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1 **Effects of soils, topography and geographic distance in**
2 **structuring central Amazonian tree communities**
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1 **Abstract**

2 **Question:** What is the relative contribution of geographic distance, soil and topographic
3 variables in determining the community floristic patterns and individual tree species
4 abundances in the nutrient-poor soils of central Amazonia?

5 **Location:** Central Amazonia near Manaus, Brazil

6 **Methods:** Our analysis was based on data for 1105 tree species (≥ 10 cm dbh) within 40
7 1-ha plots over a ~ 1000 -km² area. Slope and 26 soil-surface parameters were measured
8 for each plot. A main soil-fertility gradient (encompassing soil texture, cation content,
9 nitrogen and carbon) and five other uncorrelated soil and topographic variables were used
10 as potential predictors of plant-community composition. Mantel tests and multiple
11 regressions on distance matrices were used to detect relationships at the community level,
12 and ordinary least square (OLS) and conditional autoregressive (CAR) models were used
13 to detect relationships for individual species abundances.

14 **Results:** Floristic similarity declined rapidly with distance over small spatial scales (0-5
15 km), but remained constant ($\sim 44\%$) over distances of 5 to 30 km, which indicates lower
16 beta diversity than in western Amazonian forests. Distance explained 1/3 to 1/2 more
17 variance in floristics measures than environmental variables. Community composition
18 was most strongly related to the main soil-fertility gradient and C:N ratio. The main
19 fertility gradient and pH had the greatest impact of species abundances. About 30% of
20 individual tree species were significantly related to one or more soil/topographic
21 parameters.

1 **Conclusions:** Geographic distance and the main fertility gradient are the best predictors
2 of community floristic composition, but other soil variables, particularly C:N ratio, pH,
3 and slope, have strong relationships with a significant portion of the tree community.

4 **Keywords:** Beta diversity; Brazil; floristics; tropical forest; species abundances; spatial
5 dependence

1 **Introduction**

2 Much research has focused on the relative role of distance-related processes such
3 as dispersal versus environmental factors in shaping community floristics in tropical
4 forests (Clark et al. 1999; Chust et al. 2006; Hubbell 2001; Phillips et al. 2003; Potts et al.
5 2002; Webb & Peart 2000 and others). Certainly no one answer has emerged as to which
6 is more important. Instead, the importance of different factors varies from region to
7 region (Palmer 2006; Tuomisto et al. 2003a, b, c), and even at different spatial scales in
8 the same region (Normand et al. 2006; Potts et al. 2002), in response to past and recent
9 geologic history, human and natural disturbance patterns, and evolutionary processes.

10 Nonetheless, some environmental variables, particularly soil texture, cation
11 composition and topography, are repeatedly reported as important in determining tropical
12 forest composition (Costa et al. 2005; Sri-Ngernyuang et al. 2003; Tuomisto et al. 2002;
13 Valencia et al. 2004; Vormisto et al. 2004). The availability of complete floristics and
14 soils data sets that are needed to address these questions is very limited given the vast
15 extent of the tropics. In the large area covering the central Amazon basin, only one study
16 with detailed soil and floristic data examines the relationship between environmental
17 variables and floristics (Costa et al. 2005) and this is limited to understory herbs. To put
18 the floristic community in context of other studies in the Neotropics (Condit et al. 2000;
19 Pitman et al. 2001; ter Steege et al. 2006), an analysis of trees (> 10 cm dbh) is needed.

20 In the relatively nutrient-rich soils of western Amazonia, evidence is
21 accumulating that tree communities exhibit surprisingly little turnover in floristic
22 composition (beta diversity) at larger spatial scales (10–1000 km), despite having high
23 local species richness (alpha diversity) and relatively rapid turnover at local scales (<10

1 km) (Condit et al. 2002; Macia and Svenning 2005; Pitman et al. 2001). Species
2 diversity patterns are less clear for the vast region of central Amazonia, where soil
3 fertility tends to be very low (Brown 1987; Chauvel et al. 1987). Central Amazonia has
4 one of the highest alpha diversities of trees recorded for any tropical forest (Oliveira &
5 Mori 1999), but species turnover at landscape scales (distances of <100 km) has not yet
6 been assessed.

7 Here, we describe the patterns of beta diversity in a central Amazonian forest and
8 address two questions: What is the relative role of geographic distance and
9 environmental variables in shaping floristic similarity and individual species abundances?
10 Which soil and topographic variables are most important? Because spatial structuring in
11 the environmental variables can cause inflated levels of environmental associations (e.g.
12 Harms et al. 2001, Lichstein et al. 2002), our statistical approaches aimed to disentangle
13 the effects of broad-scale geographic trends, local spatial autocorrelation and
14 environmental variables. We also placed our findings in the context of other studies in
15 the tropics. In particular, we examined how beta diversity patterns compare to other plot
16 networks in the Neotropics, and how the proportion of species with significant soil or
17 topographic relationships varied among different tropical forest communities.

18

19 **Methods**

20 *Study area and floristics*

21 Our study area is part of the Biological Dynamics of Forest Fragments Project
22 (BDFFP), a large-scale experimental study of habitat fragmentation, located in the central
23 Amazon, 80 km N of Manaus, Brazil (2° 30' S, 60° W) at 50-100 m elevation (Laurance

1 et al. 2002; Lovejoy et al. 1986). The study area spans ~1000 km². The topography
2 consists mostly of flat, clay-rich plateaus dissected by numerous stream and river gullies.
3 Forests in the area are not seasonally inundated. Rainfall ranges from 1900-3500 mm
4 annually with a pronounced dry season from June to October.

5 This site shows relatively few signs of human disturbance or large-scale natural
6 disturbance in the past few hundred years. There is no evidence of swidden agriculture
7 (Piperno & Becker 1996) and the majority of charcoal found at the site indicating fire
8 dates from at least 1100–1500 yr ago (Piperno & Becker 1996; Santos et al. 2000). Large
9 gaps created by wind bursts have been observed in this area of the central Amazon
10 (Nelson et al. 1994) but there is no evidence of recent large blowdowns in the BDFFP
11 plot network (Laurance et al. 2005; but see Nelson 2005).

12 Since 1980, a long-term study of tree community dynamics and composition has
13 been conducted in fragmented and continuous forests in the study area. All trees ≥ 10 cm
14 diameter-at-breast-height [dbh] are being monitored within 66 permanent, square, 1-ha
15 plots spanning the experimental landscape. Live trees in each plot were marked and
16 mapped and their diameters recorded, with measurements taken above the buttresses for
17 buttressed trees. On average, 95.3% of all trees in each plot were identified to species (or
18 morphospecies) level, using sterile or fertile material from each tree. Voucher specimens
19 are housed in the BDFFP Plant Collection, Manaus, Brazil. The subset of 40 plots was
20 used for this analysis contained 1105 identified tree species with a range of 224 to 293
21 species and 546 to 754 stems per plot.

22 Because we were interested in the relationship between floristics and
23 environmental patterns in undisturbed forest, we used data only from the initial tree

1 census, in the early-mid 1980s, completed before the BDFFP fragments were created and
2 when the whole area was continuous forest. The distance between the 40 plots ranged
3 from 0.10 to 32.5 km, with an average interplot distance of 13.8 km.

4
5 *Soil data*

6 The dominant soils in the study area are xanthic ferralsols (using the
7 FAO/UNESCO system; Beinroth 1975). Ferralsols are widespread in the Amazon Basin,
8 heavily weathered, and usually have a low base saturation. They often are well
9 aggregated, porous, and friable, with variable clay contents. Clay particles in ferralsols
10 can form very durable aggregations, giving the soil poor water-holding characteristics,
11 even with high clay contents (Richter & Babbar 1991). Xanthic ferralsols in the Manaus
12 area are derived from Tertiary deposits and are typically acidic and very poor in nutrients
13 such as P, Ca, and K (Chauvel et al. 1987).

14 A total of 21 soil parameters and 1 topographic parameter (slope) was collected
15 from the plots. Each 1-ha plot was divided into 25 quadrats of 20 X 20 m each from
16 which soil-surface samples (0-20 cm depth) were collected. The maximum slope within
17 each quadrat was measured with a clinometer, then averaged to yield a mean value for the
18 plot. To assess soil parameters, 9-13 quadrats were selected per plot, using an alternating
19 pattern to provide good coverage of the plot. Within each quadrat, 15 surface samples
20 were collected at haphazard locations using a soil auger, then bulked and subsampled.
21 The organic matter layer was removed before sampling. The laboratory methods used for
22 soil analyses are detailed in Fearnside and Leal-Filho (2001), and briefly summarized in
23 Appendix 1.

1 Because many soil variables were intercorrelated, we used Principal Components
2 Analysis (PCA) to identify major trends in the soil data using the 40 plots with complete
3 soils data. The first PCA axis explained 50% of the total variance in the soil data and
4 was strongly correlated with soil texture (fractions of clay, sand, and silt), total N, C
5 content, total exchangeable bases, and several individual cation concentrations (Zn, Mg,
6 Mn) (Appendix 2; Laurance et al. 1999). Because these parameters are strongly
7 associated with soil fertility (Laurance et al. 1999), we hereafter refer to PCA axis 1, and
8 the soil variables strongly correlated with it (particularly clay content), as the main soil-
9 fertility gradient.

10

11 *Statistical analyses*

12 In both the community and species level analyses, we sought to assess the relative
13 importance of geographic distance (both broad- and fine-scale) and environmental
14 variables in determining community composition and species abundance patterns. The
15 community level analysis was based on interplot geographic distances and interplot
16 differences in floristics and environmental variables. The individual species analysis was
17 based on species abundances, geographic coordinates, and levels of the environmental
18 variables in each plot. Interplot distances were used in the individual species analysis to
19 measure fine-scale spatial autocorrelation among plots. We used one topographic and
20 five soil variables in the analyses. One of the soil variables was clay content, which was
21 employed as a surrogate for the main soil-fertility gradient because it was very strongly
22 correlated with PCA axis 1 ($r = -0.97$) and was available for all 66 plots, rather than just
23 the 40 plots used in the PCA. The other five variables, pH, C:N ratio, water content,

1 phosphate, and slope, were only weakly correlated with the main soil-fertility gradient
2 and with one another. Using Tukey's ladder of powers, the soil and topographic
3 variables were transformed to reduce skewness (Table 1), so that the skewness of all
4 transformed soil variables $< |1|$. Geographic distance was log-transformed to account for
5 the main distance-dependent process, dispersal limitation, which is expected to cause
6 logarithmic distance decay (Condit et al. 2002). Analyses were performed using the R
7 2.1.1 and S-Plus 7.0 statistical packages (www.R-project.org and Insightful Corporation,
8 Seattle, WA, USA), including the S-Plus Spatial Module (Kaluzny et al. 1998).

9

10 *Community level analyses*

11 We used three methods, based on interplot similarity matrices, to assess how
12 changes in tree species composition are related to distance and soil
13 composition/topography at the community level. First, we plotted interplot floristic
14 similarity against interplot geographic distance and interplot similarity in soils. Second,
15 we used Mantel tests to identify correlations between interplot similarity matrices
16 (Legendre & Legendre 1998; Smouse et al. 1986). Mantel tests, which test for linear
17 relationships between interplot differences, were deemed appropriate because plots of
18 floristic similarity versus interplot differences in transformed distance and environmental
19 variables showed no "hump-shaped" relationships. Partial Mantel tests were used to test
20 the strength of relationship between floristic similarity and similarity in soil and
21 topographic variables, while removing the portion of correlations based on interplot
22 geographic distance alone. Third, we used multiple regressions on the interplot-distance
23 matrices, described further below, to assess the relative importance of the soil and

1 topographic variables and to partition the variance in floristic similarity among
2 geographic distance and environmental factors (Legendre et al. 1994; Legendre &
3 Legendre 1998).

4 Interplot floristic similarity was calculated with both Steinhaus' index, which
5 incorporates species abundances, and Sørensen's index, which is based on
6 presence/absence data (Legendre & Legendre 1998). For all three analyses, both indices
7 yielded similar results. The results for Steinhaus' index are included in the text and those
8 based on Sørensen's index can be found in the appendices. For the geographic-distance
9 matrix, log-transformed Euclidean distances between plot-center coordinates were used.
10 For the soil and topographic distance matrices, the differences between transformed
11 values were used.

12 The change in floristic similarity with distance in the BDFFP plots was compared
13 with distance decay in floristic similarity from plot networks in Panama, Ecuador and
14 Peru (Condit et al. 2002; Pitman et al. 2001). The climate in Peruvian and central
15 Amazonian sites average 2200-2300 mm/year with a 3 month dry season. The site in
16 Ecuador is aseasonal with 3200 mm/yr rainfall on average. The Panama sites have the
17 most geographic variation in rainfall (1800 – 3000 mm/rainfall). Methods in these other
18 studies and this one were similar. In all plots, trees ≥ 10 cm dbh were used and all plots
19 were 1 ha in size, except a single plot in Peru which is 0.875 ha. Because the species
20 were identified by different botanists, there is potential for incongruent species
21 identifications among the plot networks.

