

Land-cover change alters stand structure, species diversity, leaf functional traits, and soil conditions in Cambodian tropical forests

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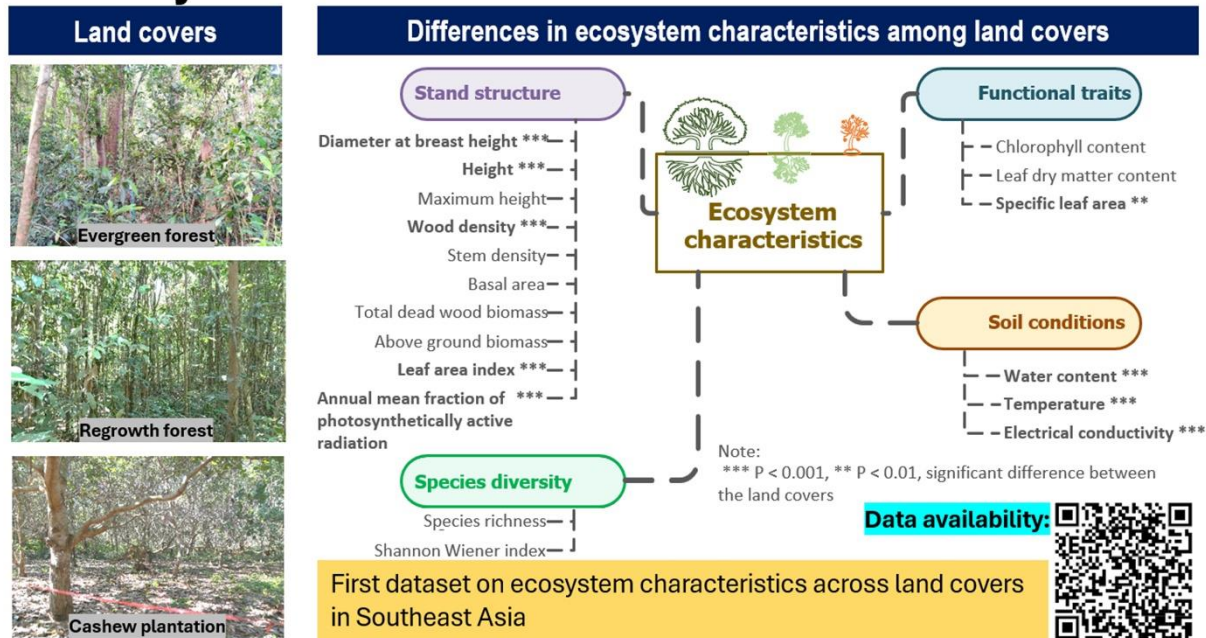
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Abstract. Given the severe land-use and land-cover change pressure on tropical forests and the high demand for field observations of ecosystem characteristics, it is crucial to collect such data both in pristine tropical forests and in the converted deforested land-cover classes. To gain insight into the ecosystem characteristics of pristine tropical forests, regrowth forests, and cashew plantations, we established an ecosystem monitoring site in Phnom Kulen National Park, Cambodia. Here, we present the first observed datasets at this site of forest inventories, leaf area index, leaf traits of woody species, a fraction of intercepted photosynthetically active radiation, and soil and meteorological conditions. Using these data, we aimed to assess how land-cover change affects stand structure, species diversity, leaf functional traits, and soil conditions among the three land-cover classes, and to evaluate the feasibility of locally calibrated diameters at breast height (*DBH*) and tree height (*H*) allometries for improving aboveground biomass estimation. We found significant differences in these ecosystem characteristics, caused by the anthropogenic land-cover conversion, which underlines land-cover change's profound impact on stand structure, species diversity, leaf functional traits, and soil conditions in these tropical forest regions. Our results further demonstrated the feasibility of locally updating aboveground biomass estimates using power-law functions based on relationships between *DBH* and *H*. These datasets and findings can contribute to enriching tropical forest research databanks and supporting sustainable forest management.

Keywords: tropical forest, forest ecosystem, forest inventory, biomass, Kulen, Cambodia.

Ecosystem characteristics of various land covers



1 Introduction

30 Tropical forests cover approximately 14 % of the Earth's surface (Fichtner and Härdtle, 2021) and contribute significantly to global terrestrial biodiversity (Giam, 2017) and biogeochemical cycles (Males et al., 2022). Tropical forests produce at least 30 % of the global terrestrial net primary production (Townsend et al., 2011; Wright, 2013) and account for approximately 70 % of the global gross carbon sink (Pan et al., 2024). In addition, they play a critical role in regulating hydrological cycles on a continental scale (Gloor et al., 2013). Tropical forests have been under severe anthropogenic pressures from agricultural land expansion, resource exploitation (logging, mining), and urbanization (Gardner et al., 2009; Laurance et al., 2014). Such disturbances have resulted in significant structural and functional degradation in tropical forests, highlighting the urgent need to assess how land-cover change alters key ecosystem characteristics (Barlow et al., 2016).

Southeast Asia, though harbouring roughly 15 % of the world's tropical forests (Stibig et al., 2014), has suffered the highest global deforestation rates over the past 15 years (Miettinen et al., 2011). This alarming trend threatens over 40 % of the region's biodiversity by 2100 (Sodhi et al., 2004). The forests are mainly disturbed by timber harvesting (Pearson et al., 2017), slash-and-burn agriculture, and agricultural plantations as a consequence of fulfilling global demands for timber production and agricultural commodities, especially rubber, cashew, oil palm, Eucalyptus and Acacia (Phompila et al., 2014; Grogan et al., 2015; Chen et al., 2016; Johansson et al., 2020). In addition to primary forests, secondary forests that regenerate after clear-cutting or other ecosystem disturbances are also important for protecting biodiversity and assuring the availability of ecosystem services and goods (Tito et al., 2022). While the ecological consequences of forest conversion are broadly recognized, relatively few studies have comprehensively examined how transitions from primary to secondary forests and plantations influence multiple ecosystem characteristics, particularly through detailed field-based observations that may inform our understanding of ecosystem functioning.

Tropical forests demonstrate remarkable ecological complexity, with high diversity in stand structure, species composition, and functional traits shaped by heterogeneous environments and varied disturbance histories (Manuel Villa et al., 2020). This complexity leads to highly site-specific and often inconsistent ecosystem characteristics and responses, making it difficult to generalize the impacts of land-cover change across regions (Wang et al., 2022). A parallel challenge in tropical forest research is the accurate estimation of aboveground biomass (*AGB*), a key metric for assessing carbon stocks. Most studies rely on generalized allometric models developed under different ecological conditions, assuming similarity in forest structure, composition, and wood density (*WD*), which these assumptions rarely hold in structurally diverse tropical forests (Vieilledent et al., 2012). These limitations introduce substantial uncertainty when models are applied across sites (Ketterings et al., 2001b). The lack of locally calibrated relationships between diameter at breast height (*DBH*) and tree height (*H*), wood density, and the scarcity of direct destructive sampling further contribute to estimation errors, highlighting the need for site-specific approaches that reflect local variation in forest structure and composition.

