial scales than ants, we saw greater diversity and greater deviation 335 from expected diversity (SES) at broader spatial scales for spiders than for ants, but there was 336 more variation between ecoregions for ant communities than for spider communities. While 337 functional and taxonomic diversity exhibited similar patterns across scales, taxonomic diversity exceeded functional diversity at coarser spatial scales (Site and Ecoregion), suggesting trait 338 339 clustering and functional redundancy at broader spatial scales, consistent with the findings of 340 Jarzyna & Jetz (2018). Essentially, the loss or addition of species at the site and ecoregion scales 341 has little influence on the loss or addition of ecological functions – as determined by functional groups – at these same scales. 342

The finding that broad-scale (Ecoregion) beta components of diversity were not different 343 344 than expected suggests the effects of ecoregions do not structure the taxonomic and functional 345 diversity of canopy-dwelling ants and spiders. Our results contrast with the significant 346 differentiation of beetle diversity between ecoregions from these same samples (Gering *et al.*, 347 2003) but are consistent with those of Summerville et al. (2003) who found no significant deviation from expected diversity in forest moth species richness among ecoregions. Nonetheless, 348 both ant and spider communities showed similar patterns of decreasing community variation with 349 350 increasing spatial scale, indicating strong differences in beta diversity at the tree and stand levels, 351 consistent with the findings of Gering et al. (2003) and Summerville et al. (2003). Here, we also

352 conducted multivariate partitions of Bray-Curtis dissimilarity across scales, which showed
353 slightly larger components of variation at the site and ecoregion levels than univariate partitions.
354 This suggests that shifts in the relative abundance of species or functional group across sites and
355 ecoregions were more important than shifts in species or functional group composition.

356 Lower variation than expected at the individual tree-level implies that tree species support 357 distinct levels of species richness and, likely, functional diversity of arthropod taxa, as supported by rarefactions of individual tree species for both ants and spiders. Differences in tree species-358 359 arthropod species richness relationships (estimated and observed richness) indicate potential tree-360 species specific constraints such as prey species abundance and competition/territoriality in ants (Majer & Delabie, 1999; Yasuda & Koike, 2009) and nesting/web site limitations for spiders 361 362 (Nicolai, 1986; Larrivée & Buddle, 2010). Thus, the maintenance of temperate, canopy-dwelling 363 arthropod communities is, at least somewhat, dependent on maintaining the diversity of host trees. As such, the subsequent emerald ash borer (Agrilus planipennis) driven loss of ash trees 364 (Fraxinus spp.; Herms & McCullough, 2014) and long-term declines of several oak species in 365 366 eastern forests (McEwan et al., 2011) since the time of sampling has likely resulted in the loss of 367 distinct arboreal arthropod communities and, possibly, species since the early 2000s.

368

369 4.2 Landscape and environmental influences

Previous studies have found differential environmental drivers between taxonomic and
functional diversity (Longhi & Beisner, 2010; Pool *et al.*, 2010; Meynard *et al.*, 2011),
suggesting that different environmental filters act on taxonomic and functional diversity of the
same taxa. While we saw similar scaling patterns between ants and spiders, taxonomic and

functional diversity, and community structure, the climatic, landscape, and vegetation drivers of 374 375 these patterns did, indeed, differ (Table 3, Figure 3). Climatic characteristics showed strong 376 influence on alpha and beta diversity as well as both taxonomic and functional composition, 377 while landscape characteristics were important in explaining patterns of functional alpha diversity 378 and functional composition. Nonetheless, taxonomic and functional community composition was 379 explained by shared climatic variables (precipitation seasonality and isothermality), indicating 380 that both taxonomic and functional assemblages are partially driven by the same climatic 381 conditions.

Climate variables were more common than landscape variables in univariate (diversity) 382 and multivariate (community composition) ant and spider models. Precipitation seasonality was 383 384 the most common predictor in explaining variation in taxonomic and functional diversity (Table 385 3), while isothermality (temperature evenness) was the most common predictor for explaining multivariate variation in taxonomic and functional community composition (Figure 3). These 386 differences in climate variables as predictors of univariate and multivariate variation in diversity 387 may also reflect variation in species distributions versus shifts in relative abundances. The 388 389 broader implication of climate variability as an important driver of diversity and composition of 390 canopy arthropod communities is that climate change mediated increases in climate variability 391 (increased precipitation seasonality and decreased isothermality) will likely greatly alter these 392 canopy-dwelling communities (Westerling, 2016; Jump et al., 2017; Neumann et al., 2017).