22 The relative contributions of distance and environmental variables were estimated
23 using multiple regressions on the distance matrices. In this procedure, the distance

1 matrix of interplot floristic similarity is the response variable, and the explanatory
2 variables are interplot geographic distance and similarity in environmental variables. The
3 matrix regression mimics normal regression, except the probability of the strength of
4 relationship between response and explanatory variables is determined by permuting the
5 original matrices 999 times and comparing the distribution of r values from the
6 permutations against the observed value. Both forward and backward selection was used
7 to choose the distance and environmental variables in the final model as follows. First,
8 forward selection was used to add variables at a Bonferroni-corrected significant level of
9 0.005. Then backward elimination was applied to the model with all variables from the
10 forward selection. The resulting final model was the most parsimonious model that
11 explained the greatest variance in floristic similarity (RT) and included both distance and
12 environmental and geographic variables together. We generated two additional models
13 with a reduced number of variables. A distance-only model (RG) removed the
14 environmental variables from the full model and the environment-only model (RE)
15 removed distance. The results of these three models were used to partition the variation
16 in floristic similarity into four fractions: pure geographic (RPG), purely environmental
17 (RPE), mixed geographic-environmental (RPX), and unknown (RUN). These were
18 calculated as follows: $RPG = RT - RE$; $RPE = RT - RG$; $RMX = RE + RG - RT$; and
19 $RUN = 1 - RT$ (cf. Borcard et al. 1992; Duivenvoorden et al. 2002; Tuomisto et al.
20 2003c). All multiple regressions on distance matrices were done using Permute 3.4
21 (Legendre et al. 1994).

22

23 *Individual species level analyses*

1 To assess patterns in individual species, we used a series of regression models that
2 include the effects of environmental and geographic distance variables on species
3 abundances. Then, we used information-theoretic approach to determine which models
4 provided the most explanatory power for the species abundance data (Burnham &
5 Anderson 2002; Svenning et al. 2006). Two types of geographic distance variables were
6 used (Bahn et al. 2006; Cressie 1993; Kaluzny et al. 1998; Lichstein et al. 2002;
7 Svenning et al. 2006). First, the geographic trend variable measures how species
8 abundance is related to geographic positions of the plots in the study area and thus
9 influenced by broad-scale spatial trends. Second, the spatial autocorrelation variable
10 measures how species abundance in one plot is related to abundances in neighboring
11 plots.

12 We used ordinary least square regression (OLS) and conditional autoregressive
13 models (CAR) models in a maximum likelihood framework to determine which
14 combinations of environmental, geographic trend, and spatial autocorrelation variables
15 best fit the patterns of abundance for individual species (Lichstein et al. 2002; Svenning
16 et al. 2006). In species for which local spatial autocorrelation is important, a CAR model
17 will show a large effects of spatial autocorrelation term with a corresponding reduction in
18 either unexplained variance and/or the importance of the broad-scale trend or
19 environmental variables. If spatial autocorrelation is not important, then OLS model will
20 perform as well as the CAR models.

21 We restricted the analyses to 135 relatively common species, defined as having a
22 mean density of at least one individual per hectare and being present in at least half of the
23 40 plots. The species abundance was square-root transformed to reduce skewness,

1 resulting in only four of 135 species with skewness $| > 1 |$. For geographic trend variables
2 (T), we used the UTM coordinates of the plots (X and Y) centered on their means and
3 scaled to unit variance. We used the transformed values of the four most important
4 environmental variables (E) from the community level analysis (Table 1) – the main
5 fertility gradient, slope, pH and C:N ratio - which were centered and their variance
6 scaled.

7 Conditional autoregressive (CAR) models are similar to OLS models, except they
8 include a term in the explanatory variables to model the local spatial dependence in the
9 residuals of the model. The CAR model is specified as $\mathbf{Y} = \mathbf{X}\beta + \rho\mathbf{C}(\mathbf{Y}-\mathbf{X}\beta) + \varepsilon$, where \mathbf{Y}
10 is the vector of observations, \mathbf{X} is the matrix of explanatory variables, β is the vector of
11 regression coefficients, ρ is the spatial autoregression parameter, \mathbf{C} is a symmetric
12 neighborhood matrix of weights (w_{ij}) that quantifies the degree to which the abundance of
13 plot i is influenced by its neighbor j , and ε is vector of random errors. The neighborhood
14 matrix of w_{ij} weights is based on the distance between neighbors and can take different
15 forms depending on the nature of distance dependence in the system. We used the form
16 $w_{ij} = 1/d_{ij}^k$ and tested models that allowed k to vary between 1 and 4. In the end, we used
17 $k=1$ ($w_{ij} = 1/d_{ij}$) because it generated the highest overall the highest r^2 and ρ values.

18 We initially developed a set of eight models, referred to below as the
19 “trend/environment/rho” models, to explore the relative contribution of broad-scale
20 geographic trend (X and Y together), all environment variables together and fine-scale
21 spatial autocorrelation (ρ). The “trend/environment/rho” models included three OLS
22 models that contained the following response variables but lacked the spatial
23 autocorrelation term: intercept, geographic trend and environment (M_{E+T}); intercept and

1 geographic trend (M_T); and intercept and environment (M_E). The three analogous CAR
2 models - $M_{E+T+\rho}$, $M_{T+\rho}$ and $M_{E+\rho}$ - had the same combinations of explanatory variables
3 plus the spatial autocorrelation parameter ρ . The final two models were the ρ -only CAR
4 model (intercept + ρ) (M_ρ) and the null model with only an intercept term (M_0).

5 To explore the contribution of the four environmental variables individually and
6 in all possible combinations, we also built an additional set of OLS-only models that
7 included all permutations of geographic trend (X and Y together) and the four individual
8 soil and topographic parameters. The outcome of this analysis, referred to below as the
9 “all environmental variables” models, resulted in 32 models.

10

11 *Model assessment*

12 The relative support for each model within the two sets of models was determined
13 using Akaike’s Information Criterion (AIC), which takes into account both the relative
14 likelihood (L) of each model and the model complexity, ie number of explanatory
15 variables (k) included in the model (Akaike 1981). Because we had a low number of
16 observations (n) but large number of explanatory variables (k), we used the small-sample
17 bias-corrected form of the AIC, $AIC_c = -2\ln(L) + 2K + 2K(K + 1)/(n - K - 1)$. Because
18 the OLS and CAR functions in S-plus used different error structures and methods to
19 compute likelihoods, we used the approach of Svenning et al. (2006) to calculate the OLS
20 likelihood estimates of the CAR models by adjusting the null CAR model ($\rho=0$) to equal
21 the null OLS model. The model with the lowest AIC_c among the OLS and adjusted CAR
22 likelihoods was considered the best model (AIC_{cmin}). The difference between the AIC_c
23 of each model and the best model (ΔAIC_c) can be used to interpret the strength of support

1 for alternative model compared to the best model. Models with $\Delta AIC_c \leq 2$ have strong
2 support and should be considered along with the best model (Burnham & Anderson
3 2002).

4 Another approach to comparing models is Akaike weights, which can be
5 interpreted as the probability that model i is the best model for the observed data.
6 Akaike weights were calculated as: $W(i) = \exp(-1/2\Delta AIC_c(i)) / \sum \exp(-1/2\Delta AIC_c(i))$
7 where the denominator is summed over all models (Burnham & Anderson 2002). The
8 AIC_c weights were calculated for each species, then averaged over all species to produce
9 a species-averaged Akaike weight. The probability that the best model includes a
10 specific explanatory variable was determined by summing the species-averaged Akaike
11 weights for all models that include that explanatory variable. For the OLS models of the
12 “trend/environment/rho” model, we determined the amount of variance explained by each
13 explanatory variable using the variance partitioning scheme described in the community
14 analysis section of this paper.

15 When there is substantial support for a number of models, the relative importance
16 of individual explanatory variables among the models can be assessed by computing
17 weighted averages of the standardized regression coefficients. This procedure weights
18 the regression coefficient for each parameter β_i in each model by the AIC weight $W(i)$ for
19 that model, then sums over all models $\beta_{MA} = \sum W(i)\beta_i$. Standardized regression coefficients
20 for each explanatory variable were determined for each species, then averaged over
21 species to yield a species-average coefficient for each explanatory variable (Burnham &
22 Anderson 2002).

1 For the “all environmental variables” models only, we tallied the number of
2 common species that had significant relationships with each environmental variable using
3 a Bonferroni-adjusted p-value. First, for each species, we determined if the individual
4 environmental variables included in the best model were significant. Then for each
5 environmental variable, we summed significant relationship over the species. The same
6 procedure was done for models with strong support.

7

8 **Results**

9 *Community composition*

10 Geographic distance was the most important variable shaping floristic similarity
11 in the central Amazonian plots. A Mantel test between floristic similarity (Steinhaus'
12 index) and geographic distance yielded a relatively strong correlation ($r= 0.57$), which
13 was mostly due to the sharp decline in floristic similarity between 0 - 5 km interplot
14 distances (Fig. 1). Interplot floristic similarity was also correlated with interplot similarity
15 in the soil and topographic variables (Table 1, Fig. 2). Half of the six soil and
16 topographic variables (main fertility gradient, pH and C:N ratio) showed a significant
17 correlation with floristic similarity using the simple Mantel tests. [Slope was not
18 significant in the 40 plots, but was significant in the full 66 plots (Appendix 3).] Using
19 partial Mantel tests to account for trends in floristic similarity related solely to geographic
20 distance, only the main fertility gradient and C:N ratio were significant with correlation
21 coefficients at 0.41 and 0.21, respectively (Table 1).

22 In the multiple regression analysis considering all the explanatory variables
23 together, the final model contained interplot geographic distance and the main fertility

1 gradient (Table 2, Appendix 4). Geographic distance had a higher standardized regression
2 coefficient than the main fertility gradient. Partitioning the variance, pure geographic
3 distance explained 26% of the community floristic variation, environmental factors (soil
4 and topographic variables) explained 12%, and the interaction of distance and
5 environment explained 7%, with 56% of the variation unexplained.

6 The plots in central Amazonia had lower overall beta diversity than did
7 comparable plot networks in Peru, Ecuador, and Panama (Fig. 1). At small interplot
8 distances (0-2.5 km), the three Amazonian networks had comparable floristic similarity
9 between 0.45 and 0.52 fraction of shared species (the similarity of the Panama plots was
10 much higher), but they diverged at larger distances (>2.5 km). At interplot distance 10 -
11 30 km, the similarity of the central Amazon plots remained fairly constant at 0.45,
12 whereas the similarity within the Peruvian plots decreases steeply from 0.45 to 0.32. The
13 floristic similarity of the Ecuador and Panama plots between 10 and 30 km was
14 considerably lower than the central Amazon between 0.25 – 0.35 shared species.

15

16 *Individual species*

17 As in the community level analysis, geographic trend had somewhat greater
18 importance than environmental variables in determining individual species abundances.
19 Geographic trend was included in the best model for 41% of species compared to 30%
20 for the environmental variables (Table 3, Appendix 5). However, considering all models
21 with strong support, which are plausible alternatives to the best model, 89% of species
22 contained the environmental variables compared to 59% that included the geographic
23 trend variables (Table 3). The summed AIC weights, which is a measure of the weight of

1 evidence that a variable should be included in the final model, were somewhat greater for
2 broad-scale geographic trend (46%) than for the environmental variables (35%) (Table
3 3).

4 There was some, but not strong, support that spatial autocorrelation (ρ), ie fine-
5 scale structuring, is important in this data set. The summed AIC weight for any model
6 with ρ was 37%, equal to that of models containing the environmental variables (Table
7 3). But spatial autocorrelation was included in only 15% of the best models (CAR
8 models); 55% of best models had no rho parameter (OLS models) (Table 3). In 80-95%
9 of species where a specific CAR models (for example, geographic trend + ρ) had strong
10 support, the corresponding OLS model (for example, just geographic trend) also had
11 strong support, suggesting that either a CAR or OLS model could be chosen for each
12 species. The null model had quite strong support with an AIC weight of 20% (Table 3).
13 For 30% of species, the null model was the best model, and for over half the species, the
14 null model had strong support, meaning that a model that included at least one
15 environmental or distance variable was only slightly better than a model with just an
16 intercept.

17 The partitioning of the variance from the OLS models showed that 15% of the
18 variance was explained by environmental variables, 8% by geographic trend and 8% by
19 trend + environment. On average across species, the large majority of variance in species
20 abundance (70%) was unexplained. For just the CAR models, the average variance
21 explained (r^2) for the full model (RT) was 33%, thus the average unexplained variance
22 (1-RT) was 67%, similar to the unexplained variance in the OLS models.

23

1 *Importance of individual environmental variables*

2 We used the “all environmental variables” OLS models to determine the relative
3 importance of the four environmental variables for species abundances. Of the four
4 environmental variables, the main fertility gradient and pH were more important in
5 structuring species abundance than slope and C:N ratio. Both the main fertility gradient
6 and pH were included in the best model for 39% of the species (Table 4, Appendix 5, 6).
7 Also, the main fertility gradient and pH had higher AIC_c weights than the other two
8 environmental variables (Table 4). It is important to note, however, the data could not
9 definitively arbitrate between the different models, as each environmental variable was
10 included in at least one model that had strong support.

11 In the models with the best support, the main fertility gradient had a significant p-
12 value for the most number of species (16%) compared to pH (9%), C:N ratio (3%) and
13 slope (1%). Similarly, for models with strong support, the significant p-values were
14 greatest for the main fertility gradient (21%), followed by pH (10%), slope (5%) and C:N
15 ratio (4%). One quarter of the species had at least one environmental variable that was
16 significant in the best model and 33% of the species had at least one significant
17 environmental variable included in a model with strong support. There was evidence that
18 certain species-rich families and genera have strong trends with particular environmental
19 variables. The seven common Moraceae, 11 *Eschweilera* (Lecythidaceae), and nine
20 *Protium* (Burseraceae) species showed a strong relationship with the main fertility
21 gradient (c.f. Fine et al. 2004, who also found clay and sand specialists in the genus
22 *Protium*), the six *Licania* (Chrysobalanaceae) species with slope, and the five
23 Myristicaceae species with C:N ratio (Appendix 5).

