How many tropical forest tree species are there?

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137 The high species richness of tropical forests has long been rec-138 ognized, yet there remains substantial uncertainty regarding the 139 actual number of tropical tree species. Using a pan-tropical tree 140 inventory database from closed canopy forests, consisting of 657,630 trees belonging to 11,371 species, we use a fitted value 142 of Fisher's alpha and an approximate pan-tropical stem total to 143 estimate the minimum number of tropical forest tree species to 144 fall between ${\sim}40{,}000$ and ${\sim}53{,}000{,}$ i.e. at the high end of previ-145 ous estimates. Contrary to common assumption, the Indo-Pacific 146 region was found to be as species rich as the Neotropics, with both regions having a minimum of \sim 19,000-25,000 tree species. 148 Continental Africa is relatively depauperate with a minimum of 149 \sim 4,500-6,000 tree species. Very few species are shared among 150 the African, American and the Indo-Pacific regions. We provide a 151 methodological framework for estimating species richness in trees 152 that may help refine species richness estimates of tree dependent 153 taxa. 154

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Diversity estimation | Fishers's log-series | Pan-tropical | Spatial richness patterns | Tropical tree species richness

Significance statement People are fascinated by the amazing diversity of tropical forests and will be surprised to learn that robust estimates of the number of tropical tree species are lacking. We show that there are at least 40,000, but possibly more than 53,000 tree species in the tropics, in contrast to only 124 across temperate Europe. Almost all tropical tree species are restricted to their respective continents, while the Indo-Pacific region appears to be as species rich as tropical America, with each of these two regions being almost five times as rich in tree species as African tropical forests. Our study shows that most tree species are extremely rare, meaning that they are under serious risk of extinction at current deforestation rates.

Despite decades of biological inventories worldwide, we still do not know how many species exist and how they are distributed (1). Although global patterns of estimated vascular plant species richness and distribution have become more clear (2-5), no previous study has focused on trees as a distinct growth form. As a consequence, our estimation of the number of tree species in tropical forests still depends on untested expert opinions (6-8), rather than on an appropriate methodological framework and data set.

Given the importance of trees as key structural components of forest ecosystems, sources of timber and non-timber products, and providers of vital ecosystem services (9, 10), the lack of reliable estimates of the total number of tropical tree species represents a critical knowledge gap that has direct consequences for estimating the diversity of other tree dependent taxa (11). A classic example is Erwin's (6) estimate of the existence of 30 million arthropod species, which was based on observed host specificities of arthropods with individual tropical tree species combined with an estimate of the total number of tropical tree species. Global arthropod richness has subsequently been revised downward (7), but current estimates still suffer from the lack of information on the number of tropical tree species.

In recent decades, the number of tree inventory plots across the tropics has grown to such an extent that species richness estimation at the continental and pan-tropical scale can now be addressed using standardized species lists with abundance data. Prior estimates of plant richness at such broad scales have been based on analyses of incidence data obtained from herbarium collections and Flora treatments (2-5). However, these methods are highly sensitive to collecting biases and ignore valuable information on species' abundances (12). Abundance data enable extrapolation of richness from local to global scales using diversity estimators that fit the observed species rank-abundance data (13-15).

Results and Discussion

We estimate the number of tropical tree species from a standardized dataset of old-growth tropical forest tree inventories (including gymno- and angiosperms with diameters at breast height $(dbh) \ge 10$ cm). This dataset contains tree species abundance data for 207 one-degree grid cells (locations) originally dominated by closed canopy forests across the tropics (Fig. 1). By calculating Fisher's-alpha (16) at the pan-tropical scale and combining this value with the estimated potential number of stems present within 500 km of each location, we arrive at a minimum number of tropical tree species of at least \sim 40,000 and possibly more than \sim 53,000 (Table 1), i.e. at the high end of current total estimates of 37,000 (7), 43,000 (8) and 50,000 (6), which are based on expert opinion.