60 In the context of tackling the current challenges of global land cover change, it is necessary to conduct field observations in order to investigate the responses of ecosystems to changing environmental conditions on fine spatial and temporal scales.

Field observations of key ecosystem characteristics such as forest inventory, leaf functional traits, leaf area index (*LAI*), fraction of photosynthetically active radiation (*fPAR*), and soil conditions provide crucial insights into ecosystem functions and services, including vegetation productivity, carbon sequestration, hydrological cycle, ecosystem stability and resilience to disturbances, nutrient reservoir capacity, and the abundance of habitats of organisms (Naeem et al., 1994; Hector, 1998; Cardinale et al., 2012; Chen et al., 2016; Liang et al., 2016; Parisi et al., 2018b; Woodall et al., 2020). In addition, the field data on leaf functional traits, *LAI*, and *fPAR* are important for the parameterization and evaluation of remote sensing products and dynamic vegetation models, essential for modelling and upscaling ecosystem responses to anthropogenic disturbances and climate change (Feng et al., 2018; Fang et al., 2019; Pei et al., 2022). Recognizing the significant role and high demand for field observations of ecosystem characteristics, open data repositories such as FLUXNET, ICOS Carbon Portal, SpecNet, and the TRY database have been established to facilitate data sharing (Gamon et al., 2010; Kattge et al., 2020; Pastorello et al., 2020). Despite those global initiatives, observed data from tropical forests that support multi-class, pairwise comparisons for capturing ecosystem changes across gradients of forest degradation and land-use conversion remain limited. This gap is especially critical in Southeast Asia, where rapid forest-to-agriculture transitions threaten key ecosystem functions, and understanding these ecosystem changes is essential for informing evidence-based conservation and restoration strategies (Fan et al., 2024).

Within this context, Phnom Kulen National Park (Kulen) in Cambodia emerges as a critical landscape for investigating ecosystem responses to land cover change. Kulen is a hotspot for ecosystem service provisioning in Cambodia, mainly for water supply, potential carbon sink, and cultural services (Jacobson et al., 2022; Kim et al., 2023). It is the origin of the Khmer Empire and contains numerous archaeological sites. The stream water from the mountain is not only used to support local livelihoods in water supply and irrigation downstream (Somaly et al., 2020). It is also the primary water source to recharge surface water and groundwater aquifers in the Angkor Wat, UNESCO World Heritage Site. Hence, the area is of high importance to ensure that the temples' foundations remain stable and maintain their surrounding forest ecosystem (Hang et al., 2016). However, previous studies revealed that the forestland in and around Kulen has been disturbed, largely driven by agricultural expansion, particularly the spread of cashew plantations (Chim et al., 2019; Sovann et al., 2025).

Given increasing concerns about land use and land cover change threatening the high-value ecosystem functions of tropical forests like those in Kulen, our study aims to gain insight into the impact of land-cover conversion on key ecosystem characteristics. Specifically, our first objective is to assess the differences in stand structure, species diversity, leaf functional traits, and soil conditions between pristine tropical forests and the land cover the deforested regions are converted into (regrowth forests and cashew plantations). We hypothesise that there will be a reduction in stand structural complexity, species composition, and leaf functional traits, and a marked change in soil conditions. Additionally, our second objective is to evaluate the feasibility of locally updating aboveground biomass estimates by applying power-law functions derived from site-specific relationships between diameter at breast height and tree height, with the hypothesis that locally calibrated *DBH-H* relationships will have a substantial effect on estimated aboveground biomass compared to regional or generalized allometric models. To

95 test these hypotheses, we will analyse a novel in situ dataset collected from pristine tropical forests, regrowth forests, and
cashew plantations at a newly established ecosystem monitoring site in Phnom Kulen National Park, Cambodia.

2 Materials and Methods

2.1 Study area and selection of plots

100 The selected study area is the Phnom Kulen National Park located in the Siem Reap Province in north-west Cambodia (Fig.
1). It covers 37,380 ha predominantly on Jurassic-Cretaceous sandstone plateaus with the highest peak of 496 m (Matschullat,
2014; Geissler et al., 2019). In 2021, 72 % of Phnom Kulen National Park was forested, dominated by nearly intact tropical
evergreen forests (EF) (30 %) and forests that regrow naturally after clear-cutting (RF) (7 %). The remaining 35 % of forest
cover consisted of semi-evergreen, deciduous, and bamboo stands. Non-forest areas were dominated by household-scale
cashew plantations (CP) (15 %), with the remaining 13 % consisting of croplands, paddy fields, settlements, and tree and
105 rubber plantations (Sovann et al., 2025).