1 Finally, for model-averaged standardized regression parameters in the OLS
2 models, the main fertility gradient and pH had the highest absolute coefficients averaged
3 over all species (Table 5). Of the species with high coefficients (>0.10) for slope (32
4 species), 72% had positive coefficients, indicating that species tended to have greater
5 abundance on steeper slopes. The main fertility gradient also had more species with
6 positive than negative coefficients (Table 5). Of the species-rich taxa, *Licania* had high
7 abundance at steeper slopes and *Eschweilera* tended to have higher abundance at sites
8 with low clay content, and low amounts of nitrogen, carbon and cations (Appendix 5).
9 pH and C:N ratio had more equal numbers of species with strong negative and positive
10 relationships (Table 5).

11 We used the value of 34% as the number of species with a significant relationship
12 with environmental variables to compare the results of this study to other studies. This
13 was the percentage of species that had significant relationships with any environmental
14 variables in the best or strongly-supported in the “all environmental variables” model.
15 The 34% value also matches the percentage of species for which environment was
16 included in the best model for the “trend/environment/rho” model (Table 3).

17

18 **Discussion**

19 *Community floristic similarity*

20 In our study, both geographic distance and environmental variables were
21 important in determining community floristics patterns. The effect of geographic
22 distance on floristic similarity was strongest at distances of 0-5 km, but then became
23 relatively unimportant at distances of 5-30 km. A steep decline in similarity at local

1 distances and a more gradual decline at larger distances has been reported throughout the
2 Neotropics (Condit et al. 2002; Normand et al. 2006; Tuomisto et al. 2003c; Vormisto et
3 al. 2004), and is consistent with the important role of dispersal limitation in determining
4 beta diversity (Condit et al. 2002).

5 Of the soil and topographic variables, the main soil-fertility gradient influenced
6 floristic similarity at all spatial scales, with widely separated sites on similar soils tending
7 to be more similar floristically than are nearby sites on differing soils (see also Rankin-de
8 Merona et al. 1992). The dominant soil trend in our central Amazonian landscape
9 appears to be a gradient between clay-rich soils with higher organic matter, cation
10 concentrations and nitrogen, and sand-dominated soils with opposite characteristics
11 (Chauvel et al. 1987; Laurance et al. 1999; Richter & Babbar 1991). Comparable trends
12 are found in other tropical forests where soil texture and cation concentrations strongly
13 influence plant species composition (Phillips et al. 2003; Tuomisto et al. 2003b;
14 Tuomisto et al. 2003a; Webb & Peart 2000). The C:N ratio of soils, which reflects soil
15 nitrogen mineralization, also had a impact on community floristics. The relationship
16 between C:N ratio and floristic composition has not been reported in other tropical
17 forests, but can be important in temperate forests (Elgersman & Dhillion 2002; Schuster
18 & Diekman 2005).

19

20 *Landscape and regional patterns*

21 The plots in central Amazonia exhibit two notable trends compared to the plot
22 networks in northwestern and southwestern Amazonia: consistent levels of floristic
23 similarity for plots >5 km apart and relatively low beta diversity (Condit et al. 2002;

1 Macia & Svenning 2005; Pitman et al. 2001). The differences in floristic similarity
2 between the sites may result from different dispersal or speciation dynamics in central
3 Amazon versus western Amazon forests (Condit et al. 2002). The uniformity in floristic
4 similarity with increasing distance suggests that the central Amazonian landscape is
5 dominated by a small group or “oligarchy” of common species (cf. Pitman et al. 2001).
6 Similar to the western Amazon sites, 61% of the stems in central Amazonia are common
7 species with densities of at least one stem per hectare. The oligarchy of common species
8 on nutrient-starved soils of central Amazonia tends to be quite distinct from that of the
9 geologically younger and richer soils of the western Amazon (Gentry 1990; Terborgh &
10 Andresen 1998). Yasuní and Manu share 42 of their 150 commonest species (Pitman et
11 al. 2001), but the central Amazon shares only 18 common species with Yasuni and 8
12 species with Manu. Although the common species of central Amazonia do not overlap
13 with the common species in western Amazonia, we do not expect the common central
14 Amazonian species to be specialist species localized to the BDFFP plots, as common
15 species in other Neotropical forests tend to be widespread regionally and locally abundant
16 (Duque et al. 2003; Eilu et al. 2004; Pitman et al. 2001).

17

18 *Distance vs. environment in individual species*

19 Broad-scale geographic trend and the environmental variables were both
20 important in determining individual species distributions. Spatial autocorrelation was
21 less important, found in only 15% of the best models. The majority of the variance in
22 individual species was not explained by geographic distance, environmental variables or

1 spatial autocorrelation, indicating that importance of stochastic processes and possibly
2 unmeasured environmental variables for species distributions.

3 Because of the high community floristic similarity at interplot distances less than
4 5 km, we expected distance variables to be important for individual species. Fine-scale
5 spatial autocorrelation was less important and broad-scale trend more important than we
6 expected. The lack of spatial autocorrelation may be due to the relatively large distances
7 between plots. The minimum distance between plots was 100 m, whereas density
8 dependent processes, such as dispersal limitations and Janzen-Connell effects, influence
9 tree densities and dynamics at shorter spatial scales of 30 m or less (Condit et al. 2000,
10 Hubbell et al. 2001, Queenborough et al. 2007).

11 The strong relationship between broad scale spatial trend and abundance for many
12 species could have several underlying environmental or historical causes. The flora of
13 the Manaus area is composed of species from several phytogeographic regions and many
14 species are at the limit of their known range boundaries (Oliveira & Daly 1999). The
15 expansion or contraction of these boundaries may leave geographic trends in some
16 species' densities. The geographic trend in species abundance could result from
17 unmeasured environmental variation. The western portion of the BDFFP is part of a
18 different watershed and geological formation than the eastern portion. The road between
19 Manaus and Caracas, which was constructed in the late 1960's, runs through the western
20 portion of the study area, although not within 5 km of any plot. Although the wide
21 disturbed right-of-way of this road could be a source of pioneer species, we believe that
22 the likelihood is small that this factor could have affected the populations of trees ≥ 10
23 cm dbh in the study plots by the 1980s when our survey data were collected. Clues to the

1 causes of the geographic variation might be obtained by investigating the species-rich
2 families and genera that showed strong relationship geographic trend. In particular, the
3 nine common *Protium* and 16 common Lecythidaceae species showed a strong
4 relationships with geographic trend (Appendix 5).

5

6 *Important environmental variables for species abundances*

7 As with community floristic similarity, the main fertility gradient was the most
8 important environmental variable in relation to individual species abundances. However,
9 pH was clearly the second most important environmental variable, in contrast to the
10 community floristic similarity analysis in which C:N ratio was the second most important
11 explanatory variable. Unlike cation concentration and soil texture, which seem to be
12 widely important in influencing floristic patterns in tropical forests, the role of soil pH
13 may be limited to certain regions or taxa. Soil pH has been found to have significant
14 relationships with floristics in some tropical forests (Baillie 1987; Hall et al. 2004; Jones
15 et al. 2006) but is unimportant in others (Paoli et al. 2006; Ruokolainen & Tuomisto
16 1998; Tuomisto et al. 2003a), including the Ducke Reserve in central Amazonia (Costa et
17 al. 2005). Higher species abundances on soils of low pH might be expected due to the
18 evolutionary origin of Amazonian flora of soils of low pH (Pärtel 2002). However, there
19 were equal numbers of species with positive and negative species coefficients for pH,
20 which does not lend support for greater species abundances at lower pH in this
21 community.

22 It is important to note that for each species, each environmental variable was
23 included in at least one model with strong support (Table 4), indicating that no single

1 model definitively fit the data better than the other models. To narrow the candidates for
2 the best fit model would require either more data or reducing the number of models under
3 consideration (32 for the “all environmental variables” set of models). The analysis
4 provided here can help guide the choice of the appropriate environmental variables for
5 different species, genera and families for further study.

6

7 *Meta-analyses across tropical forests*

8 The proportion of species having significant relationships with soil factors varied
9 widely (25-82%) among studies of tropical forests (Fig. 3, Appendix 7). The
10 methodologies of different studies varied tremendously, making it difficult to compare
11 among them. For example, the number of plant taxa analyzed ranged from just a single
12 genus (*Entandrophragma*) (Hall et al. 2004), a single family (Lauraceae,
13 Dipterocarpaceae, Sterculiaceae) (Paoli et al. 2006; Sri-ngernyuang et al 2003; Yamada
14 et al. 2006), and a single plant form (ferns or palms) (Svenning et al.1999; Tuomisto et al.
15 2002), to all trees (Cannon & Leighton 2004; Clark et al. 1998, 1999; Harms et al. 2001;
16 Miyamoto et al. 2003; Phillips et al. 2003; Webb & Peart 2000). Despite these
17 differences among studies (Appendix 7), one trend was apparent. The percentage of
18 species with significant soil and topographic relationships was negatively correlated with
19 the total number of species examined (Fig. 3). Among four studies that examined 100 or
20 more species, all had a comparatively low percentage (25-55%) of species with
21 significant soil or topographic relationships. The other ten studies had 55 or fewer
22 species and the range of species with significant soil or topographic relationships was
23 higher, between 65-82% for nine of the ten studies. Our study, in which 34% of the

1 species were significantly related to at least one soil variable, was within the range of
2 studies examining a large number of species. Other factors, such as the taxa, geographic
3 region, soil variables or habitat type, study-area size, or total plot-area sampled, did not
4 influence the proportion of identified species. It is possible that studies with a small
5 number of species focus on species with previous observations of habitat specialization.
6 Another possibility is that the studies with large numbers of species contain more rare
7 species and fail to detect environmental relationships in the rare because of low sample
8 size. The environmental variation within each study site was difficult to compare with
9 the information provided in the literature, but would certainly be an important next step
10 in a meta-analysis. To definitively study community floristics relationships with soil and
11 topographic variables, more studies with large numbers of species and with soil data
12 collected in a standardized manner are needed.

13

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20

21 **References**

22 Akaike, H. 1981. Modern development of statistical methods. in Eykhoff, P., (eds.)
23 *Trends and progress in systems identification*. pp. 169-184, Pergamon Press, London.

- 1 Bahn, V., O'Connor, R. J. & Krohn, W. B. 2006. Importance of spatial autocorrelation in
2 modeling bird distributions at a continental scale. *Ecography* 29:835-844.
- 3 Baillie, I. C., Ashton, P. S., Court, M. N., Anderson, J. A. R., Fitzpatrick, E. A. &
4 Tinsley, J. 1987. Site characteristics and the distribution of tree species in mixed
5 dipterocarp forest on tertiary sediments in Central Sarawak, Malaysia. *Journal of*
6 *Tropical Ecology* 3:201-220.
- 7 Beinroth, F. H. 1975. Relationships between U.S. soil taxonomy, the Brazilian system,
8 and FAO/UNESCO units. in Bornemisza, E. & Alvarado, A., (eds.) *Soil Management*
9 *in Tropical America*. pp. 97-108, North Carolina State University, Raleigh, NC.
- 10 Borcard, D., Legendre, P. & Drapeau, P. 1992. Partialling out the spatial component of
11 ecological variation. *Ecology* 73:1045-1055.
- 12 Brown, K. S. 1987. Soils and vegetation. in Whitmore, T. C. & Prance, G. T., (eds.)
13 *Biogeography and Quaternary History in Tropical America*. pp. 19-45, Oxford
14 Monographs in Biogeography, Oxford, UK.
- 15 Burnham, K. P. & Anderson, D. R. 2002. *Model selection and multi-model inference: a*
16 *practical information-theoretic approach*. Springer-Verlag, New York.
- 17 Cannon, C. H. & Leighton, M. 2004. Tree species distributions across five habitats in a
18 Bornean rain forest. *Journal of Vegetation Science* 15:257-266.
- 19 Chauvel, A., Lucas, Y. & Boulet, R. 1987. On the genesis of the soil mantle of the region
20 of Manaus, Central Amazonia, Brazil. *Experientia* 43:234-240.
- 21 Chust, G., Chave, J., Condit, R., Aguilar, S., Lao, S. & Perez, R. 2006. Determinants and
22 spatial modeling of tree beta-diversity in a tropical forest landscape in Panama.
23 *Journal of Vegetation Science* 17:83-92.

- 1 Clark, D. B., Clark, D. A. & Read, J. M. 1998. Edaphic variation and the mesoscale
2 distribution of tree species in a neotropical rain forest. *Journal of Ecology* 86:101-
3 112.
- 4 Clark, D. B., Palmer, M. W. & Clark, D. A. 1999. Edaphic factors and the landscape-
5 scale distributions of tropical rain forest trees. *Ecology* 80:2662-2675.
- 6 Condit, R., Ashton, P. S., Baker, P., Bunyavejchewin, S., Gunatilleke, S., Gunatilleke, N.,
7 Hubbell, S. P., Foster, R. B., Itoh, A., LaFrankie, J. V., Lee, H. S., Losos, E.,
8 Manokaran, N., Sukumar, R. & Yamakura, T. 2000. Spatial patterns in the
9 distribution of tropical tree species. *Science* 288:1414-1418.
- 10 Condit, R., Pitman, N., Leigh, E., Chave, J., Terborgh, J., Foster, R. B., Nunez, P.,
11 Aguilar, S., Valencia, R., Villa, G., Losos, E., Muller-Landau, H. & Hubbell, S. P.
12 2002. Beta diversity in tropical forest trees. *Science* 295:666-669.
- 13 Costa, F. R. C., Magnusson, W. E. & Luizao, R. C. 2005. Mesoscale distribution patterns
14 of Amazonian understorey herbs in relation to topography, soil and watersheds.
15 *Journal of Ecology* 93:863-878.
- 16 Cressie, N. A. C. 1993. *Statistics for spatial data*. John Wiley, New York.
- 17 Duivenvoorden, J. F., Svenning, J.-C. & Wright, S. J. 2002. Beta diversity in tropical
18 forests. *Science* 295:636-637.
- 19 Duque, A., Cavelier, J. & Posada, A. 2003. Strategies of tree occupation at a local scale
20 in terra firme forests in the Colombian Amazon. *Biotropica* 35:20-27.
- 21 Eilu, G., Hafashimana, D. L. N. & Kasenene, J. M. 2004. Tree species distribution in
22 forests of the Albertine Rift, western Uganda. *African Journal of Ecology* 42:100-
23 110.