Habitat configuration was more important in explaining patterns of ant and spider alpha and beta diversity and community structure than total habitat availability, but spider functional richness was related to patch size and shape. In line with our results, previous research suggests

that ants show weak support for species-area relationships, while spiders tend to exhibit strong 396 397 positive species-area relationships (Crist, 2009; Cardoso et al., 2010). Similarly, previous studies 398 have found higher ant richness in habitat fragments with higher connectivity, while habitat 399 connectivity does not influence spider species richness (Abensperg-Traun et al., 1996; Suarez et 400 al., 1998; Cardoso et al., 2010). The disparities in the influence of habitat fragmentation and 401 patch size and shape between ants and spiders is likely due to differences in dispersal ability 402 (Thompson & Townsend, 2006). Our results suggest that patterns of arthropod diversity among 403 habitat patches is influenced by dispersal ability, with connectivity being an important predictor 404 of poor dispersing arthropods and patch size and shape being an important predictor of diversity of stronger dispersing arthropods. Taken together, this suggests that the importance of habitat 405 406 configuration and area of habitat in determining species richness (Haddad et al., 2017) may be 407 somewhat dependent upon the dispersal ability of organisms. Yet, connectivity and fragmentation of habitat patches were important for composition of both taxa, suggesting that while spider 408 richness may not be influenced by habitat configuration, the identities of spider species is 409 410 influenced by habitat configuration. Therefore, conservation efforts should focus on increasing 411 both patch connectivity and size to provide the largest benefits to diversity of all arthropods. 412 Nevertheless, future research should expand upon the relative roles of habitat configuration and patch size and shape in driving patterns of diversity for taxa across a dispersal gradient. 413

414 *4.3 Conclusions* 

Our findings demonstrate stronger scaling patterns of taxonomic diversity than functional
diversity from local to regional scales, suggesting functional redundancy at broader spatial scales.
Further, taxonomic and functional diversity and community assemblages change along different

- 418 environmental gradients. The controls of climate and landscape fragmentation on the diversity
- 419 and structure of canopy-dwelling ants and spider communities indicate that climate change via
- 420 increased variability will likely further alter the diversity, composition, and function of arboreal
- 421 arthropods that are already threatened by forest fragmentation and land use changes. Our findings
- 422 provide further support for the consideration of functional components of diversity and multiple
- 423 measures of beta diversity in monitoring and conservation (Cadotte *et al.*, 2011; Socolar *et al.*,
- 424 2016; Isbell *et al.*, 2017; Jarzyna & Jetz, 2018).

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Code and Data accessibility: All data and R code are available on github:

https://github.com/mahonmb/CanopyArthropodDiversity. PRIMER and PERMANOVA+ code is available by request.

# Tables.

Table 1. Multiplicative partition of alpha and beta diversity components across a hierarchically nested study of canopy ant and spider diversity. *P* values <0.05 are bolded

Source	Observed Diversity	Expected Diversity	р	SES	
Ant Species		<b>v</b>			
Ecoregion ( $\beta_{region}$ )	1.10	1.10	0.687	-0.01	
Site ( $\beta_{site}$ )	1.34	1.33	0.386	0.52	
Stands ( $\beta_{stand}$ )	1.66	1.55	0.001	3.94	
Among Trees ( $\beta_{tree}$ )	1.98	1.57	0.001	24.67	
Within Trees ( $\alpha_{tree}$ )	4.78	6.02	0.001	-19.6	
Ant Functional Group					
Ecoregion ( $\beta_{region}$ )	1.00	1.00	0.500	0.00	
Site ( $\beta_{site}$ )	1.06	1.04	0.348	1.27	
Stands ( $\beta_{stand}$ )	1.16	1.16	0.520	0.15	
Among Trees ( $\beta_{tree}$ )	1.56	1.33	0.001	15.23	
Within Trees ( $\alpha_{tree}$ )	3.13	3.66	0.001	-13.01	
Spider Species					
Ecoregion ( $\beta_{region}$ )	1.30	1.28	0.291	0.92	
Site ( $\beta_{site}$ )	1.74	1.67	0.003	2.98	
Stands ( $\beta_{stand}$ )	2.40	2.23	0.001	5.97	
Among Trees ( $\beta_{tree}$ )	2.79	2.60	0.001	7.01	
Within Trees ( $\alpha_{tree}$ )	6.42	6.88	0.001	-6.56	
Spider Functional Group					
Ecoregion ( $\beta_{region}$ )	1.15	1.15	0.803	0.19	
Site ( $\beta_{site}$ )	1.24	1.23	0.475	0.36	
Stands ( $\beta_{stand}$ )	1.32	1.26	0.001	4.27	
Among Trees ( $\beta_{tree}$ )	1.38	1.32	0.001	7.12	
Within Trees ( $\alpha_{tree}$ )	8.89	9.32	0.001	-6.80	