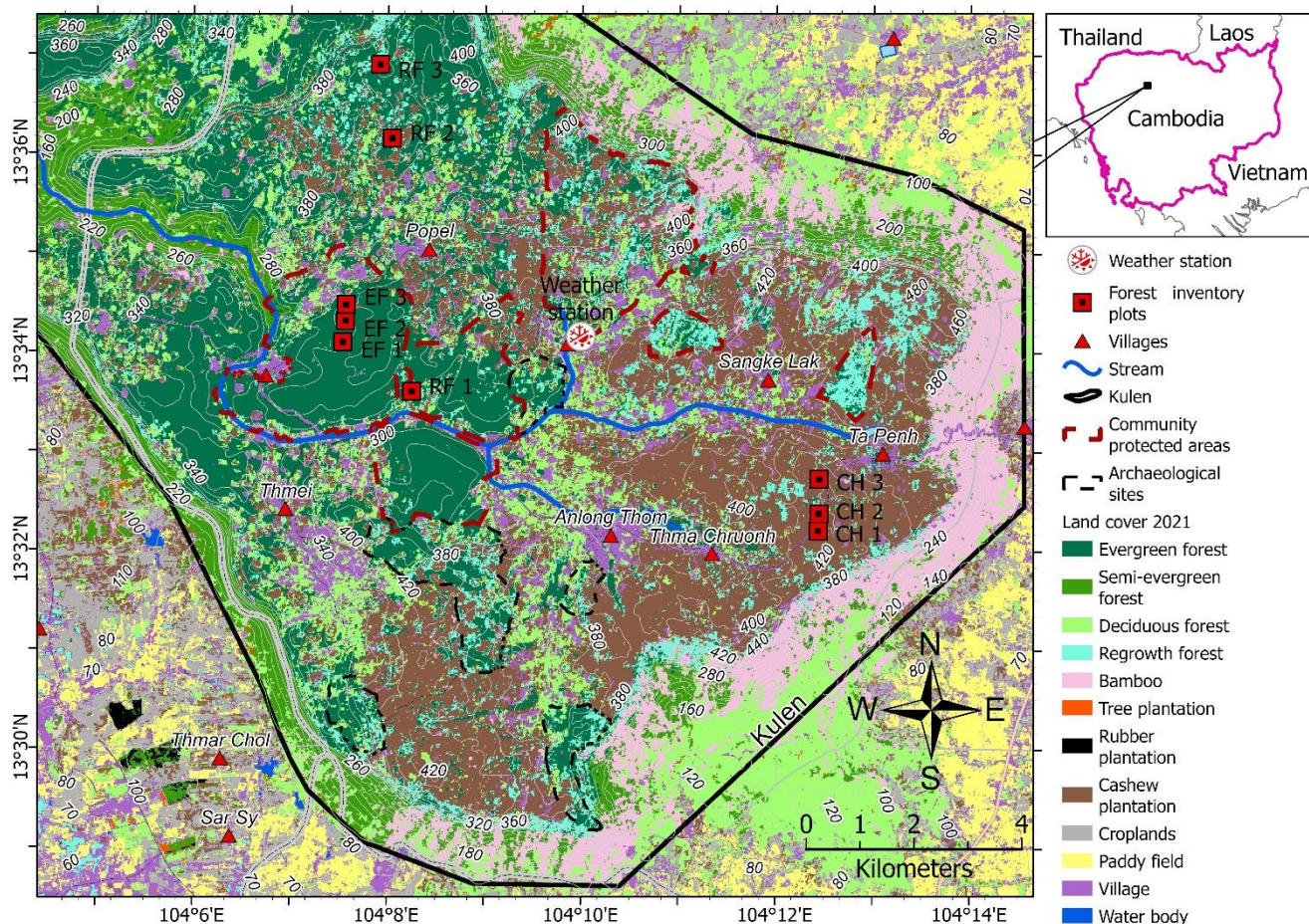


Figure 1. The locations of the nine forest inventory plots and the meteorological station in the Phnom Kulen National Park, Cambodia. Note: the background land cover 2021 was derived from Sovann et al. (2025).

110 Nine forest inventory plots were established in Kulen in December 2020, three within each of the EF, RF, and CP land-cover classes (Fig. 1, Table 1, Fig. S1.1), with a minimum separation of 250 meters to capture stand structure variation for each land-cover class. The EF plots represented tropical evergreen forests with no clear-cut history. The RF plots were dominated by at least 10-year-old natural regrowth forests, RF1 was clear-cut in 2009, while RF2 and RF3 experienced timber harvesting, burning, and fuelwood collection from 2006 to 2013. The CP plots were permanent rainfed cashew plantations, with cashew trees planted in 2013 in CP1 and in 2012 for the other two.

115 **Table 1. Characteristics of the forest inventory plots in Phnom Kulen National Park. Note: EF = evergreen forests, RF = regrowth forests, and CP = cashew plantations. Data source: Soil type and geology data from Matschullat (2014). Disturbance history information is obtained from field observation, discussion with local people, and combining with the Global Forest Change dataset of Hansen et al. (2013) and LandTrendr Pixel Time Series Plotter tool of Kennedy et al. (2018).**

Plot ID	Latitude, Longitude	Elevation (m)	Soil type	Disturbance history
EF1	N 13° 34' 12.4680" E 104° 7' 18.6096"	331	Acid Lithosols	No clear-cut history; affected by wind disturbance and human collection of wild honey, lychee, and other wild fruits in 2006, 2012, and 2014. Fewer large tree stands and lower vegetation cover density compared to EF2 and EF3.
EF2	N 13° 34' 25.3452" E 104° 7' 20.2872"	349	Acid Lithosols	No clear-cut history; past disturbances include cutting one lychee tree for fruit harvesting in 2022.
EF3	N 13° 34' 35.0508" E 104° 7' 20.6148"	339	Acid Lithosols	No clear-cut history; a wind-driven disturbance occurred in 2023.
RF1	N 13° 33' 42.6132" E 104° 8' 1.2408"	331	Red-yellow podzols	Evergreen forest clear-cut in 2009.
RF2	N 13° 36' 15.6924" E 104° 7' 48.8928"	371	Acid Lithosols	Timber harvesting, burning, and fuelwood collection of an evergreen forest from 2006 to 2013.
RF3	N 13° 37' 0.3612" E 104° 7' 41.358"	401	Acid Lithosols	Timber harvesting, burning, and fuelwood collection of an evergreen forest from 2006 to 2013.
CP1	N 13° 32' 18.8988" E 104° 12' 12.5568"	429	Red-yellow podzols	Cashew plantation established in 2013.
CP2	N 13° 32' 29.3100" E 104° 12' 13.0284"	422	Red-yellow podzols	Cashew plantation established in 2012.
CP3	N 13° 32' 50.1864" E 104° 12' 13.1544"	430	Red-yellow podzols	Cashew plantation established in 2012.

2.2 Data collection

120 2.2.1 Forest inventory

The forest inventory was performed by following the standard method of the National Forest Inventory of Cambodia (Than et al., 2018). Each plot was designed as a rectangle with 50 m x 30 m long edges in the south-north and west-east directions. The plots were further subdivided into five subplots with the following dimensions: 2 m x 2 m, 5 m x 5 m, 10 m x 10 m, 30 m x 15 m, and 30 m x 50 m (Fig. S1.2). In the 2 m x 2 m subplots, seedlings with diameters at breast height (1.3 m above ground) of less than 1 cm were recorded. In the 5 m x 5 m, 10 m x 10 m, 30 m x 15 m, and 30 m x 50 m subplots, trees with *DBH* ranges of 1–5 cm, 5–15 cm, 15–30 cm, and greater than 30 cm were measured, respectively.

For seedlings, we only recorded the total numbers of each species. For the *DBH* range of 1–5 cm, we noted the *DBH*, tree height, species, local name (Khmer), and position of each tree. For trees with a *DBH* greater than 5 cm, we collected the same data as for trees with a *DBH* of 1–5 cm, plus bole height (the height from the ground to the first main lowest stem), health (healthy or infected), quality (straight, bent, or crooked stem), origin (natural or planted), and stump diameter and height (measured 15 cm above ground for annual tree growth monitoring).

Deadwood is a significant indicator of decomposition and nutrient cycling processes in a forest ecosystem (Shannon et al., 2021). Data on lying and standing deadwood with a *DBH* greater than 10 cm in the 30 m x 15 m subplots were also collected. The deadwood decomposition levels were classified into five scales, based on harmonizing the scaling systems of the National Forest Inventory of Sweden (Swedish NFI, 2019) and Cambodia (Than et al., 2018) (Table S1.1). For standing deadwood, we recorded their species, local name, location, height, and decomposition level. For lying deadwood, we counted the number of pieces and measured their lengths, base and tree diameters, and decomposition levels.