- 1 Elgersman, A. M. & Dhillion, S. S. 2002. Geographical variability of relationships
2 between forest communities and soil nutrients along a temperature-fertility gradient in
3 Norway. *Forest Ecology and Management* 158:155-168.
- 4 Fearnside, P. M. & Leal-Filho, N. 2001. Soil and development in Amazonia: lessons
5 from the Biological Dynamics of Forest Fragments Project. in Bierregaard, R. O.,
6 Gascon, C., Lovejoy, T. E. & Mesquita, R., (eds.) *Lessons from Amazonia: The*
7 *Ecology and Conservation of a Fragmented Forest*. pp. 291-312, Yale University
8 Press, New Haven, CT.
- 9 Fine, P. V. A., Mesones, I. & Coley, P. D. 2004. Herbivores promote habitat
10 specialization by trees in Amazonian forests. *Science* 305:663-665.
- 11 Gentry, A. H. 1990. Floristic similarities and differences between Southern Central
12 America and Upper and Central Amazonia. in Gentry, A. H., (eds.) *Four Neotropical*
13 *Rainforests*. pp. 141-157, Yale University Press, New Haven, CT.
- 14 Hall, J. S., McKenna, J. J., Ashton, P. M. S. & Gregoire, T. G. 2004. Habitat
15 characterizations underestimate the role of edaphic factors controlling the distribution
16 of *Entandrophragma*. *Ecology* 85:2171-2183.
- 17 Harms, K. E., Condit, R., Hubbell, S. P. & Foster, R. B. 2001. Habitat associations of
18 trees and shrubs in a 50-ha neotropical forest plot. *Journal of Ecology* 89:947-959.
- 19 Hubbell, S. P. 2001. *The Unified Theory of Biodiversity and Biogeography*. Princeton
20 University Press, Princeton, New Jersey, USA.
- 21 Hubbell, S. P., Ahumada, J. A., Condit, R. & Foster, R. B. 2001. Local neighborhood
22 effects on long-term survival of individual trees in a neotropical forest. *Ecological*
23 *Research* 16:859-875.

- 1 Jones, M. M., Tuomisto, H., Clark, D. B. & Olivas, P. 2006. Effects of mesoscale
2 environmental heterogeneity and dispersal limitation on floristic variation in rain
3 forest ferns. *Journal of Ecology* 94:181-195.
- 4 Kaluzny, S. P., Vega, S. C., Cardoso, T. P. & Shelly, A. A. 1998. *S+ Spatial Stats User's*
5 *Manual*. MathSoft, Seattle.
- 6 Laurance, W. F., Fearnside, P. M., Laurance, S. G., Delamonica, P., Lovejoy, T. E. &
7 Rankin, d. 1999. Relationship between soils and Amazon forest biomass: a landscape-
8 scale. *Forest Ecology and Management* 118:1-3.
- 9 Laurance, W. F., Lovejoy, T. E., Vasconcelos, H. L., Bruna, E., Didham, R., Stouffer, P.,
10 Gascon, C., Bierregaard, R., Laurance, S. & Sampaio, E. 2002. Ecosystem decay of
11 Amazonian forest fragments: a 22-year investigation. *Conservation Biology* 16:605-
12 618.
- 13 Laurance, W. F., Oliveira, A. A., Laurance, S. G., Condit, R., Dick, C. W., Andrade, A.,
14 Nascimento, H. E. M. & Lovejoy, T. E. 2005. Altered tree communities in
15 undisturbed Amazonian forests: A consequence of global change? *Biotropica* 37:160-
16 162.
- 17 Legendre, P., Lapointe, F.-J. & Casgrain, P. 1994. Modeling brain evolution from
18 behaviour: a permutational regression approach. *Evolution* 48:1487-1499.
- 19 Legendre, P. & Legendre, L. 1998. *Numerical Ecology*. Elsevier, Amsterdam.
- 20 Lichstein, J. W., Simons, T. R., Shriner, S. A. & Franzreb, K. E. 2002. Spatial
21 autocorrelation and autoregressive models in ecology. *Ecological Monographs*
22 72:445-463.

- 1 Lovejoy, T. E., Bierregaard, R. O., Rylands, A. B., Malcolm, J. R., Quintela, C. E.,
2 Harper, L. H., Brown, K. S., Powell, A. H., Powell, G. V. N., Schubart, H. O. R. &
3 Hays, M. B. 1986. Edge and other effects of isolation on Amazon forest fragments. in
4 Soule, M. E., (eds.) *Conservation Biology: The Science of Scarcity and Diversity*. pp.
5 257-285, Sinauer, Sunderland, MA.
- 6 Macia, M. J. & Svenning, J. C. 2005. Oligarchic dominance in western Amazonian plant
7 communities. *Journal of Tropical Ecology* 21:613-626.
- 8 Miyamoto, K., Suzuki, E., Kohyama, T., Seino, T., Mirmanto, E. & Simbolon, H. 2003.
9 Habitat differentiation among tree species with small-scale variation of humus depth
10 and topography in a tropical heath forest of Central Kalimantan, Indonesia. *Journal of*
11 *Tropical Ecology* 19:43-54.
- 12 Nelson, B. W. 2005. Pervasive alteration of tree communities in undisturbed Amazonian
13 forests. *Biotropica* 37:158-159.
- 14 Nelson, B. W., Kapos, V., Adams, J. B., Oliveira, W. J., Braun, O. P. g. & Amaral, I. L.
15 1994. Forest disturbance by large blowdowns in the Brazilian Amazon. *Ecology*
16 75:853-858.
- 17 Normand, S., Vormisto, J., Svenning, J.-C., Grandez, C. & Balslev, H. 2006.
18 Geographical and environmental controls of palm beta diversity in paleo-riverine
19 terrace forests in Amazonian Peru. *Plant Ecology* 186:161-176.
- 20 Oliveira, A. A. & Daly, D. C. 1999. Geographic distribution of tree species occurring in
21 the region of Manaus, Brazil: implications for regional diversity and conservation.
22 *Biodiversity and Conservation* 8:1245-1259.

- 1 Oliveira, A. A. & Mori, S. A. 1999. A central Amazonian terra firme forest. I. High tree
2 species richness on poor soils. *Biodiversity and Conservation* 8:1219-1244.
- 3 Palmer, M. W. 2006. Distance decay in an old-growth neotropical forest. *Journal of*
4 *Vegetation Science* 16:161-166.
- 5 Paoli, G. D., Curran, L. M. & Zak, D. R. 2006. Soil nutrients and beta diversity in the
6 Bornean Dipterocarpaceae: evidence for niche partitioning by tropical rain forest
7 trees. *Journal of Ecology* 94:157-170.
- 8 Pärtel, M. 2002. Local plant diversity patterns and evolutionary history at the regional
9 scale. *Ecology* 83:2361-2366.
- 10 Phillips, O. L., Nunez Vargas, P., Monteagudo, A. L., Pena Cruz, A., Chuspse Zans, M.-
11 E., Galiano Sanchez, W., Li-Halla, M. & Rose, S. 2003. Habitat association among
12 Amazonian tree species: a landscape-approach. *Journal of Ecology* 91:757-775.
- 13 Piperno, D. R. & Becker, P. 1996. Vegetational history of a site in the Central Amazon
14 basin derived from phytolith and charcoal records from natural soils. *Quaternary*
15 *Research* 45:202-209.
- 16 Pitman, N. C. A., Terborgh, J. W., Silman, M. R., Nunez Vargas, P., Neill, D. A., Ceron,
17 C. E., Palacios, W. A. & Aulestia, M. 2001. Dominance and distribution of tree
18 species in upper Amazonian terra firme forests. *Ecology* 82:2102-2117.
- 19 Potts, M. D., Ashton, P. S., Kaufman, L. S. & Plotkin, J. B. 2002. Habitat patterns in
20 tropical rain forests: A comparison of 105 plots in Northwest Borneo. *Ecology*
21 83:2782-2797.

- 1 Queenborough, S. A., Burslem, D. F. R. P., Garwood, N. C. & Valencia, R. 2007.
2 Neighborhood and community interactions determine the spatial pattern of tropical
3 seedling survival. *Ecology* 88:2248-2258.
- 4 Rankin-de Merona, J. M., Prance, J. M., Hutchings, R. W., Silva, M. F., Rodrigues, W.
5 A. & Uehling, M. A. 1992. Preliminary results of a large scale inventory of upland
6 rain forest in the central Amazon. *Acta Amazonica* 22:493-534.
- 7 Richter, D. D. & Babbar, K. I. 1991. Soil diversity in the tropics. *Advances in Ecological*
8 *Research* 21:315-389.
- 9 Ruokolainen, K. & Tuomisto, H. 1998. Vegetacion natural de la zona de Iquitos. in
10 Kalliola, R. & Flores Paitan, S., (eds.) *Geoecologia y desarrollo amazonico: estudio*
11 *integrado en la zona de Iquitos, Peru*. pp. 253-365.
- 12 Santos, G. M., Gomes, P. R. S., Anjos, R. M., Cordeiro, R. C., Turcq, B. J., Sifeddine, A.,
13 de Tada, M. L., Creswell, R. G. & Fifield, L. K. 2000. ^{14}C AMS dating of fires in the
14 central Amazon rain forest. *Nuclear Instruments and Methods in Physics Research B*
15 172:761-766.
- 16 Schuster, B. & Diekmann, M. 2005. Species richness and environmental correlates in
17 deciduous forests of Northwest Germany. *Forest Ecology and Management* 206:197-
18 205.
- 19 Smouse, P. E., Long, J. C. & Sokal, R. R. 1986. Multiple regression and correlation
20 extensions of the Mantel test of matrix correspondence. *Systematic Zoology* 35:627-
21 632.
- 22 Sri-Ngernyuang, K., Kanzaki, M., Mizuno, T., Noguchi, H., Teejuntuk, S., Sungpalee, C.,
23 Hara, M., Yamakura, T., Sahunalu, P., Dhanmanonda, P. & Bunyavejchewin, S.

- 1 2003. Habitat differentiation of Lauraceae species in a tropical lower montane forest
2 in northern Thailand. *Ecological Research* 18:1-14.
- 3 Svenning, J. C. 1999. Microhabitat specialization in a species-rich palm community in
4 Amazonian Ecuador. *Journal of Ecology* 87:55-65.
- 5 Svenning, J.-C., Engelbrecht, B. M. J., Kinner, D. A., Kursar, T. A., Stallard, R. F. &
6 Wright, S. J. 2006. The relative roles of environment, history and local dispersal in
7 controlling the distributions of common tree and shrub species in a tropical forest
8 landscape, Panama. *Journal of Tropical Ecology* 22:575-586.
- 9 ter Steege, H., Pitman, N. C. A., Phillips, O. L., Chave, J., Sabatier, D., Duque, A.,
10 Molino, J.-F., Prevoist, M.-F., Spichiger, R., Castellanos, H., von Hildebrand, P. &
11 Vasquez, R. 2006. Continental-scale patterns of canopy tree composition and function
12 across Amazonia. *Nature* 443:444-447.
- 13 Terborgh, J. & Andresen, E. 1998. The composition of Amazonian forests: patterns at
14 local and regional scales. *Journal of Tropical Ecology* 14:645-664.
- 15 Tuomisto, H., Poulsen, A. D., Ruokolainen, K., Moran, R. C., Quintana, C., Celi, J. &
16 Canas, G. 2003a. Linking floristic patterns with soil heterogeneity and satellite
17 imagery in Ecuadorian Amazonia. *Ecological Applications* 13:352-371.
- 18 Tuomisto, H., Ruokolainen, K., Aguilar, M. & Sarmiento, A. 2003b. Floristic patterns
19 along a 43-km long transect in an Amazonian rain forest. *Journal of Ecology* 91:743-
20 756.
- 21 Tuomisto, H., Ruokolainen, K., Poulsen, A. D., Moran, R. C., Quintana, C., Canas, G. &
22 Celi, J. 2002. Distribution and diversity of pteridophytes and Melastomataceae along

- 1 edaphic gradients in Yasuni National Park, Ecuadorian Amazonia. *Biotropica* 34:516-
- 2 533.
- 3 Tuomisto, H., Ruokolainen, K. & Yli-Halla, M. 2003c. Dispersal, environment, and
- 4 floristic variation of Western Amazonian forests. *Science* 299:241-244.
- 5 Valencia, R., Foster, R. B., Villa, G., Condit, R., Svenning, J. C., Hernandez, C.,
- 6 Romoleroux, K., Losos, E., Magard, E. & Balslev, H. 2004. Tree species distributions
- 7 and local habitat variation in the Amazon: large forest plot in eastern Ecuador.
- 8 *Journal of Ecology* 92:214-229.
- 9 Vormisto, J., Svenning, J. C., Hall, P. & Balslev, H. 2004. Diversity and dominance in
- 10 palm (Arecaceae) communities in terra firme forests in the western Amazon basin.
- 11 *Journal of Ecology* 92:577-588.
- 12 Webb, C. O. & Peart, D. R. 2000. Habitat associations of trees and seedlings in a
- 13 Bornean rain forest. *Journal of Ecology* 88.
- 14 Yamada, T., Tomita, A., Itoh, A., Yamakura, T., Ohkubo, T., Kanzaki, M., Tan, S. &
- 15 Ashton, P. S. 2006. Habitat associations of Sterculiaceae trees in a Bornean rain
- 16 forest plot. *Journal of Vegetation Science* 17:559-566.