When the analysis was restricted to each of the three main tropical regions, we found that the Indo-Pacific had comparable tree species richness to that found in tropical America (Table 1). Moreover, these two regions show similar rates of species turnover for a given increase in geographical distance between locations (Fig. 2). This result contradicts the widely held view that the Neotropics are the most diverse and species-rich region for tropical trees (8, 15, 17, 18). This underestimation of Indo-Pacific tree species richness, and our inclusion of dry as well as moist and wet forests, may explain why some of the previous estimates (7, 8) are lower than ours. Nevertheless, the high species richness in the Indo-Pacific is understandable given the highly variable topography, complex geological history, steep environmental gradients, past and ongoing merging of several contrasting floras from Madagascar, India, Southeast Asia, and New Guinea-Australia (19, 20), as well as the large current and time integrated forest area (8).

Tropical continental Africa has a relatively depauperate tree flora, a finding consistent with earlier studies (21, 22). This region shows comparatively low species turnover, in other words, as sample area increases, the number of tree species increases at a much slower rate than in either the Indo-Pacific or the Neotropics (Fig. 2). The differences in species richness and spatial turnover, comparing continental Africa with the other tropical regions, cannot be explained solely by Africa's smaller forest area or lower environmental variability (Table 1; Fig. S2). Rather, these disparities support the hypothesis that African forests have

Significance

People are fascinated by the amazing diversity of tropical forests and will be surprised to learn that robust estimates of the number of tropical tree species are lacking. We show that there are at least 40,000, but possibly more than 53,000 tree species in the tropics, in contrast to only 124 across temperate Europe. Almost all tropical tree species are restricted to their respective continents, while the Indo-Pacific region appears to be as species rich as tropical America, with each of these two regions being almost five times as rich in tree species as African tropical forests. Our study shows that most tree species are extremely rare, meaning that they are under serious risk of extinction at current deforestation rates.

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Fig. 1. Overview of sample locations and their floristic affinities (point colors correspond to scores on the first DCA-axis with similar colours indicating similar generic composition, while the lines indicate the floristic affinities as determined by cluster analysis).

 Table 1. Species richness estimates for the tropics and the three main tropical sub-regions.

| Region Africa America | Species observed | Stems observed | Unidentified stems (%) | Fisher's-alpha minimum | Fisher's- alpha maximum | Stems estimated | Species minimum 4626 18589 | Species maximum 5984 24580 |
|------------------------------|---------------------|-------------------|---------------------------|---------------------------|----------------------------|--|-------------------------------------|-------------------------------------|
| | 1376 4375 | 117902 116754 | 8.4 13.5 | 218.7 897.2 | 286.6 1203.4 | 3.4x10 ¹¹ 8.9x10 ¹¹ | | |
| Indo-Pacific Pan-tropical | 5672 11371 | 422974 657630 | 9.6 10.1 | 925.8 1953.0 | 1225.2 2607.7 | 7.7x10 ¹¹ 2.0x10 ¹² | 19014 40517 | 24819 53345 |

Observed values represent the numbers in the original dataset. Unidentified stems were excluded from estimation of minimum Fisher's-alpha, but used to calculate maximum Fisher's-alpha. Estimated stems represent the number of stems predicted to occur within 500 km of each of the 207 sample locations.



Fig. 2. Increase in Fisher's-alpha with (a) increasing numbers of locations (average of 50 replicates per region with random input order of locations), i.e. regional diversity and (b) increasing distance around locations (based on 50 replicates per region each with a randomly selected starting location), i.e. species turnover. Error bars indicate standard deviation among location reorderings. Fisher's-alpha can decline if the number of stems added to the sample increase disproportionally to the number of new species detected.

experienced severe extinction events due to repeated shrinkage of forest area during the Pleistocene (19, 23). When these forests expanded to their present size, they could only be repopulated by a severely depleted species pool derived from a limited number of refugia. In contrast, tropical America retained considerable forest

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cover and equatorial forests of the Indo-Pacific may even have expanded during the same period (19, 20, 23, 24).