2.2.2 Leaf sample collection and measurement

A total of 453 leaf samples from 30 woody species were collected inside and 500 m around the forest inventory plots in December 2019 and August 2022. Each species was represented by five to 47 leaf samples. Each leaf's fresh mass, chlorophyll content, and photo were taken in the field. A Chlorophyll Meter (SPAD 502 Plus; Konica Minolta Sensing Inc., Japan) was used in situ to measure chlorophyll content five times on each leaf surface to retrieve a leaf mean value. The given measurement unit was in SPAD value (Soil Plant Analysis Development) and later converted to chlorophyll a and b content (*Chl*) in $\mu\text{g cm}^{-2}$ (Coste et al., 2010). We obtained fresh leaf mass by weighting in the field and leaf dry mass by oven-drying the leaves at 60 °C until the leaf mass remained constant (oven-dried for at least three days) (Garnier et al., 2001). The leaf photos were used for estimating leaf lengths and areas using ImageJ (Schindelin et al., 2012; Schneider et al., 2012).

2.2.3 Meteorological and photosynthetically active radiation data

A meteorological station was installed in an open area to continuously record meteorological conditions, and incoming photosynthetically active radiation (*PAR*) for the wider area (the Kulen National Park). Data were sampled at one minute

150 intervals and stored as 15 minute averages (sum for rainfall). The installation was done in November 2020 in Khnang Phnom Commune, Svay Luer District, Siem Reap Province, at 13° 34' 16.1148" N, 104° 9' 45.6768" E, and an altitude of 314 m above mean sea level. The station has one Atmos 41 meteorological station (Meter Group Inc. WA, USA), installed 2.2 m above ground level, measuring rainfall, wind speed, wind direction, global radiation, atmospheric pressure, and air temperature. Additionally, four *PAR* sensors (SQ-110-SS, Apogee Instruments, Inc., UT, USA) were positioned 2 m above the ground to
 155 record incoming *PAR* (PAR_{inc}) (Fig. S2.1).

Six additional loggers with five *PAR* sensors (SQ-521-SS and SQ-110-SS, Apogee Instruments, Inc., UT, USA) and one Tيروس 12 soil moisture sensor each (Meter group Inc. WA, USA), collecting data at a 15 minute mean timestep, were installed in six of the forest inventory plots in April 2022. The soil moisture sensors were installed at a depth of 20 cm to measure soil water content (*SWC*), soil temperature (*T_s*), and soil electrical conductivity (*ECs*). Two loggers were placed in each land-cover
 160 class (EF, RF, and CP). The selection of plots in each land-cover class was based on previous measurements of leaf area index and the loggers were placed at the plots with the highest and lowest *LAI* for each land cover, respectively. Thus, the selected plots for installing *PAR* sensors were EF1, EF3, RF1, RF3, CP2, and CP3 (Fig. 1). The *PAR* sensors were placed with one in the centre of the plot and the other four placed 15 ± 1 m apart at 30°, 150°, 220°, and 330° from the north. In cases of unfavourable field conditions, such as high termite nests or being too close to a tree, the locations were adjusted 0.5–1 m east
 165 or west of the planned position. Each *PAR* sensor was mounted on 1.3 m poles to record *PAR* below canopy data. We calculated the fraction of *PAR* intercepted by the stand canopy for each plot using Eq. (1) (Olofsson and Eklundh, 2007). Each Tيروس 12 soil moisture sensor was installed at a depth of 20 cm in the middle of the six plots to measure *SWC*, *T_s*, and *ECs*. The data of *fPAR* and soil conditions from two plots within the same land-cover classes were averaged to represent those classes.

$$fPAR = \frac{(PAR_{inc} - PAR_{below})}{PAR_{inc}} \quad (1)$$

Where PAR_{inc} and PAR_{below} are photosynthetically active radiation above and below canopy ($\mu\text{mol m}^{-2} \text{s}^{-1}$). *fPAR* is in
 170 percentage.

2.2.4 Leaf area index measurements

We measured each plot's total one-sided leaf surface area per unit ground area, *LAI*, using a LAI-2000 Plant Canopy Analyzer (LI-COR, NE, USA). The measurements were conducted six times across two seasons: four times during the dry season (November/December 2019, November 2020, December 2020, and March 2021) and twice during the rainy season (September
 175 2020 and June 2021). The measurements were taken both at ground level to capture the total *LAI* (LAI_T) and at breast height to specifically assess tree canopy *LAI* (LAI_C) within two diagonal transects across the 50 m x 30 m rectangular plots. On each measurement occasion, we collected between 32 and 75 samples, except for the ground-level measurements of the RF3 plot in December 2020, where only ten samples were collected due to technical issues.

2.3 Data analysis

180 2.3.1 Species diversity

We investigated the species diversity of various land covers by calculating species richness (S_R) and the Shannon-Wiener index (S_H) (Shannon, 1948). The S_R was determined by summing the number of tree species in each plot. The S_H is commonly used to quantify species richness and evenness in a community by representing the number of species and how equally individuals are distributed among them (Hill, 1973). The value of S_H increases as the number of species and the degree of evenness increase. The S_H was calculated by:

$$S_H = - \sum_{i=1}^n P_i \ln(P_i) \quad (2)$$

Where S_H is Shannon-Wiener index (unitless), P_i is a proportion of i species in a community (unitless), and n is the number of species in a plot (unitless). We calculated the S_R and S_H at the plot level and then averaged the values for each land-cover class.

2.3.2 Functional traits and diversity

190 We computed the specific leaf area (SLA) for each of the 453 leaf samples as the ratio of leaf area to leaf dry mass. Likewise, leaf dry matter content ($LDMC$) was calculated by the ratio of dry leaf mass to fresh leaf mass (Garnier et al., 2001; Akram et al., 2023). We estimated the trait community-weighted means and standard deviations of SLA_{cwm} , $LDMC_{cwm}$, and Chl_{cwm} to represent ecosystem functions and their diversity at the land-cover level (Garnier et al., 2004; Leoni et al., 2009; Wang et al., 2020) with:

$$T_{cwm} = \frac{\sum_{i=1}^n W_i T_i}{\sum_{i=1}^n W_i} \quad (3)$$

195 Where T_{cwm} is trait community-weighted mean for SLA , $LDMC$, or Chl , T_i is the species-specific trait value tree i , n is total number of trees, W_i is the weight (volume based) value of the tree, assuming that larger trees have a greater impact on the ecosystem function (Chave et al., 2005; Feldpausch et al., 2011). Before computing T_{cwm} for each trait, we addressed missing species traits within each plot by first taking values from a different plot with the same land-cover class. If unavailable, we used values from the same species across all nine plots, followed by values from the genus and family levels. When multiple
200 genera or families were available, we averaged the values. If neither was available, we used the mean trait value of the plot.