1 Table 1. Results of simple and partial Mantel tests between floristic similarity (as
 2 measured by Steinhaus' index), interplot differences in environmental variables and
 3 interplot geographic distance in central Amazonia. The environmental variables and
 4 distance have been transformed. The Mantel correlation coefficient (r) and
 5 significance level (p) is shown for each test. Relationships in bold are statistically
 6 significant at a Bonferroni-adjusted level of $p < 0.007$.

Independent Variable	Transfor- mation	Variable ~		Floristic Similarity ~		Floristic Similarity ~	
		Distance		Variable		Distance	
		r	p	r	p	r	p
Geographic distance	$\log_{10}(x)$	--	--	0.57	0.0005	--	--
Main fertility gradient	x^2	0.17	0.0020	0.43	0.0005	0.41	0.0005
pH	x	0.41	0.0005	0.31	0.0005	0.10	0.0640
Slope	$x^{0.5}$	0.06	0.0645	0.15	0.0220	0.14	0.0340
C:N ratio	$-x^{-3}$	0.19	0.0005	0.28	0.0005	0.21	0.0030
Water	$x^{0.5}$	0.06	0.0630	0.17	0.0090	0.16	0.0115
Phosphate	x^2	0.09	0.0240	0.11	0.0445	0.07	0.1280

1 Table 2. Standardized regression coefficients from multiple regression analyses with
 2 interplot floristic similiary (Steinhaus index) as the response variable and interplot
 3 geographic distance and interplot variation in six environmental variables as the
 4 explanatory variables. A Bonferroni-corrected p-value of 0.005 was used in forward and
 5 backward selection to derive the final model from the full model.
 6

	Full model	Final model
	Coefficient (p)	Coefficient (p)
<hr/>		
Full model		
Distance	0.43 (<0.0001)	0.52 (0.0010)
Main fertility gradient	0.33 (<0.0001)	0.34 (0.0010)
pH	0.14 (0.0040)	
C:N ratio	0.09 (0.0340)	
Water	0.09 (0.0420)	
Slope	0.05 (0.1690)	
Phosphate	0.10 (0.0190)	
r^2	0.49 (0.0001)	0.44 (0.0005)

Table 3. Comparison of “trend/environment/rho” models with all possible combinations of broad-scale geographic trend, environmental variables (main fertility gradient, slope, pH and C:N ratio together) and fine-scale spatial autocorrelation (ρ). The AIC_c weights were averaged over all species. The AIC_c weights can be interpreted as the probability that the model is the best model in this set for the observed data.

Model	All models	Best Model	Best model or strong support	Best model
	AIC _c weights	# spp	# spp	r ²
Geographic trend + environment	8%	9 (7%)	37 (27%)	0.46
Geographic trend	22%	41 (30%)	97 (72%)	0.23
Environment	14%	24 (18%)	65 (48%)	0.34
Geographic trend + environment + ρ	5%	1 (1%)	30 (22%)	NA
Geographic trend + ρ	12%	7 (5%)	48 (36%)	0.46
Environment + ρ	8%	5 (4%)	76 (56%)	0.45
P	12%	7 (5%)	71 (53%)	0.18
Null	20%	41 (30%)	75 (56%)	NA
Geographic trend in any model	46%	56 (41%)	79 (59%)	0.30
Environment in any model	35%	41 (30%)	120 (89%)	0.39
ρ in any model (CAR)	37%	20 (15%)	135 (100%)	0.37
No ρ in model (OLS)	44%	74 (55%)	133 (99%)	0.30

Table 4. Summary of results from “all environmental variables” ordinary least squares (OLS) models, which includes all possible combinations of geographic trend and the four environmental variables as explanatory variables resulting in 32 models. For each species, AIC_c weights were calculated as the sum of AIC_c weights from all models that include the variable. Then the AIC_c weight for each variable was averaged over the species. Individual model results are given in Appendix A2.

Model	Sum of AIC_c weights	Best model # spp	Best model or strong support # spp
Null	5.1%	18 (13%)	58 (43%)
Main fertility gradient in any model	49.0%	52 (39%)	135 (100%)
Slope in any model	39.8%	33 (24%)	135 (100%)
pH in any model	46.2%	53 (39%)	135 (100%)
C:N ratio in any model	36.2%	20 (15%)	134 (99%)
Geographic trend in any model	43.5%	49 (36%)	134 (99%)
Environmental variable in any model	89.6%	104 (77%)	135 (100%)

Table 5. Weighted standardized regression coefficients for individual environmental variables averaged across all species. To get these values, the regression coefficient for each standardized parameter β_i in each model is weighted by the AIC weight $W(i)$ for that model, then summed over all models $\beta_{MA} = \sum W(i)\beta_i$. The models used to calculate these coefficients was the “all environmental variables” models, which included all possible combinations of one geographic trend and four environmental variables for a total of 32 models (Appendix 5). The coefficients can be positive or negative, depending on the slope of the relationship between species abundance and the environmental variable. The mean is derived from the absolute value of the coefficient.

Environmental Variable	Mean	No. of species with coefficients > 0.10	
		Positive slope	Negative slope
Main fertility gradient	0.141	34 (25%)	20 (15%)
Slope	0.069	23 (17%)	9 (7%)
pH	0.112	32 (24%)	24 (18%)
C:N ratio	0.059	12 (9%)	11 (8%)

FIGURE CAPTIONS

Figure 1. Floristic similarity (measured with Sørensen's index) as a function of geographic distance between plots in central Amazonia and three other Neotropical plot networks. Data for Panama, Peru, and Ecuador were adapted from Condit *et al.* (2002). The data points are for central Amazonia only. The lines were generated by classifying the floristic similarity data based on clusters of interplot distances and taking the average in each distance class.

Figure 2. Floristic similarity (measured with Steinhaus' index) as a function of difference in percent clay (an indicator of the main fertility gradient) between 40 plots in central Amazonia. Inset is a plot of floristic similarity versus difference in C:N ratio between plots. The lines are based on a linear regression.

Figure 3. Comparison of the percentage of individual species with significant relationships with soil and topographic variables versus the total number of species for 14 studies in tropical forests. Squares indicate studies where individual soil or topographic variables were used. Circles indicate studies where plots were divided into two or more distinct habitats. "X" indicates that the species in the study were only drawn from one or two particular taxa, such as only Dipterocarpaceae, or only palms. Information on the studies can be found in Appendix 7.

Fig. 1

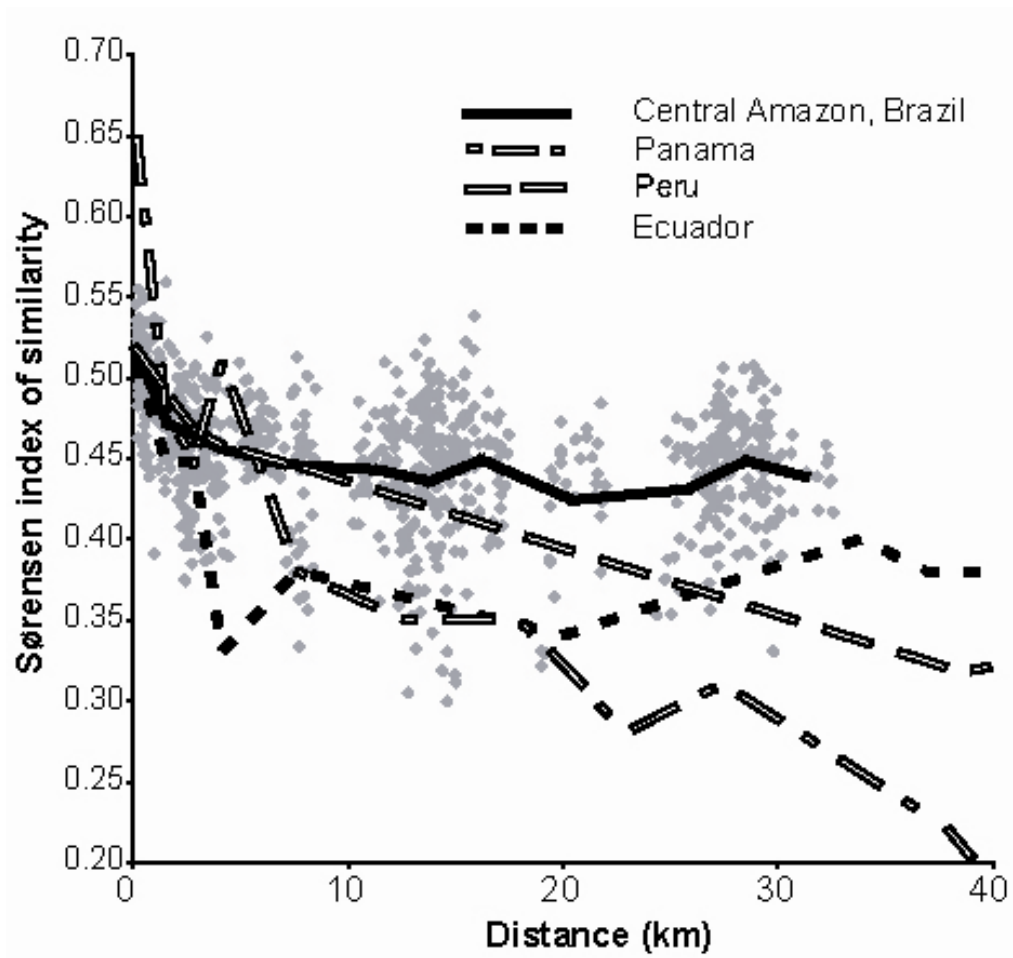


Fig. 2

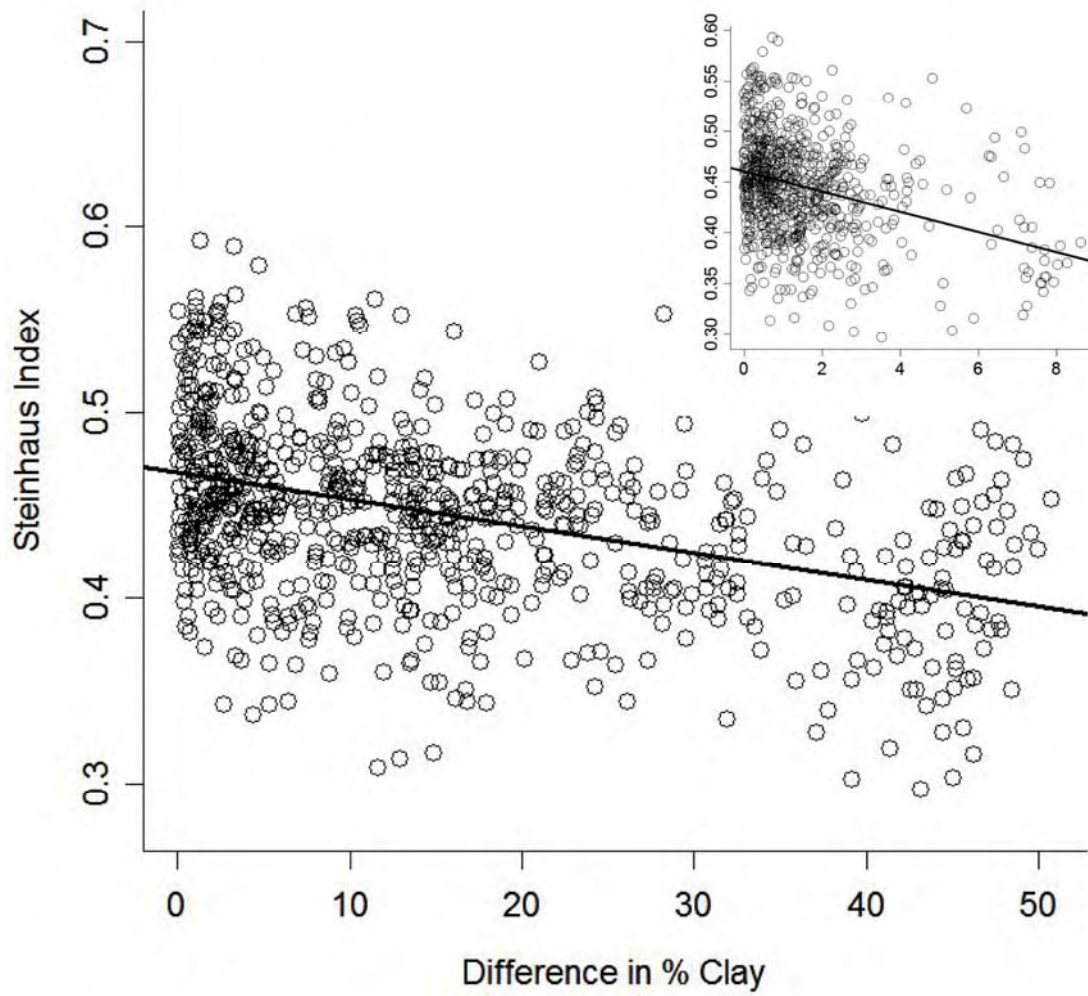
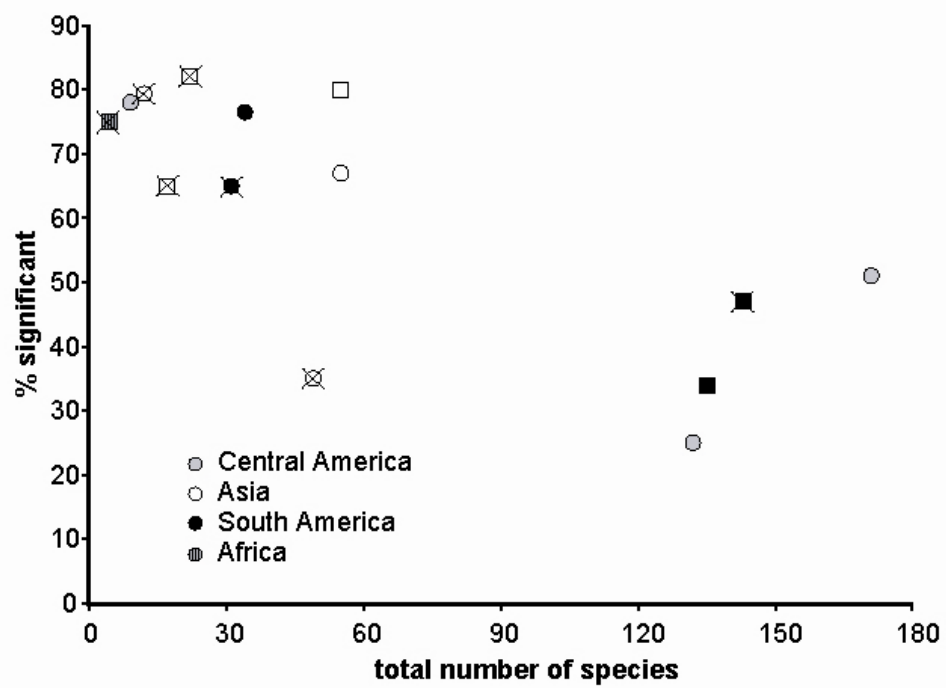