We provide the first survey-based minimum estimate of tropical tree species richness and its distribution. We acknowledge, however, that the current estimate is just a first step in an ongoing effort. Estimates of species richness will become more refined and increasingly accurate as forest surveys continue to expand. This study highlights the usefulness and critical importance of forest surveys, and we emphasize once more the existence of large numbers of tree species with exceptionally small population sizes, which may necessitate novel conservation approaches for effective preservation of current tree diversity (25, 26).

Materials and Methods

Data set. Tree inventory data (gymnosperms and angiosperms only; trunk diameter at breast height ≥ 10 cm) from old-growth forest plots without signs of recent human disturbance were compiled from across the tropics and subtropics (i.e., within 30° north and south of the Equator). Individual trees from the inventories were pooled within their respective one degree grid cells (henceforth called locations). Species names were standardized using 'The Plant List' <www.theplantlist.org>, 'Taxonomic Name Resolution Service' <http://tnrs.iplantcollaborative.org/TNRSapp.html> and 'The Asian Plant Synonym Lookup' http://phylodiversity.net/fslik/synonym_lookup.htm. Unknown taxa were not used in diversity and composition analyses. We removed all locations with fewer than 250 identified individuals to minimize effects of sample size, resulting in 207 retained locations (Dataset S1), each showing a reasonably high sample coverage (0.96 \pm 0.3, mean \pm SE) that did not differ among geographic areas (Table S1), thus suggesting that our estimations of species richness were not biased by differences in sample coverage among regions.

Phylogeographic analyses. To identify the main tropical regions for species richness comparison, we performed 'Minimum Variance Clustering' with 'Squared Euclidean Distances' on square-root transformed relative abundance data at the genus level (Fig. S1). These analyses were conducted at the genus level because virtually no overlap existed between continents at the species level. We subsequently ran a Detrended Component Analysis to visualize these floristic patterns across the tropics (Fig. 1). To assess if difference in species richness and turnover among regions are related to differences in environmental variability among regions, we performed a Principal Component Analysis (PCA) of locations versus environmental variables (climatic data (27) and edaphic data (28)) (Fig. S2).

Sample coverage and non-parametric estimators of species richness. Be-
cause estimates of species richness can be strongly dependent on differences
in inventory completeness (29, 30), we estimated the inventory completeness
for the complete database and for each region separately using the sample405
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409coverage estimator recommended by Chao & Shen (31), which is a reduced-
bias estimator of sample completeness:

$$\hat{C}_n = 1 - \frac{f_1}{n} \left[\frac{(n-1)f_1}{(n-1)f_1 + 2f_2} \right]$$

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414 where f_1 and f_2 are the number of species represented by one (singletons) 415 and two (doubletons) individuals in the sample, respectively, n is the total 416 number of individuals in the sample, and \hat{C}_n is the proportion of the total 417 number of individuals in an assemblage (observed and not observed) that 418 belong to the species represented in the sample. Sample coverage was almost identical in all regions (Table S1), suggesting that our regional comparisons of species richness were not biased by differences in sample coverage among 419 420 regions, nor was our pan-tropical estimate disproportionately influenced by 421 any one region (30).