Table 2. Permutational multivariate analyses of variance (PERMANOVA) based on Bray-Curtis dissimilarity of square-root transformed abundance data for species and functional groups for both taxa. *P*-values are based on Monte-Carlo randomization, *p* values <0.05 are bolded. Variance is the square root of the estimated component of variance, to put values on the scale of Bray-Curtis dissimilarities (percent difference among assemblages).

Endpoint and Source of Variation	df	MS	Pseudo- F	р	Variance	Percent Variance Explained
Ant Species						<b>▲</b>
Ecoregion	1	14407	2.41	0.042	13.3	15.9%
Site(Ecoregion)	4	5970.3	2.31	<0.001	14.5	17.4%
Stand(Site(Ecoregion))	18	2589.7	1.65	<0.001	16.0	19.2%
Residual	72	1566.1			39.6	47.5%
Ant Functional Groups						
Ecoregion	1	10354	2.71	0.066	11.7	17.8%
Site(Ecoregion)	4	3821.6	2.61	0.003	12.1	18.5%
Stand(Site(Ecoregion))	18	1463.4	1.63	0.003	11.9	18.1%
Residual	72	899.2			30.0	45.7%
Spider Species						
Ecoregion	1	15822	1.82	0.058	12.2	11.7%
Site(Ecoregion)	4	8706	1.82	<0.001	15.7	15.1%
Stand(Site(Ecoregion))	18	4778.7	1.53	<0.001	20.4	19.6%
Residual	72	3120			55.9	53.7%
Spider Functional Groups	5					
Ecoregion	1	4849.3	2.01	0.089	7.1	13.5%
Site(Ecoregion)	4	2415.2	1.80	0.023	8.2	15.5%
Stand(Site(Ecoregion))	18	1343.8	2.42	<0.001	14.0	26.5%
Residual	72	554.9			23.6	44.5%

Table 3. Multiple regressions of tree-level alpha and beta diversity as predicted by stand-level measurements of climate, landscape structure and tree species richness. Regression coefficients of best models as indicated by AICc model selection. All predictor variables were converted to Z scores (SE) to allow for comparison of strength of effects.

	Predictor Variables							
	df	Patch Connectedness	Patch Edge:Area	Tree Richness	Isothermality	Temp Warm Month	Precipitation Seasonality	R <sup>2</sup>
Alpha							-	
Ant Species	2						-0.73 (0.22)	0.53
Ant Func. Group	3	0.44 (0.18)					-0.60 (0.18)	0.72
Spider Species	1							
Spider Func. Guild	3		-0.99 (0.25)				-1.27 (0.24)	0.70
Beta								
Ant Species	4	0.82 (0.21)		0.49 (0.20)			0.86 (0.21)	0.63
Ant Func. Group	2					-0.58 (0.26)		0.33
Spider Species	2				-0.54 (0.27)			0.29
Spider Func. Guild	3			-0.52 (0.23)			0.49 (0.23)	0.44

Figure 1. Standard effect sizes (SES) of the observed diversity partitions and the diversity deviations from a null model across the hierarchical scales (within trees, among trees, stand, site, and region) for taxonomic and functional diversity. Blue triangles are spiders, pink circles are ants.