Fig. 3



Appendix 1. Laboratory methods for soil analysis.

Composite samples for each quadrat were oven-dried for 24 hours at 105°C, cleaned by removing stones and charcoal fragments, and then passed through 20 mm and 2 mm sieves. In all cases, values for soil parameters were derived separately for each quadrat, then combined to yield a mean value for each 1-ha plot.

Textural analyses were conducted to determine percentages of clay (particles <0.002 mm diameter), silt (0.002-0.05 mm), fine sand (0.05-0.2 mm), and coarse sand (0.2-2 mm), using the pipette method. Available water capacity (equivalent to "maximum plant-available water"), a measure of the amount of water the soil can hold in a form extractable by plant roots, was estimated as the difference between the field capacity (moisture content retained in soil under a suction of 0.33 atmospheres) and the wilting point (moisture content retained at 15 atmospheres), using a pressure membrane apparatus. As is standard practice, samples were dried, sieved, and re-wetted before determining available water capacity, making the results only an index of water available to plants in the field.

A pH meter was used to measure soil pH in water with a ratio of 1 ml soil to 25 ml distilled water (SCNLS-EMBRAPA 1979). Total N was determined by Kjeldahl digestion (Parkinson & Allen 1975). Total C was determined by dry combustion and organic C by the Walkley-Black method. Total P was determined by digestion in HNO_3^+ , HClO_4 , and HF (Lim & Jackson 1982), and reaction with ammonium molybdate. Extractable nutrients were determined as follows, using separate sub-samples for each group of analyses: PO_4^{3-} was measured in an autoanalyzer using the

molybdenum blue method (Jorgenson 1977); K^+ was determined by atomic emission spectroscopy at the Brazilian Center for Nuclear Energy in Agriculture (CENA), Piracicaba, São Paulo, while Ca^{2+} , Mg^{2+} , Na^+ , S, Cu, Fe, Mn^{2+} , Zn^+ , Al^{3+} , and H^+ in soil were measured by atomic absorption spectrophotometry at CENA after digestion in $HClO_4$, HNO_3 , and H_2SO_4 (Jorgenson 1977).

Cation measures were derived without Na^+ , which is generally a minor constituent of exchangeable bases and was recorded at only a limited number of study sites. These measures included cation exchange capacity (CEC), which is the sum of K^+ , Ca^{2+} , Mg^{2+} , Al^{3+} , and H^+ ; total exchangeable bases (TEB), the sum of K^+ , Ca^{2+} , Mg^{2+} , and Na^+ ; and aluminum saturation, which is $((Al^{3+} + H^+)/CEC) \times 100$. Samples were digested in $HClO_4$, HNO_3 , and H_2SO_4 , and extracts for cation determination were buffered to pH 7.0, the standard practice in Brazil (SNLCS-EMBRAPA 1979). A ratio of 1 ml soil to 10 ml solution was used, and the mixture was allowed to decant overnight without filtration (SNLCS-EMBRAPA 1979).

Organic carbon to total nitrogen (C:N) ratios were calculated to provide an index of N mineralization. Low C:N values indicate recalcitrant N, values in the 10-12 range are considered ideal for plant growth, while high values indicate inhibited nitrification, often due to waterlogging (Young 1976).

References

- Jorgenson, S. S. 1977. *Guia Analitico: Metodologia Utilizada para Analises Quimicas de Rotina*. Centro de Energia Nuclear na Agricultura (CENA), Piracicaba, Sao Paulo, Brazil.
- Lim, C. H. & Jackson, M. L. 1982. Dissolution for total elemental analysis. In: Page, A. L., Miller, R. H. & Keeney, D. R., (eds.) *Methods of Soil Analysis, Part 2: Chemical and Microbiological Properties*. pp. 1-12. American Society of Agronomy and Soil Science Society of America., Madison, WI.
- Parkinson, J. A. & Allen, S. E. 1975. A wet oxidation procedure suitable for the determination of nitrogen and mineral nutrients in biological material. *Communications in Soil Science and Plant Analysis* 6:1-11.
- SCNLS-EMBRAPA. 1979. *Manual de Metodos de Analise de Solo*. Servicio Nacional de Levantamento e Conservacao de Solos-Empresa Brasileira de Pesquisa Agropecuaria. SNCLS-EMBRAPA, Rio de Janeiro, Brazil.
- Young, A. 1976. *Soils and Soil Survey*. Cambridge University Press, Cambridge.

Appendix 2. Pearson correlations between the first two PCA axes and different soil and topographic variables in central Amazonia. Soil and topographic variables are listed in decreasing strength of association with PC1, and variables in boldface were used to assess the distributions of individual tree species.

Soil Variable	Units	Transfor- mation	Min	Max	Mean	Pearson	
						PC1	PC2
Clay Fraction	%	x²	18.0	73.3	53.6	0.97	-0.04
Fine Sand Fraction	%	log x	1.3	18.1	5.7	-0.95	0.02
Total Sand Fraction	%	log x	5.8	71.4	25.7	-0.95	0.09
Coarse Sand Fraction	%	log x	4.4	56.6	20.0	-0.94	0.12
Total Nitrogen	%	x ²	0.08	0.22	0.16	0.94	-0.15
Total Exchangeable Bases	m.e./100g dry soil	log x	0.064	0.303	0.192	0.91	0.24
Magnesium	m.e./100g dry soil	x	0.013	0.125	0.073	0.90	0.10
Zinc	ppm	x ²	0.63	2.99	1.65	-0.89	-0.19
Aluminum Saturation	% of CEC	-x ⁻¹	87.8	96.2	92.5	-0.79	0.00
Total Organic C	%	x ²	1.0	2.0	1.6	0.78	0.15
Sodium	m.e./100g dry soil	x ^{0.5}	0.017	0.106	0.051	0.77	-0.14
Manganese	ppm	x ²	0.87	2.49	1.73	0.76	-0.11
H	m.e./100g dry soil	x ²	0.393	0.841	0.681	0.73	-0.16
Calcium	m.e./100g dry soil	x ^{0.5}	0.015	0.131	0.063	0.71	0.57
Cation Exchange Capacity (CEC)	m.e./100g dry soil	x ³	1.07	3.31	2.44	0.69	0.57
Silt Fraction	%	x	8.3	32.8	20.6	0.66	-0.21
Iron	ppm	x ²	77	185	128	0.66	0.00
Potassium		x ^{0.5}	0.028	0.077	0.057	0.65	-0.51
Slope	%	x^{0.5}	0.3	37.6	12.3	-0.51	-0.08
Aluminum	m.e./100g dry soil	x ^{0.5}	0.42	2.22	1.57	0.51	0.75
Total Phosphorus		x	50.9	177.2	114.0	0.48	-0.81
Sulfur	ppm	x ²	11.3	14.9	12.9	0.46	0.33
Copper	ppm	x	0.10	0.54	0.30	0.41	0.05
Available Water Capacity	% by weight	x^{0.5}	3.3	11.3	7.2	0.39	-0.38
C:N Ratio	dimensionless	-x⁻³	8.3	16.6	10.0	-0.20	0.47
Phosphate	m.e./100g dry soil	x²	0.014	0.041	0.028	-0.05	-0.24
pH	pH units	x	3.7	4.5	4.1	0.05	-0.92

Appendix 3. Results of simple and partial Mantel tests between floristic similarity, interplot differences in environmental variables and interplot geographic distance in central Amazonia. The environmental variables and distance have been transformed. Floristic similarity was measured as either 1-Sørensen's index or Steinhaus' index. If the distance or environmental variable was available for more than 40 plots, the results for all plots containing that variable are listed in addition to the results for 40 plots. The results for Steinhaus' index and 40 plots are in the main text. The Mantel correlation coefficient (r) and significance level (p) is shown for each test. Relationships in bold are statistically significant at a Bonferroni-adjusted level of $p < 0.007$.

Floristic									
Floristic					Floristic				
Similarity		Transfor-	No. of	Variable ~		Similarity ~		Similarity ~	
Index	Independent Variable	mation	plots	Distance		Variable		Variable	
				<i>r</i>	<i>p</i>	<i>r</i>	<i>p</i>	<i>r</i>	<i>p</i>
Sørensen's	Geographic distance	log10(x)	66	--	--	0.53	0.0010	--	--
			40	--	--	0.57	0.0005	--	--
	Main fertility gradient	x ²	66	0.12	0.0035	0.30	0.0005	0.28	0.0005
			40	0.17	0.0020	0.42	0.0005	0.40	0.0005
	pH	x	40	0.41	0.0005	0.33	0.0005	0.13	0.0305
	Slope	x ^{0.5}	66	0.11	0.0025	0.24	0.0010	0.22	0.0005
			40	0.06	0.0645	0.15	0.0150	0.14	0.0305
	C:N ratio	-x ⁻³	40	0.19	0.0005	0.31	0.0005	0.25	0.0005
	Water	x ^{0.5}	40	0.06	0.0610	0.11	0.0500	0.09	0.0715
	Phosphate	x ²	40	0.09	0.0265	0.04	0.2810	0.00	0.5790
PC axis 1	x	40	0.18	0.0020	0.42	0.0005	0.39	0.0005	
Steinhaus'	Geographic distance	log10(x)	65	--	--	0.55	0.0005	--	--
	Main fertility gradient	x ²	65	0.12	0.0020	0.29	0.0005	0.26	0.0005
	Slope	x ^{0.5}	66	0.11	0.0030	0.23	0.0010	0.20	0.0005
	PC axis 1	x	40	0.18	0.0010	0.42	0.0005	0.39	0.0005

Appendix 4. Standardized regression coefficients from multiple regression analyses with interplot floristic similarity as the response variable and interplot geographic distance and interplot variation in six environmental variables as the explanatory variables. Floristic similarity is calculated as both 1-Sørensen's index and Steinhaus' index. Results for Steinhaus' index for 40 plots is shown in table 2 in the text. Only values for the main fertility gradient and slope were available for all 66 plots. A Bonferroni-corrected p-value of 0.005 was used in forward and backward selection to derive the final model from the full model.

	Sørensen's	Sørensen's	Sørensen's	Steinhaus'
	40 plots	40 plots	66 plots	66 plots
	Full model	Final model	Final model	Final model
	Coefficient (p)	Coefficient (p)	Coefficient (p)	Coefficient (p)
Full model				
Distance	0.42 (<0.0001)	0.49 (0.0010)	0.49 (<0.001)	0.53 (<0.001)
Main fertility gradient	0.31 (<0.0001)	0.30 (0.0010)	0.21 (<0.001)	0.22 (<0.001)
pH	0.15 (0.0030)			
C:N ratio	0.15 (0.0030)	0.15 (0.0020)		
Water	0.02 (0.3450)			
Slope	0.06 (0.1280)		0.14 (0.001)	
Phosphate	0.03 (0.0254)			
r^2	0.48 (0.0001)	0.45 (0.0005)	0.36 (0.0005)	0.35 (0.0005)

Appendix 5. Results of ordinary least squares (OLS) and conditional autoregressive (CAR) models for individual species abundances. The "trend/env/rho" consists of 8 models that are all possible combinations of geographic trend, environmental variables (main fertility gradient, slope, pH and C:N ratio together) and a spatial autocorrelation parameter (ρ). The "all env variables" consists of 32 models that are all possible combinations of geographic trend and the four environmental variables (main fertility gradient, slope, pH and C:N ratio individually). The r^2 values were averaged across any model containing the variable. The last four columns are the model-averaged standardized regression parameters for the soil and topographic variables in the "all environmental variables" models. Clay indicates the main soil fertility gradient. *** indicates that this soil variable had a significant Bonferroni-corrected p-value in a model with strong support that included this variable.