2.3.3 Stand structural attributes

We examined the differences in *DBH*, *H*, basal area (*BA*), aboveground biomass, and deadwood biomass (*DWB*) for the various land-cover classes to characterize stand structure attributes. Deadwood volumes (V_{DW} , m³) for each bole were determined by Smalian's equation:

$$V_{DW} = (\pi H_b) \frac{(D_{base}^2 + D_{top}^2)}{8} \quad (4)$$

205 Where D_{base} and D_{top} are diameters at base and top (m), and H_b is the length/height of the trunk (m).

Deadwood biomass was then received by multiplying V_{DW} with a mean deadwood density of 0.45 g cm⁻³ (Kiyono et al., 2007). Total *DWB* was computed plot-wise by taking the sum of lying and standing *DWB*. *DWB* for each land-cover class was calculated as the average of the total *DWB* across the plots within that land-cover class.

Basal area was determined plot-wise by combining the *DBH* of all living trees within a plot:

$$BA = \sum_{i=1}^n \pi \left(\frac{DBH_i}{2} \right)^2 \left(\frac{10^4}{A_i} \right) \quad (5)$$

210 Where *BA* is a plot-wise total basal area of all living trees (m² ha⁻¹), *n* is a number of trees in a plot, DBH_i is the diameter at breast height of tree *i* in a sampling plot (m), $\pi \left(\frac{DBH_i}{2} \right)^2$ is the circle basal area of tree *i* (m²), $\left(\frac{10^4}{A_i} \right)$ are the scaling factors employed to convert the sampled subplot area (A_i) to one hectare (unitless). The *BA* for each land-cover class was represented by the mean *BA* of all plots within a class.

We calculated the mean and standard deviation of *DBH* and *H* for each plot and land cover. We further used these for
215 establishing relationships between *DBH* and *H*, as such relationships serve as functional traits characterizing tree growth patterns and successional stages within forest communities (Nyirambangutse et al., 2017; Howell et al., 2022). We used natural logarithms and then converted them to power-law relationships both plot- and land-cover class-wise (West and Brown, 2005). An ordinary least-square linear regression (OLS) was applied to investigate the *DBH-H* relationship, followed by transforming the relationship into a power-law relationship (Huxley, 1932).

$$H = K_1 DBH^{K_2} \quad (6)$$

220 Where K_1 and K_2 are the power-law intercept and slope, respectively. The K_1 captures the overall scaling relationship between *H* (m) relative to *DBH* (cm) within a forest community while K_2 regulates the rate of *H* increase relative to *DBH* growth. The obtained K_1 and K_2 values were further used to estimate *AGB* (AGB_h) Eq. (7) in Table 2. We also computed the *AGB* using existing equations (Table 2, Eqs. (9–11)) (AGB_i) adopted for the three different land-cover classes. These EF and RF allometric equations were developed for tropical multiple species, whereas the CP was a species-specific allometric equation for the

225 cashew tree (Malimbwi et al., 2016). The wood density values required for the *AGB* estimations were species-specific and obtained from The International Council for Research in Agroforestry (2022) and Zanne et al. (2009). When multiple *WD* values for a tree species were available, the mean value was used, whereas when no species-specific *WD* values were available, the average of tropical Asia (0.57 g cm^{-3}) was used (Reyes et al., 1992). The applied *WD* values for this study then ranged from $0.39\text{--}1.04 \text{ g cm}^{-3}$. Specifically, the *WD* values (mean $\pm$ a standard deviation) for EF, RF, and CP were $0.74 \pm 0.17 \text{ g cm}^{-3}$, $0.72 \pm 0.15 \text{ g cm}^{-3}$, and 0.45 g cm^{-3} , respectively. We first estimated *AGB* at the plot level in kilograms, then scaled these values to megagrams per hectare, and averaged per land-cover class.

Table 2. Allometric equations used for estimating aboveground biomass (*AGB*, kg tree⁻¹) in the different land-cover classes.

No.	Equations	Land cover	<i>AGB</i> allometric equations	Regions	n	<i>DBH</i> (range, cm)	$\overline{WD}_f$ (mean $\pm$ SD, g cm ⁻³)	References
1	Eq. (7)	All	$AGB_h = \frac{WD \pi K_1}{8} DBH^{2+K_2} + \varepsilon$	-	-	-	-	This study
2	Eq. (8)	All	$AGB_{wd} = \frac{WD}{\overline{WD}_f} AGB_f$	-	-	-	-	This study
3	Eq. (9)	EF	$AGB_f = 0.1184 DBH^{2.53}$	Pantropical	170	5.0–148.0	0.58 ± 0.02	Brown (1997)
4	Eq. (10)	RF	$AGB_f = 0.0829 DBH^{2.43}$	Sarawak, Malaysia	136	0.1–28.7	0.38 ± 0.07	Kenzo et al. (2009)
5	Eq. (11)	CP	$AGB_f = 0.8450 DBH^{1.77}$	Pwani, Tanzania	45	6.0–89.9	0.18	Malimbwi et al. (2016), Mlagalila (2016)

235 **Note:** EF is evergreen forests, RF is regrowth forests, CP is cashew plantations. In Eqs. (9–11), *DBH* is diameter at breast height (cm), and $\overline{WD}_f$ is the reported mean wood density used in *AGB_f* (kg m⁻³). In Eq. (7), *K*₁ and *K*₂ are derived power-law intercept and slope values between *DBH* (cm) and tree height (*H*, m) relationship in Eq. (6), ε is a statistical error term, *WD* is wood density for each tree species (g cm⁻³), and *DBH* is in centimetres. In this study, in Eq. (7), we employed a trunk shape factor of 1/8 for calculating the volume of frustum cones, as proposed by King et al. (2006). This factor falls within the range of 1/4 (cylinder volumes) to 1/12 (cone volumes). In Eq. (8), *AGB_{wd}* is our examined aboveground biomass based on equations Eqs. (9–11) with species-specific wood density updated for our woody tree species, *WD* are the species-specific wood density of trees in each plot (g cm⁻³).

240 2.3.4 Statistical analysis

Descriptive statistics were conducted to examine the difference in ecosystem characteristics between plots and land-cover classes. One-way ANOVA tests (ANOVA) were used to assess significant differences in mean values across land-cover classes. Tukey's Honestly Significant Difference test (Tukey HSD) was further employed for pairwise comparisons between land-cover classes. Pearson correlation and ordinary least squares regression analyses were used to explore relationships between variables. All analyses were performed using R 4.2.3 (R Core Team, 2023).