422 Estimates of sample coverage indicated that more than 90% of all the trees present in the tropics belonged to species included in our recorded 423 samples (Table S1). We also computed estimates of species richness based 424 on several non-parametric estimators (Table S2). In particular, we considered nine of the estimators available in the software SPADE (31), and they 425 426 estimated that, on average, we have recorded between 79% and 91% of the species present in all regions (Table S2). However, consistent with ter Steege 427 et al. (15), we found that these methods underestimated the species richness, 428 as estimates for each tropical region (1,539 to 1,680 species in Africa; 4,959 to 429 5.540 in America: and 6.232 to 6.784 species in Asia) were between one-half 430 and one-third of previous estimates based on expert opinion and available floras (6-8). In fact, recent estimates suggest that there are approximately 431 16,000 tree species in the Amazon (15), so having fewer than 5,540 tree species in the whole Neotropics is highly unlikely (Table S2). As ter Steege 432 433 et al. (15) argued, the failure of these non-parametric methods to yield plausible estimates arises from fact that these estimators are designed to 434 estimate the expected number of species at a local scale, based on samples 435 that are fully representative of the area sampled (15, 32). However, like ter 436 Steege et al. (15), we are attempting to estimate the number of species for 437 the whole tropics, including areas that have been poorly sampled.

438 Selection of species-abundance distribution (SAD) model to predict 439 species richness. Several models have been proposed to describe observed species-abundance distributions (SADs) within a community (33). SAD mod-440 els allow for an understanding of the abundance structure of biological 441 communities, and can be useful for predicting unsampled portions of com-442 munities. The fit of SAD models to the data depends, among other factors, on community evenness and sampling intensity (33, 34). For instance, extremely 443 uneven SADs are predicted by the geometric series (35), while unusually even 444 SADs are predicted by the broken stick model (36). The log series (16) and log 445 normal (37) models are intermediate, differing in the assumed proportions of 446 rare species: Fisher's log series assumes very high proportions of rare species, while the log normal model assumes very low proportions of rare species (33). 447 Regarding sampling intensity, complete surveys usually follow log-normal 448 types of SADs, whereas incomplete sampled communities usually deviate 449 from log-normality (34).

450 Here we used Fisher's log series to estimate the expected number of species within each region. The log series distribution is one of the most 451 frequently used and thoroughly investigated models of the relationship 452 between species richness and the relative abundance of species (33), and it 453 has been successfully used to estimate the number of species at different 454 spatial scales (15, 25), though not previously at an intercontinental scale. 455 To fit and compare the log series model with other commonly used SAD 456 models (i.e., log normal, Broken-stick and Pareto [power-law] distributions) we used maximum likelihood tools with the sads package for R 3.0.3 (38). We 457 ranked the models from the best to the worst based on Akaike's information 458 criterion (AIC). The set of models with a difference in AIC (Δ) < 2 can be considered to have equivalently strong empirical support and similar 459 plausibility (39). 460

Graphical comparison of the models showed that the log series provided the best fit to our data (Fig. S3); this result was confirmed by the AIC analyses, in which the log series model had the strongest support (Table S3). The visual analysis of the frequency of species in octaves of abundance (so-called Preston plots) predicted by each SAD model also supported the idea that log series fit the data well and was the best model to predict the proportion of rare species (Figs. S4 - S7). Therefore, we can conclude that the log series is an appropriate SAD model to fit our data.

Application of the Fisher's log-series. Fisher's-alpha values can be used to extrapolate species richness of a defined region if the number of individuals is known. Extrapolations with Fisher's-alpha, however, rely on two assumptions: (i) species abundances are distributed in a log-series manner, and (ii) the plant community is homogeneous at the scale of the sampling. The first assumption held true in our case and is generally valid even for small sample sizes in tropical forests (13). The second assumption may be an adequate approximation at small spatial scales for most lowland tropical forests (15, 40) but is unquestionably violated at larger spatial scales due to increasingly biogeographically, environmentally, and spatially structured plant communities. In such cases a large number of randomly placed sample sites are needed to capture this heterogeneity, which might adequately approximate the homogeneity assumption at a large spatial scale. Despite477the large number of plots used in this study, the second assumption was not478completely met at the continental scale, as shown by the incomplete leveling479off of the Fisher's-alpha curves (Fig. 2a). This means that our regional Fisher's-480alpha values, and thus our species richness estimates remain minimum481