Species	Family	Best "Trend/Env/ Rho" Model	r ²	Best "All Env Variables" Model	r ²	Main Fertility Gradient	Slope	pH	C:N Ratio
<i>Bocageopsis multiflora</i> (Mart.) R. E. Fries	Annonaceae	trend+ρ	0.19	C:N ratio	0.13	-0.012	-0.033	0.010	-0.161
<i>Duguetia chrysea</i> Maas	Annonaceae	trend	0.21	slope+pH	0.27	-0.009	0.234	-0.110	0.026
<i>Duguetia echinophora</i> R. E. Fries	Annonaceae	env	0.33	clay+pH	0.29	0.229***	-0.009	0.094	-0.083
<i>Duguetia manausensis</i> Maas & Boon	Annonaceae	env	0.24	clay	0.15	0.187	-0.120	0.085	0.009
<i>Fusaea longifolia</i> (Aubl.) Saff.	Annonaceae	env	0.27	trend+clay+pH	0.35	0.318***	-0.030	0.111	-0.123
<i>Unonopsis duckei</i> R. E. Fries	Annonaceae	trend	0.16	pH	0.17	0.128	-0.030	0.306	0.027
<i>Geissospermum argenteum</i> Woods.	Apocynaceae	env	0.42	clay+pH	0.38	0.522***	-0.077	0.228	0.031
<i>Euterpe precatoria</i> Mart.	Arecaceae	trend	0.22	trend	0.22	-0.010	0.060	0.017	-0.048
<i>Oenocarpus bacaba</i> Mart.	Arecaceae	null	NA	C:N ratio	0.08	-0.002	0.008	0.031	-0.120
<i>Jacaranda copaia</i> (Aubl.) D. Don	Bignoniaceae	ρ	0.21	clay+pH	0.26	0.098	0.016	0.267***	-0.003

<i>Scleronema micranthum</i> (Ducke) Ducke	Bombacaceae	null	NA	slope	0.12	-0.022	0.159	-0.054	-0.012
<i>Protium altsonii</i> Sandw.	Burseraceae	trend	0.11	trend+clay+pH	0.28	0.235	-0.019	0.177	-0.042
<i>Protium apiculatum</i> Swart	Burseraceae	trend	0.18	trend	0.18	-0.108	-0.001	-0.059	-0.022
<i>Protium decandrum</i> (Aubl.) March.	Burseraceae	trend	0.17	clay+pH	0.22	0.201	0.006	0.078	-0.098
<i>Protium guianense</i> (Aubl.) March.	Burseraceae	ρ	0.20	trend+clay	0.35	0.289	0.036	-0.062	0.003
<i>Protium hebetatum</i> Daly	Burseraceae	trend	0.32	trend+clay	0.37	-0.249	0.067	-0.082	-0.014
<i>Protium occultum</i> Daly	Burseraceae	env	0.30	trend+clay+pH	0.32	0.260***	0.052	0.228	0.035
<i>Protium paniculatum</i> Engl.	Burseraceae	trend	0.32	trend+clay	0.36	-0.270***	0.031	-0.106	0.060
<i>Protium strumosum</i> Daly	Burseraceae	trend	0.19	pH	0.17	0.076	-0.026	-0.158	0.005
<i>Protium trifoliolatum</i> Engl.	Burseraceae	trend	0.23	trend	0.23	0.035	-0.019	-0.079	0.045
<i>Tetragastris panamensis</i> (Engl.) Kuntze	Burseraceae	env+p	0.47	clay	0.37	-0.349***	0.070	0.141	0.030
<i>Pourouma bicolor</i> Mart.	Cecropiaceae	null	NA	null	NA	0.003	0.026	0.031	-0.015

<i>Pourouma guianensis</i> Aubl. ssp. guianensis	Cecropiaceae	env	0.36	clay+slope+pH	0.36	-0.342***	0.144	0.112	-0.045
<i>Pourouma tomentosa</i> Miq.	Cecropiaceae	null	NA	clay	0.07	-0.197	-0.002	-0.010	-0.124
<i>Goupia glabra</i> Aubl.	Celastraceae	null	NA	null	NA	0.009	-0.023	0.022	-0.044
<i>Couepia caryophylloides</i> Benoist	Chryso- balanaceae	env	0.32	clay+slope	0.31	-0.433***	-0.241	-0.010	0.015
<i>Couepia guianensis</i> Aubl. (Miq.) Prance	Chryso- balanaceae	env+p	0.44	clay+pH	0.33	-0.356***	0.031	-0.152	0.062
<i>Couepia longipendula</i> Pilger	Chryso- balanaceae	trend	0.49	trend+C:N ratio	0.56	-0.176	-0.072	0.001	0.196
<i>Hirtella bicornis</i> Mart. & Zucc. var. pubescens Ducke	Chryso- balanaceae	env	0.38	pH	0.36	0.015	-0.028	0.431***	0.027
<i>Hirtella rodriguesii</i> Prance	Chryso- balanaceae	trend	0.28	clay+pH+C:N ratio	0.37	0.166***	-0.053	0.243***	0.116
<i>Licania canescens</i> Benoist	Chryso- balanaceae	null	NA	clay	0.10	0.129	-0.097	-0.003	-0.044
<i>Licania caudata</i> Prance	Chryso- balanaceae	null	NA	null	NA	0.092	0.007	0.119	0.009
<i>Licania heteromorpha</i> Benth.	Chryso- balanaceae	trend	0.33	trend+slope	0.38	-0.033	0.156	-0.089	0.004

<i>Licania impressa</i> Prance	Chryso- balanaceae	trend+env	0.58	trend+clay+slo pe+pH	0.57	0.700***	0.245	0.210	0.019
<i>Licania micrantha</i> Miq.	Chryso- balanaceae	trend	0.35	trend+slope+p H	0.47	-0.074	0.134	0.194	0.016
<i>Licania oblongifolia</i> Standl.	Chryso- balanaceae	env	0.28	slope+C:N ratio	0.23	0.051	0.353***	-0.090	0.191
<i>Buchenavia grandis</i> Ducke	Combretaceae	null	NA	slope	0.08	-0.032	-0.137	0.036	-0.018
<i>Sloanea floribunda</i> Spruce ex Benth.	Elaeocarpaceae	null	NA	null	NA	0.014	0.017	-0.048	-0.028
<i>Conceveiba hostmannii</i> Benth.	Euphorbiaceae	trend+ρ	0.62	trend	0.59	0.002	0.003	0.033	0.006
<i>Drypetes variabilis</i> Uittien	Euphorbiaceae	trend+env	0.36	trend+clay+pH	0.34	0.314***	0.017	0.272***	-0.069
<i>Hevea guianensis</i> Aubl.	Euphorbiaceae	null	NA	null	NA	-0.002	-0.006	0.111	-0.079
<i>Mabea caudata</i> Pax & K. Hoffm.	Euphorbiaceae	env	0.28	C:N ratio	0.18	-0.061	0.114	0.180	-0.389***
<i>Endopleura uchi</i> (Huber) Cuatr.	Humiriaceae	null	NA	null	NA	0.062	-0.015	-0.024	0.009
<i>Sacoglottis</i> <i>mattogrossensis</i> Malme var. <i>subintegra</i> (Ducke)	Humiriaceae	ρ	0.25	clay+slope	0.20	0.294***	0.107	0.052	0.006

<i>Vantanea parviflora</i> Lam.	Humiriaceae	env	0.28	clay+C:N ratio	0.28	0.078	-0.025	-0.025	0.321***
<i>Lacistema aggregatum</i> (Berg.) Rusby	Lacistemataceae e	null	NA	null	NA	-0.008	-0.029	0.007	-0.031
<i>Aniba burchellii/panurensis</i> s.l.	Lauraceae	null	NA	slope	0.10	-0.013	0.170	0.006	0.007
<i>Licaria cannella</i> (Meissn.) Kosterm.	Lauraceae	null	NA	pH	0.13	0.055	-0.021	-0.232	0.002
<i>Mezilaurus</i> sp. nov. 1	Lauraceae	null	NA	pH	0.06	0.079	-0.059	0.189	-0.007
<i>Ocotea ceanothifolia</i> (Nees) Mez	Lauraceae	env	0.32	slope+C:N ratio	0.32	0.021	0.362***	0.016	-0.122
<i>Ocotea cinerea</i> van der Werff	Lauraceae	trend	0.12	trend+C:N ratio	0.21	-0.003	-0.005	-0.079	0.119
<i>Ocotea cujumari</i> Mart.	Lauraceae	null	NA	clay+pH	0.21	0.228	-0.015	-0.101	0.001
<i>Ocotea percurrens</i> Vicentini	Lauraceae	ρ	0.26	trend+clay	0.33	0.122	0.005	-0.007	-0.099
<i>Corythophora alta</i> R. Knuth	Lecythidaceae	trend+ ρ	0.63	trend+clay	0.61	0.132	-0.050	-0.011	-0.094
<i>Couratari stellata</i> A. C. Smith	Lecythidaceae	trend+ ρ	0.53	trend+slope	0.49	-0.017	-0.113	-0.043	-0.020
<i>Eschweilera atropetiolata</i> Mori	Lecythidaceae	trend	0.65	trend+clay	0.68	0.104	-0.054	-0.016	0.007

<i>Eschweilera collina</i> Eyma	Lecythidaceae	trend+env	0.46	trend+clay+slope+C:N ratio	0.45	-0.466***	-0.253	-0.021	-0.353
<i>Eschweilera coriacea</i> (A. DC.) Mori	Lecythidaceae	env	0.59	clay	0.58	-1.246***	-0.015	0.017	-0.037
<i>Eschweilera cyathiformis</i> Mori	Lecythidaceae	trend	0.23	trend	0.23	0.154	0.050	0.066	0.009
<i>Eschweilera grandiflora</i> (Aubl.) Sandw.	Lecythidaceae	null	NA	clay	0.10	0.112	-0.069	-0.090	-0.002
<i>Eschweilera micrantha</i> (Berg) Miers	Lecythidaceae	trend+env+p	0.67	trend+clay+pH+C:N ratio	0.63	0.466***	0.013	-0.454	0.234
<i>Eschweilera pseudodecolorans</i> Mori	Lecythidaceae	env+p	0.51	trend+clay+pH	0.45	0.634***	-0.033	0.270	-0.075
<i>Eschweilera rankiniae</i> Mori	Lecythidaceae	trend	0.40	trend+slope	0.49	-0.020	-0.172	-0.117	0.042
<i>Eschweilera romeu-cardosoi</i> Mori	Lecythidaceae	trend+env	0.56	trend+clay	0.56	0.830***	-0.015	-0.042	0.007
<i>Eschweilera truncata</i> A. C. Smith	Lecythidaceae	trend	0.64	trend+pH	0.71	-0.003	0.006	-0.478	0.014
<i>Eschweilera wachenheimii</i> (Benoist) Sandw.	Lecythidaceae	null	NA	null	NA	0.025	-0.053	0.024	-0.020
<i>Gustavia elliptica</i> Mori	Lecythidaceae	trend+p	0.33	pH	0.26	0.171	0.142	-0.291***	0.043

<i>Lecythis barnebyi</i> Mori	Lecythidaceae	env	0.30	pH+C:Nratio	0.27	0.071	0.037	-0.355***	-0.235
<i>Lecythis poiteaui</i> Berg	Lecythidaceae	ρ	0.07	trend+clay+C: N ratio	0.28	-0.195	-0.014	-0.020	-0.176
<i>Bocoa viridiflora</i> (Ducke) Cowan	Leguminosae	env+ ρ	0.53	trend+slope+p H	0.53	0.006	0.196***	-0.280***	-0.040
<i>Eperua glabriflora</i> (Ducke) Cowan	Leguminosae	ρ	0.16	trend	0.20	-0.004	0.006	0.102	0.069
<i>Inga paraensis</i> Ducke	Leguminosae	null	NA	pH	0.08	-0.023	-0.024	-0.100	-0.016
<i>Macrolobium limbatum</i> Benth.	Leguminosae	env	0.30	slope+pH	0.29	0.008	0.259***	0.141	-0.016
<i>Paramachaerium</i> <i>ormosioides</i> (Ducke) Ducke	Leguminosae	trend	0.34	trend+clay	0.44	0.401***	-0.004	0.170***	-0.016
<i>Sclerolobium</i> sp. 2	Leguminosae	trend	0.18	trend	0.18	-0.060	-0.006	-0.013	0.012
<i>Swartzia ingifolia</i> Ducke	Leguminosae	trend	0.12	pH	0.14	0.001	0.032	-0.204	-0.016
<i>Swartzia polyphylla</i> DC.	Leguminosae	null	NA	null	NA	0.020	0.011	0.014	0.050
<i>Swartzia recurva</i> Poepp.	Leguminosae	trend+env	0.40	trend+slope+p H	0.38	-0.022	0.157	-0.422***	0.014