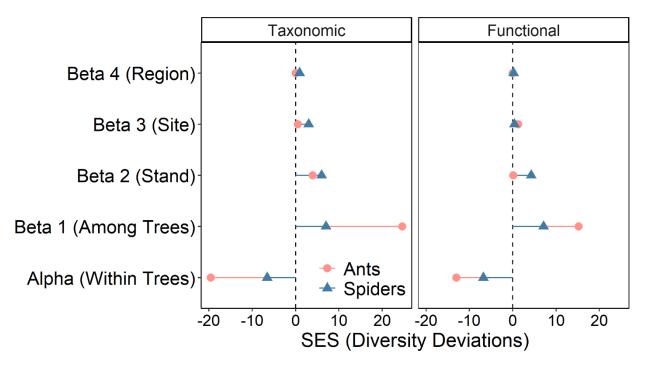


Figure 2. Sample-based rarefaction curves of ant (A) and spider (B) species by number of host trees sampled. Ash, Hackberry, Sycamore, and Walnut trees are not shown, because of few individual trees sampled (< 5 for each species). Number of points on a curve and length of the curve represent the number of individual trees fogged. Ants and spiders show differential patterns among tree species of observed (O) and Chao1 estimated (E) species richness.

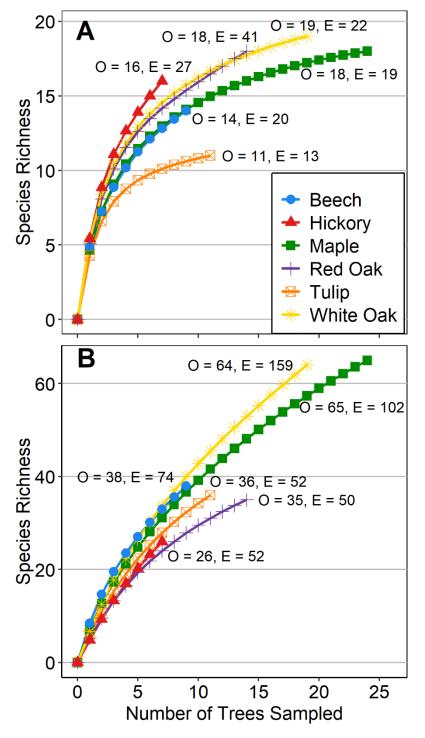
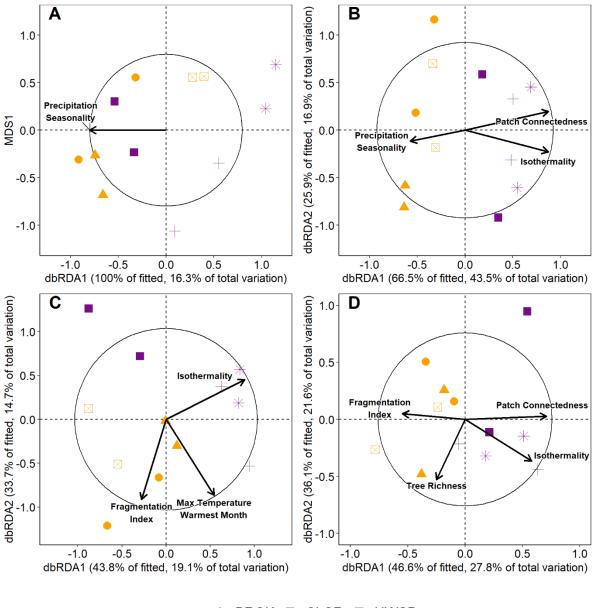


Figure 3. Distance-based redundancy analysis (dbRDA) of stand-level variation in composition based on Bray-Curtis dissimilarities of square-root transformed abundance of ant species (A), ant functional groups (B), spider species (C), and spider functional groups (D). Vectors correspond to predictor variables in best model as determined by model selection. The circles correspond to vector lengths that would have a correlation coefficient of 1 with a given axis, strength of correlation for vectors are scaled to this circle. Yellow symbols are stands within sites in the North-Central Till Plains and purple symbols are stands within sites in the Western Allegheny Plateau. Sites are designated by different symbol shapes.



Site ● BROK ■ CLCR ⊠ HWSP ▲ CACR + EOA \* SSFO