Lower-bound estimation of species richness. For each one degree grid 482 cell located within 500 km of our 207 locations (excluding major water 483 bodies), we calculated the total number of stems as the inverse distance-484 weighted average of the stem density observed in the five nearest locations. 485 multiplied by the size of each grid cell (Fig. S8). Latitudinal change in grid cell size was taken into account. Because our analysis was focused on original (po-486 tential) tree species richness, we ignored recent deforestation. Uncertainty 487 in this spatial extrapolation was assessed with a jackknife approach using 488 100 runs. Jackknifing removes each data point in turn and re-computes the spatial surface based on the remaining points. The differences between the original data values and the cross-validated values indicate the prediction 489 490 accuracy of the surface model (Fig. S8). 491

The lower-bound of species richness at the pan-tropical and continental 492 scales was then calculated using Fisher's log series, with Fisher's-alpha and 493 total number of stems as input variables at each spatial scale. Since the total number of stems was upscaled by several orders of magnitudes, one might 494 expect that this extreme extrapolation could potentially cause a large error 495 in our species richness estimates. However, a simple sensitivity test, in which 496 we varied stem numbers between 0.1 and 1.9 times the estimated values for 497 each spatial scale (a range much larger than the observed error in our stem number predictions), showed that this extreme perturbation in stem number resulted in only a 3-11% difference in number of species predicted (Table S4).

Upper-bound estimation of species richness. Our species richness estimates are near the true value if the unidentified species (individual recorded, but not identified) in the tree inventories, which we excluded from the analyses, follow the same rank abundance pattern as the identified species. However, if rare species are disproportionately represented in the unidentified category, our analyses may have underestimated the true Fisher'salpha and thus the species richness estimates. Because we did not know the relative abundance of rare species in the category of unidentified individuals, we selected a dataset of multiple tree inventories with a total of 10,647 individual trees from eastern Borneo for which this information was available. All individuals in this dataset were either identified to species or only to morpho-species (unidentified). A Mann-Whitney W test showed that, not surprisingly, rare species were disproportionately represented in the unidentified category (df = 1, N = 1103, W = 194798, p < 0.0001).

Using logistic regression of species square root transformed abundance versus identification status (identified versus unidentified) we predicted the probability that a tree species would be classified as identified. This 'classification probability' was then assigned to each individual tree belonging to that species. After adding a random number between zero and one to the 'classification probability' of each individual tree, to account for the fact that even individuals belonging to common species may remain unidentified, we sorted the whole list of individuals from high to low. We produced ten such sorted lists, each time varying the random number added to the 'classification probability' of an individual tree. For each list we could then count the number of species present within any level of identified individuals and calculate a minimum Fisher's-alpha. The average of these ten minimum Fisher's-alpha values could then be compared to the actual Fisher's-alpha observed for the whole list. Dividing the actual Fisher's-alpha by the minimum Fisher's-alpha gives an inflation factor with which observed Fisher's-alpha values can be multiplied to estimate the upper boundary of Fisher's-alpha for any percentage of identified species. The inflation factor (Y) showed a power-function relationship with ratio of identified individuals (X) given by: $Y = 1.2237 \times X^{-0.767}$. Using the appropriate inflation factor, we calculated the maximum expected species richness at all spatial scales (Table 1)

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 Figure 1. Overview of sample locations and their floristic affinities (point colors correspond

to scores on the first DCA-axis with similar colours indicating similar generic composition, while the lines indicate the floristic affinities as determined by cluster analysis).

Figure 2. Increase in Fisher's-alpha with (a) increasing numbers of locations (average of 50 replicates per region with random input order of locations), i.e. regional diversity and (b) increasing distance around locations (based on 50 replicates per region each with a randomly selected starting location), i.e. species turnover. Error bars indicate standard deviation among location re-orderings. Fisher's-alpha can decline if the number of stems added to the sample increases disproportionally to the number of new species detected. = EndOfDocument = =

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