3 Results

3.1 Meteorological and soil conditions

The observed annual daily mean air temperature from April 2022 to April 2023 at Kulen meteorological station was 24.2 ± 2.0 °C, varying between 17.8 °C and 28.6 °C (Fig. 2a). The total annual rainfall was 2290 mm, significantly surpassing nearby
250 lowland stations: Banteay Srei station, located 22 km west, recorded 1160 mm, and Siem Reap City station, situated 40 km southwest, recorded 1475 mm (Chim et al., 2021). About 90 % of the annual precipitation fell during the rainy season from May to November, with September being the wettest month (505 mm). The daily maximum rainfall can reach up to 141 mm, but the daily mean during the rainy season was 11.2 ± 19.7 mm (Fig. 2b). The annual daily mean of global radiation, relative humidity, vapour pressure deficit, and wind speed were 172 ± 44 W m⁻², 88 ± 12 %, 0.45 ± 0.21 kPa, and 0.68 ± 0.22 m s⁻¹,
255 respectively (Fig. 2c–f).

Soil conditions varied significantly among land-cover classes (ANOVA and Tukey HSD, p-value < 0.001). Annual daily mean soil temperature was highest in CP (25.8 °C), exceeding values in EF (24.3 °C) and RF (24.2 °C). In contrast, annual daily mean soil water content was lowest in RF (0.14 m³ m⁻³) compared to EF (0.23 m³ m⁻³) and CP (0.21 m³ m⁻³) (Table 3). Annual daily mean soil electrical conductivity was highest in EF (0.039 dS m⁻¹), followed by RF (0.032 dS m⁻¹) and CP (0.025 dS m⁻¹).
260 Overall, daily mean values across land-cover classes ranged between 0.14 – 0.23 m³ m⁻³ for SWC, 24.2 – 25.8 °C for T_s , and 0.025 – 0.039 dS m⁻¹ for EC_s (measured at 20 cm depth, Fig. 2g–i).

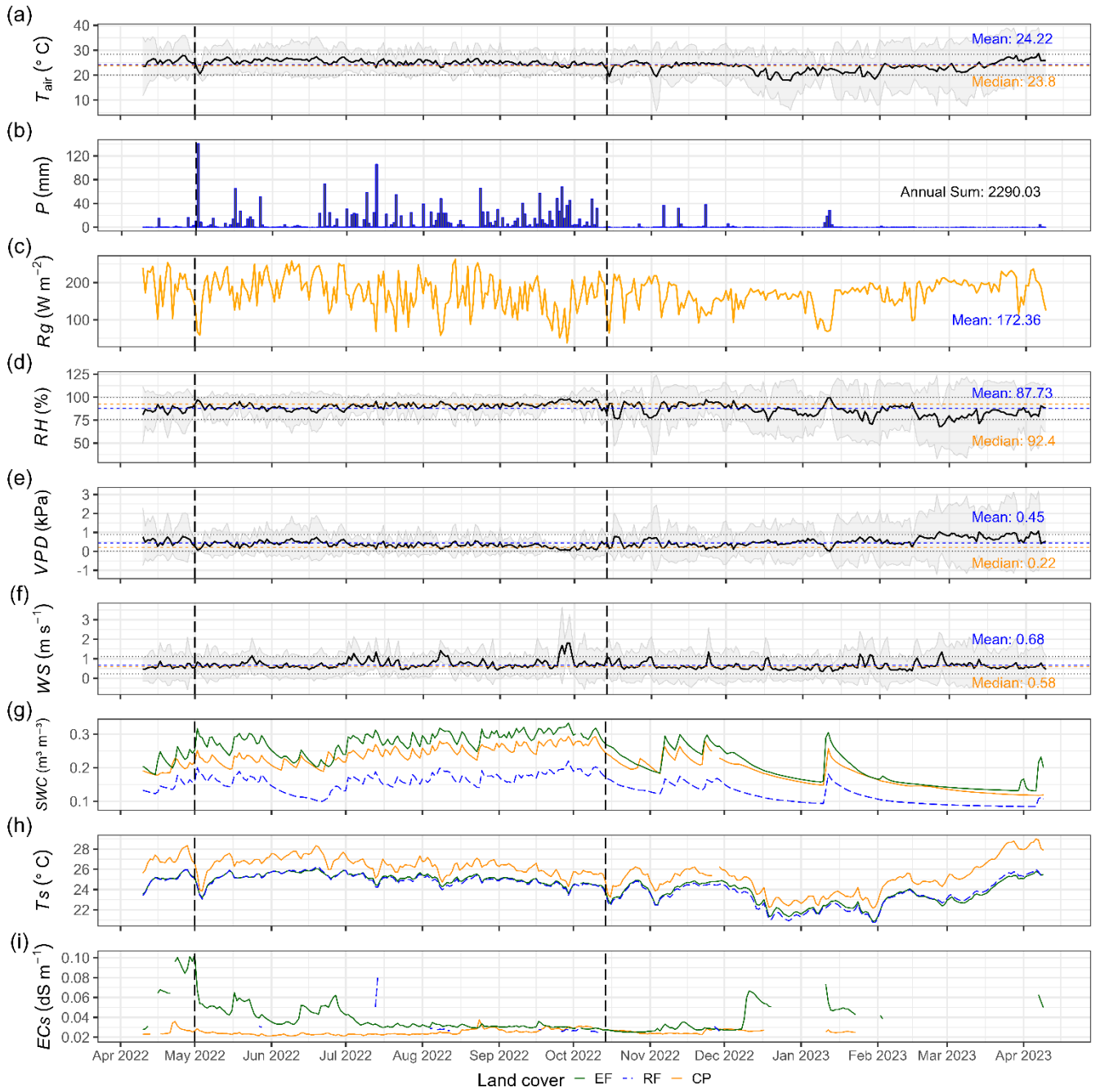


Figure 2. The meteorological conditions at Kulen meteorological station (a–f), and soil conditions at each land-cover class (g–i) from April 10, 2022, to April 9, 2023. (a) Daily mean air temperature (T_{air} , °C), (b) daily total precipitation (P , mm), (c) daily mean global radiation (R_g , W m^{-2}), (d) daily mean relative humidity (RH , %), (e) daily mean vapour pressure deficit (VPD , kPa), and (f) daily mean wind speed (WS , m s^{-1}), (g) daily mean soil water content (SWC , $\text{m}^3 \text{m}^{-3}$), (h) daily mean soil temperature (T_s , °C), (i) daily mean soil saturation extraction electrical conductivity (EC_s , dS m^{-1}). The vertical dashed line region in all the plots highlighted the rainy season period in Cambodia from May to October. The grey-shaded regions around the mean in (a), (d), (e), and (f) represent the 95 % confidence interval (using a standard deviation) from the daily mean, whereas the blue horizontal dashed line represents

the yearly mean, the brown horizontal dashed line represents the yearly median, and the black horizontal dotted line represents a yearly standard deviation (see Table S2.1 and Fig. S2.2 present the Kulen meteorological station's annual and monthly meteorological data. Figs. S3.1–S3.3 shows monthly mean soil conditions by land-cover class, and Fig. S4.1 depicts correlations between meteorological and soil conditions).