<i>Swartzia reticulata</i> Ducke	Leguminosae	env	0.46	clay	0.42	0.674***	-0.082	0.048	-0.037
<i>Tachigali plumbea</i> Ducke	Leguminosae	null	NA	null	NA	-0.003	-0.028	0.024	-0.024
<i>Zygia racemosa</i> (Ducke) Barn. & Grimes	Leguminosae	null	NA	null	NA	0.019	0.016	-0.055	-0.010
<i>Mouriri angulicosta/duckeana/duckeanoides</i>	Melastomataceae	null	NA	slope	0.11	0.007	-0.171	0.098	-0.001
<i>Guarea humaitensis</i> Penn.	Meliaceae	trend+env	0.43	trend+clay+pH +C:N ratio	0.41	-0.311***	-0.092	-0.108	0.223***
<i>Siparuna decipiens</i> (Tul.) A. DC.	Monimiaceae	null	NA	pH	0.07	0.021	0.011	0.102	-0.004
<i>Brosimum rubescens</i> Taub.	Moraceae	trend	0.15	clay	0.16	-0.247***	0.008	-0.110	-0.048
<i>Helianthostylis sprucei</i> Baill.	Moraceae	trend+env	0.41	trend+clay+slope +C:N ratio	0.41	0.285***	0.080	0.006	0.420***
<i>Helicostylis tomentosa</i> (Planch. & Endl.) Rusby	Moraceae	trend	0.24	trend	0.24	-0.019	0.012	0.003	-0.043
<i>Maquira sclerophylla</i> (Ducke) C. C. Berg	Moraceae	env	0.28	clay+pH	0.25	0.275***	-0.062	0.116	-0.050
<i>Naucleopsis caloneura</i> (Huber) Ducke	Moraceae	trend	0.11	pH	0.06	-0.023	0.027	-0.084	0.064

<i>Pseudolmedia laevis</i> (Ruiz & Pav.) Macbr.	Moraceae	env+p	0.33	clay	0.08	-0.111	0.013	0.011	0.004
<i>Trymatococcus amazonicus</i> Poepp. & Endl.	Moraceae	trend	0.12	clay	0.12	-0.148	0.024	-0.033	0.068
<i>Iryanthera dialyandra</i> Warb.	Myristacaceae	null	NA	null	NA	0.032	0.019	0.009	-0.042
<i>Iryanthera juruensis</i> Warb.	Myristacaceae	env	0.26	clay+slope	0.25	-0.127	0.228***	0.013	0.023
<i>Virola calophylla</i> Warb.	Myristacaceae	null	NA	slope	0.13	0.004	0.212	-0.024	0.054
<i>Virola sebifera</i> Aubl.	Myristacaceae	env+p	0.38	C:N ratio	0.22	-0.013	0.022	-0.069	-0.373***
<i>Virola venosa</i> (Benth.) Warb.	Myristacaceae	null	NA	null	NA	0.006	-0.045	0.018	0.017
<i>Neea</i> sp. 2	Nyctaginaceae	null	NA	null	NA	-0.003	-0.024	-0.027	0.024
<i>Heisteria laxiflora</i> Engl.	Olacaceae	trend	0.11	trend+pH	0.24	-0.042	0.050	-0.308	0.047
<i>Minuartia guianensis</i> Aubl.	Olacaceae	null	NA	C:N ratio	0.06	0.066	0.040	0.072	-0.075
<i>Chrysophyllum sanguinolentum</i> (Pierre) Baehni	Sapotaceae	env	0.47	clay+slope+pH	0.47	-0.452***	0.177	-0.109	-0.006

<i>Ecclinusa guianensis</i> Eyma	Sapotaceae	null	NA	trend+clay+pH	0.23	0.131	-0.009	0.110	0.013
<i>Manilkara bidentata</i> (A. DC.) Chev.	Sapotaceae	null	NA	null	NA	0.005	0.011	0.016	-0.008
<i>Manilkara huberi</i> (Ducke) Chev.	Sapotaceae	trend+env	0.59	trend+clay+slo pe+pH	0.57	0.445***	-0.119	0.373***	-0.050
<i>Micropholis</i> <i>casiquiarensis/mensalis</i>	Sapotaceae	null	NA	clay	0.08	-0.094	0.012	0.017	-0.018
<i>Micropholis guyanensis</i> (A. DC.) Pierre	Sapotaceae	env	0.29	clay	0.26	-0.445***	-0.002	-0.024	0.040
<i>Micropholis venulosa</i> (Mart. & Eichl.) Pierre	Sapotaceae	null	NA	null	NA	-0.051	-0.003	0.059	0.021
<i>Micropholis williamii</i> Aubr. & Pellegr.	Sapotaceae	trend	0.16	trend	0.16	0.001	0.032	-0.071	0.053
<i>Pouteria ambelaniifolia</i> (Sandw.) Penn.	Sapotaceae	ρ	0.11	pH	0.06	-0.010	-0.003	-0.063	0.006
<i>Pouteria anomala</i> (Pires) Penn.	Sapotaceae	env	0.54	slope+pH	0.52	-0.075	0.470***	-0.820***	0.075
<i>Pouteria caimito</i> (R. & P.) Radlk.	Sapotaceae	trend	0.19	trend	0.19	-0.036	0.035	0.015	0.005
<i>Pouteria campanulata</i> Baehni	Sapotaceae	null	NA	pH	0.06	-0.052	0.013	-0.065	0.007
<i>Pouteria cladantha</i> Sandw.	Sapotaceae	trend	0.20	clay+pH	0.22	0.110	-0.067	0.098	-0.048

<i>Pouteria cuspidata</i> (A. DC.) Baehni	Sapotaceae	null	NA	slope	0.08	-0.023	0.137	0.057	0.061
<i>Pouteria engleri</i> Eyma	Sapotaceae	trend	0.14	slope+pH	0.21	0.071	-0.137	0.210	0.040
<i>Pouteria ericoides</i> Penn.	Sapotaceae	null	NA	null	NA	0.068	0.006	0.020	-0.005
<i>Pouteria erythrochrysa</i> Penn.	Sapotaceae	trend	0.20	trend	0.20	0.046	0.005	-0.026	0.006
<i>Pouteria eugeniifolia</i> (Pierre) Baehni	Sapotaceae	trend	0.18	trend+slope+pH	0.32	0.014	0.133	0.140	0.009
<i>Pouteria filipes</i> Eyma	Sapotaceae	null	NA	pH	0.09	-0.019	0.033	-0.131	0.027
<i>Pouteria freitasii</i> Penn.	Sapotaceae	trend	0.20	trend+pH	0.26	0.027	0.023	0.137	0.019
<i>Pouteria guianensis</i> Aubl.	Sapotaceae	trend+env	0.39	trend+slope+pH	0.36	0.007	0.229***	0.294***	-0.032
<i>Pouteria hispida</i> Eyma	Sapotaceae	null	NA	slope	0.05	0.016	-0.066	-0.001	0.027
<i>Pouteria macrophylla</i> (Lam.) Eyma	Sapotaceae	env+p	0.49	clay+pH	0.39	0.353***	-0.079	0.139	0.011
<i>Pouteria minima</i> Penn.	Sapotaceae	trend	0.23	clay+pH	0.28	-0.143	0.006	-0.136	0.109
<i>Pouteria reticulata</i> (Engl.) Eyma	Sapotaceae	trend	0.29	pH+C:Nratio	0.34	-0.019	0.004	-0.202***	0.107

<i>Pouteria retinervis</i> Penn.	Sapotaceae	trend	0.12	C:N ratio	0.13	0.008	-0.009	-0.003	0.184
<i>Pouteria rostrata</i> (Huber) Baehni	Sapotaceae	env	0.30	pH	0.29	-0.013	0.031	-0.410***	0.005
<i>Pouteria venosa</i> (Mart.) Baehni	Sapotaceae	null	NA	null	NA	-0.013	-0.032	0.043	-0.041
<i>Pouteria vernicosa</i> Penn.	Sapotaceae	trend	0.24	trend	0.24	-0.036	0.054	-0.080	0.118
<i>Sterculia parviflora</i> (Ducke) Taylor	Sterculiaceae	trend	0.12	trend+C:N ratio	0.20	-0.017	0.007	0.042	-0.121
<i>Theobroma sylvestre</i> Mart.	Sterculiaceae	trend	0.17	trend	0.17	0.040	0.003	-0.039	0.011
<i>Amphirrhox surinamensis</i> Eichl.	Violaceae	null	NA	pH	0.06	0.076	-0.009	0.137	0.020
<i>Rinorea guianensis</i> Aubl. var. <i>subintegrifolia</i>	Violaceae	env	0.26	clay+slope	0.25	0.443***	0.203	0.037	-0.017
<i>Rinorea racemosa</i> (Mart.) Kuntze	Violaceae	null	NA	slope	0.06	0.010	-0.087	-0.023	-0.014
<i>Erisma bicolor</i> Ducke	Vochysiaceae	trend	0.13	trend+slope	0.20	0.057	-0.093	-0.007	0.036
<i>Qualea labouriauna</i> Paule	Vochysiaceae	env	0.30	clay+pH	0.28	0.198	-0.048	0.212	-0.043

Appendix 6. Results from ordinary least squares (OLS) models with all possible combinations of one geographic trend variable and four environmental variables (main fertility gradient, slope, pH and C:N ratio individually) as explanatory variables. Because PO_4 and available water were the least important variables in the community level analysis, they were not included in the species level analyses. The last three rows contain the sum of AIC_c weights and the sum of species with the best or strongly-supported model for any model containing the variable listed.

Model	AIC _c weights	Best model # species	Best model or strong support # species
Null	5.1%	18 (13%)	58 (43%)
pH	5.0%	15 (11%)	63 (47%)
Trend	5.3%	13 (10%)	82 (61%)
Main fertility gradient	5.4%	12 (9%)	75 (56%)
Main fertility gradient+pH	5.7%	11 (8%)	76 (56%)
Slope	3.8%	8 (6%)	54 (40%)
Trend+main fertility gradient	4.5%	8 (6%)	67 (50%)
C:N ratio	3.2%	6 (4%)	50 (37%)
Trend+main fertility gradient+pH	3.9%	6 (4%)	46 (34%)
Trend+slope+pH	2.6%	5 (4%)	30 (22%)
Main fertility gradient+slope	3.5%	4 (3%)	57 (42%)
Slope+pH	4.0%	4 (3%)	64 (47%)
Trend+slope	3.4%	4 (3%)	50 (37%)
Trend+C:N ratio	3.1%	3 (2%)	54 (40%)
trend+pH	3.4%	3 (2%)	59 (44%)
Main fertility gradient+slope+pH	3.5%	2 (1%)	55 (41%)
pH+C:N ratio	2.8%	2 (1%)	45 (33%)
Slope+C:N ratio	2.6%	2 (1%)	47 (35%)
Trend+main fertility gradient+pH+C:N ratio	2.0%	2 (1%)	31 (23%)
Trend+main fertility gradient+slope+C:N ratio	1.8%	2 (1%)	16 (12%)
trend+main fertility gradient+slope+pH	2.3%	2 (1%)	34 (25%)
Main fertility gradient+C:N ratio	3.3%	1 (1%)	60 (44%)
Main fertility gradient+pH+C:N ratio	3.1%	1 (1%)	54 (40%)
Trend+main fertility gradient+C:N ratio	2.6%	1 (1%)	37 (27%)
Main fertility gradient+slope+C:N ratio	2.0%	0 (0%)	36 (27%)
Main fertility gradient+slope+pH+C:N ratio	1.7%	0 (0%)	27 (20%)
Slope+pH+C:N ratio	2.1%	0 (0%)	37 (27%)
Trend+main fertility gradient+slope	2.6%	0 (0%)	35 (26%)
Trend+main gradient+slope+pH+C:N ratio	1.2%	0 (0%)	10 (7%)
Trend+pH+C:N ratio	1.7%	0 (0%)	31 (23%)
Trend+slope+C:N ratio	1.7%	0 (0%)	30 (22%)
Trend+slope+pH+C:N ratio	1.2%	0 (0%)	15 (11%)

Appendix 7. Comparison of the percentage of individual species with significant relationships with soil and topographic variables from different tropical forests. For studies with “trees” listed as the species type, no subset of species was based on the abundance of species rather than on a specific taxon. Total area sampled was calculated by adding up the size of the individual plots. The statistical methods used for determining whether a species had a significant relationship with a soil variable or habitat varied among studies.

Authors	Year	Country	No. of spp.	% species with significant soil relationships	Species type	Soil variables or habitats	Study area size (ha)	No. of plots	Total area sampled (ha)
Hall et al.	2004	Central African Republic	4	75	<i>Entandrophragma</i>	chemical, physical, topographic variables	100	1	100
Sri-ngernyuang et al.	2003	Thailand	17	65	Lauraceae	topographic variables	7.5	1	7.5
Paoli et al.	2006	Borneo	22	82	Dipterocarpaceae	chemical, physical variables	340	30	4.8
Svenning	1999	Ecuador	31	65	palms	topographic-, hydrologic-, other-based habitats	50	118	4.72
Phillips et al.	2003	Peru	34	77	trees	pleistocene, holocene substrate habitats	1000000	88	8.8
Webb & Peart	2000	Borneo	49	35	trees	topographic-based habitats	150	28	4.48
Miyamoto et al.	2003	Borneo	55	80	trees	elevation, humus depth	200	2	2
Cannon & Leighton	2004	Borneo	55	67	trees	substrate, elevation, humus depth-based habitats	1500	69	6.9

1 Appendix 7 (con't)

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Authors	Year	Country	No. of spp.	% species with significant soil relationships	Species type	Soil variables or habitats	Study area size (ha)	No. of plots	Total area sampled (ha)
Clark et al.	1999	Costa Rica	132	25	trees	hydrologic-, substrate- and topographic-based habitats	573	110	1.1
Clark et al.	1998	Costa Rica	9	78	trees	soil type, topographic position, slope	216	NA	NA
Tuomisto et al.	2002	Ecuador	143	47	melostomes and pteridophytes	chemical variables	50000	27	6.75
Harms et al.	2001	Panama	171	51	trees	topographic-, hydrologic- habitats	50	1	50
Yamada et al.	2006	Borneo	10	80	trees	topographic-, soil texture- habitats	52	1	52
Bohlman et al.	this study	Brazil	135	30	trees	chemical, physical, topographic variables	100000	66	66

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