275 3.2 Species diversity

A total of 343 observations (292 trees and 51 seedlings) from 47 woody species (including 13 seedling species) and 32 families (including seven seedling families) were identified from the nine plots (Table S5.1). No statistical test of significance of differences in species diversity among land cover classes was possible due to too few sampled plots. However, species diversity declined markedly from evergreen forests to regrowth forests and was lowest in cashew plantations, as reflected in both species richness and in the Shannon-Wiener index. The average S_R per plot was 17 in EF, 13 in RF, and only 4 in CP. Similarly, the S_H was highest in EF (2.48 ± 0.33), intermediate in RF (1.97 ± 0.45), and lowest in CP (0.61 ± 0.46), with individual plot values ranging from 0.31 (CP2) to 2.68 (EF1) (Table S5.2). Species composition was more evenly distributed in EF and RF, but naturally strongly dominated by a single species in CP. In EF, the top five most abundant species, *Mesua ferrea* (n = 18), *Diospyros bejaudii* (n = 12), *Litchi chinensis* (n = 11), *Vatica odorata* (n = 11), and *Hydnocarpus annamensis* (n = 8), accounted for 46 % of the individuals. In RF, *Vatica odorata* (n = 54), *Nephelium hypoleucum* (n = 14), *Benkara fasciculata* (n = 12), *Garcinia oliveri* (n = 12), and *Mesua ferrea* (n = 5) made up 61 %. In contrast, CP was dominated by *Anacardium occidentale* (n = 46), which was the only tree species observed excluding seedlings. Additional seedling species in CP included *Strychnos axillaris* (n = 3), *Nephelium hypoleucum* (n = 1), *Melodorum fruticosum* (n = 1), *Maclura cochinchinensis* (n = 1), and *Catunaregam tomentosa* (n = 1). Furthermore, fast-growth species, as described by Ha (2015) ($WD < 0.6 \text{ g cm}^{-3}$), accounted for 40 % of EF and 44 % of RF of their total species composition.

Table 3. Mean values and statistics of ecosystem characteristics in the different land-cover classes.

Group	Variables	Land cover						Tukey HSD			
		EF		RF		CP		A	N	O	V
		(Mean $\pm$ SD)	n	(Mean $\pm$ SD)	n	(Mean $\pm$ SD)	n	p-value	p-value	p-value	p-value
Species diversity	S_R (with seedling species, count per plot)	17 $\pm$ 4	3	13 $\pm$ 2	3	4 $\pm$ 3	3	-	-	-	-
	S_R (without seedling species, count per plot)	13 $\pm$ 2	3	10 $\pm$ 3	3	1 $\pm$ 0	3	-	-	-	-
	S_H (with seedling species, unitless)	2.48 $\pm$ 0.33	3	1.97 $\pm$ 0.45	3	0.61 $\pm$ 0.46	3	-	-	-	-

Leaf functional traits	Chl_{cwm} (mg g ⁻¹)	9.14 ± 3.45	109	7.56 ± 2.03	137	4.99 ± 0.66	46	*	*	0.39	0.08
	$LDMC_{cwm}$ (mg g ⁻¹)	398.43 ± 72.24	109	370.13 ± 94.97	137	407.64 ± 21.68	46	0.51	0.50	0.94	0.69
	SLA_{cwm} (m ² kg ⁻¹)	18.18 ± 2.86	109	14.87 ± 2.06	137	11.99 ± 1.45	46	**	**	*	0.06
Stand structure	DBH (cm)	18.0 ± 20.1	109	5.8 ± 4.3	137	13.0 ± 3.9	46	***	0.14	***	***
	H (m)	17.0 ± 13.3	109	7.4 ± 3.8	137	6.3 ± 1.0	46	***	***	***	0.93
	Maximum H (m)	52.0	109	18.6	137	7.8	46	-	-	-	-
	Wood density (g cm ⁻³) [†]	0.74 ± 0.17	109	0.72 ± 0.15	137	0.45 ± 0.00	46	***	***	0.56	***
	Stem density $DBH > 1$ cm (ha ⁻¹) ^{††}	6216 ± 2177	3	10859 ± 4999	3	1067 ± 440	3	-	-	-	-
	Stem density $DBH > 5$ cm (ha ⁻¹) ^{††}	1016 ± 533	3	2193 ± 895	3	1067 ± 440	3	-	-	-	-
	Stem density $DBH \geq 10$ cm (ha ⁻¹) ^{††}	550 ± 505	3	293 ± 6	3	600 ± 164	3	-	-	-	-
	BA (m ² ha ⁻¹)	26.2 ± 3.6	3	17.0 ± 5.4	3	11.6 ± 3.5	3	-	-	-	-
	BA (m ² ha ⁻¹ , $DBH \geq 5$ cm)	23.7 ± 4.4	3	11.6 ± 2.4	3	11.6 ± 3.5	3	-	-	-	-
	BA (m ² ha ⁻¹ , $DBH \geq 10$ cm)	21.1 ± 4.4	3	4.4 ± 0.7	3	9.2 ± 1.8	3	-	-	-	-
	DWB (Total) (Mg ha ⁻¹)	27.5 ± 12.4	3	4.8 ± 7.0	3	0.4 ± 0.2	3	-	-	-	-
	AGB_f (Mg ha ⁻¹)	239 ± 92	3	42 ± 10	3	71 ± 22	3	-	-	-	-
	AGB_{wd} (Mg ha ⁻¹)	336 ± 168	3	78 ± 25	3	182 ± 57	3	-	-	-	-
	AGB_h (Mg ha ⁻¹)	312 ± 184	3	54 ± 14	3	17 ± 5	3	-	-	-	-
	LAI_C (m ² m ⁻²)	4.62 ± 0.50	21	4.66 ± 0.70	21	2.52 ± 0.42	21	***	***	1.00	***
	LAI_T (m ² m ⁻²)	6.16 ± 0.67	21	5.57 ± 0.76	21	3.07 ± 0.61	21	***	***	0.08	***
Soil conditions	Annual mean $fPAR^{\ddagger}$	0.97 ± 0.01	364	0.96 ± 0.01	365	0.76 ± 0.06	359	***	***	*	***
	Annual mean $SWC^{\ddagger}$ (m ³ m ⁻³)	0.23 ± 0.06	364	0.14 ± 0.03	365	0.21 ± 0.05	363	***	***	***	***
	Annual mean $Ts^{\ddagger}$ (°C)	24.3 ± 1.2	364	24.2 ± 1.3	365	25.8 ± 1.5	363	***	***	***	***
	Annual mean $ECs^{\ddagger}$ (dS m ⁻¹)	0.039 ± 0.015	268	0.032 ± 0.013	40	0.025 ± 0.003	260	***	***	***	***

Note: Abbreviations used in the table: EF = evergreen forests, RF = regrowth forests, CP = cashew plantations, S_R = species richness (only woody seedling species), S_H = Shannon-Wiener index, Chl_{cwm} = community-weighted mean of chlorophyll a and b content, $LDMC_{cwm}$ = community-weighted mean of leaf dry matter content, SLA_{cwm} = community-weighted mean of specific leaf area, DBH = tree's diameter at breast height, H = tree height, BA = stand basal area, AGB_f = aboveground biomass computed by adopted functions, AGB_h = aboveground biomass computed by H and DBH power-law relationship, AGB_{wd} = aboveground biomass based on equations Eqs. (9–11) with species-specific wood density updated for our woody tree species, LAI_C = canopy leaf area index, LAI_T = total leaf area index, $fPAR$ = fraction of photosynthetically active radiation, SWC = soil water content, Ts = soil temperature, ECs = soil saturation extract electrical conductivity, SD = a standard deviation, ANOVA = one-way analysis of variance, Tukey HSD = Tukey's Honestly Significant Difference test. Statistically significant code for ANOVA and Tukey HSD test: '***' p-value < 0.001, '**' p-value < 0.01, '*' p-value < 0.05, and '-' not available. [†]The species-specific wood density was derived from the ICRAF Database (2022) and Zanne et al. (2009). ^{††}Extrapolated values for one hectare were obtained from sampling DBH class subplots. [‡]Daily mean values were used to calculate the reported variables.

3.3 Leaf functional traits

305 At the species level, the mean specific leaf area for all 30 species was $16.97 \pm 5.30 \text{ m}^2 \text{ kg}^{-1}$, with *Hydnocarpus annamensis* having the highest *SLA* ($36.67 \pm 5.20 \text{ m}^2 \text{ kg}^{-1}$) and *Capparis micracantha* the lowest ($10.46 \pm 3.28 \text{ m}^2 \text{ kg}^{-1}$). For *Chl*, the mean value was $10.28 \pm 4.17 \text{ mg g}^{-1}$, with *Hydnocarpus annamensis* having the highest value ($25.75 \pm 5.28 \text{ mg g}^{-1}$) and *Anacardium occidentale* the lowest ($4.86 \pm 4.93 \text{ mg g}^{-1}$). Finally, for *LDMC* the mean value was $378.96 \pm 143.26 \text{ mg g}^{-1}$, with *Mesua ferrea* and *Hydnocarpus annamensis* having the highest ($486.90 \pm 25.03 \text{ mg g}^{-1}$) and lowest ($139.92 \pm 20.19 \text{ mg g}^{-1}$) values, respectively. For detailed descriptions of leaf functional traits of all species and plots, please refer to Tables S6.1–S6.3.

310 Across land-cover classes, mean *SLA*_{cwm} and *Chl*_{cwm} decreased from EF to RF to CP. *SLA*_{cwm} and *Chl*_{cwm} were highest in EF ($18.18 \pm 2.86 \text{ m}^2 \text{ kg}^{-1}$ and $9.14 \pm 3.45 \text{ mg g}^{-1}$) followed by RF ($14.87 \pm 2.06 \text{ m}^2 \text{ kg}^{-1}$ and $7.56 \pm 2.03 \text{ mg g}^{-1}$) and CP ($11.99 \pm 1.45 \text{ m}^2 \text{ kg}^{-1}$ and $4.99 \pm 0.66 \text{ mg g}^{-1}$). Both traits showed statistically significant differences across land covers (ANOVA p-value < 0.002 for *SLA*_{cwm}, p-value < 0.018 for *Chl*_{cwm}). In contrast, *LDMC*_{cwm} did not differ substantially among land-cover classes (p-value = 0.51), with CP having the highest value ($407.64 \pm 21.68 \text{ mg g}^{-1}$), followed by EF ($398.43 \pm 72.24 \text{ mg g}^{-1}$) and RF ($370.13 \pm 94.97 \text{ mg g}^{-1}$). See Table S6.4 for data sources and shared percentages of species trait values used to compute *SLA*_{cwm}, *Chl*_{cwm}, and *LDMC*_{cwm}.

3.4 Stand structure attributes

DBH and tree height

320 Land-cover conversion reduces both the mean and variability of tree diameter and height, indicating a loss of structural complexity in human-disturbed ecosystems. Structural measurements of 292 woody trees across three land-cover classes showed that EF had the highest structural complexity, with the highest mean and variability in *DBH* ($18.0 \pm 20.1 \text{ cm}$) and tree height ($17.0 \pm 13.3 \text{ m}$), including the largest individuals (*DBH* = 102.3 cm, *H* = 52.0 m, Fig. S7.1). However, RF and CP had substantially lower means and variability in these variables, suggesting reduced structural complexity after forest conversion.

325 While both RF and CP had similar heights (RF: $7.4 \pm 3.8 \text{ m}$, CP: $6.3 \pm 1.0 \text{ m}$), CP had a significantly greater *DBH* (CP: $13.0 \pm 3.9 \text{ cm}$, RF: $5.8 \pm 4.3 \text{ cm}$). The results of the ANOVA and Tukey HSD tests confirmed significant differences in *DBH* and height among land covers (p-value < 0.001), except for EF and CP for *DBH* and RF and CP for height (Table 3).

Aboveground and deadwood biomass

Land-cover conversion from EF to RF and CP resulted in a substantial decline in both aboveground and deadwood biomass.

330 The mean *AGB_f* estimated using the generic allometric function dropped sharply from $239 \pm 92 \text{ Mg ha}^{-1}$ in EF to $42 \pm 10 \text{ Mg ha}^{-1}$ in RF and $71 \pm 22 \text{ Mg ha}^{-1}$ in CP. Similarly, the mean total *DWB* declined from $27.5 \pm 12.4 \text{ Mg ha}^{-1}$ in EF, $4.8 \pm 7.0 \text{ Mg ha}^{-1}$ in RF, and $0.4 \pm 0.2 \text{ Mg ha}^{-1}$ in CP. See Table A1 for the contribution of lying and standing *DWB* to total *DWB*.

Stem density and basal area

Changes in land cover strongly influenced stem density, basal area, and the distribution of aboveground biomass across *DBH* classes (Fig. 3). RF exhibited twice the stem density (*DBH* > 5 cm) per hectare compared to EF and CP, driven largely by a high proportion of smaller trees in the 5–15 cm *DBH* class. Despite having a lower mean *DBH*, RF had a higher basal area ($17.0 \pm 5.4 \text{ m}^2 \text{ ha}^{-1}$) than CP ($11.6 \pm 3.5 \text{ m}^2 \text{ ha}^{-1}$). Interestingly, in EF, only 5 % of the stems with a *DBH* > 30 cm contributed to approximately 65 % of the total *AGB_f*. In contrast, the main *DBH* class contributing to the *AGB_f* in RF and CP was 5–15 cm, accounting for 57 % and 76 % of the total *AGB_f* in RF and in CP, respectively. Refer to Supplementary Table S7.1 for shared stem density percentages per hectare across *DBH* classes, and Table S7.2 for shared percentages of *AGB_f* categorized by *DBH* class.

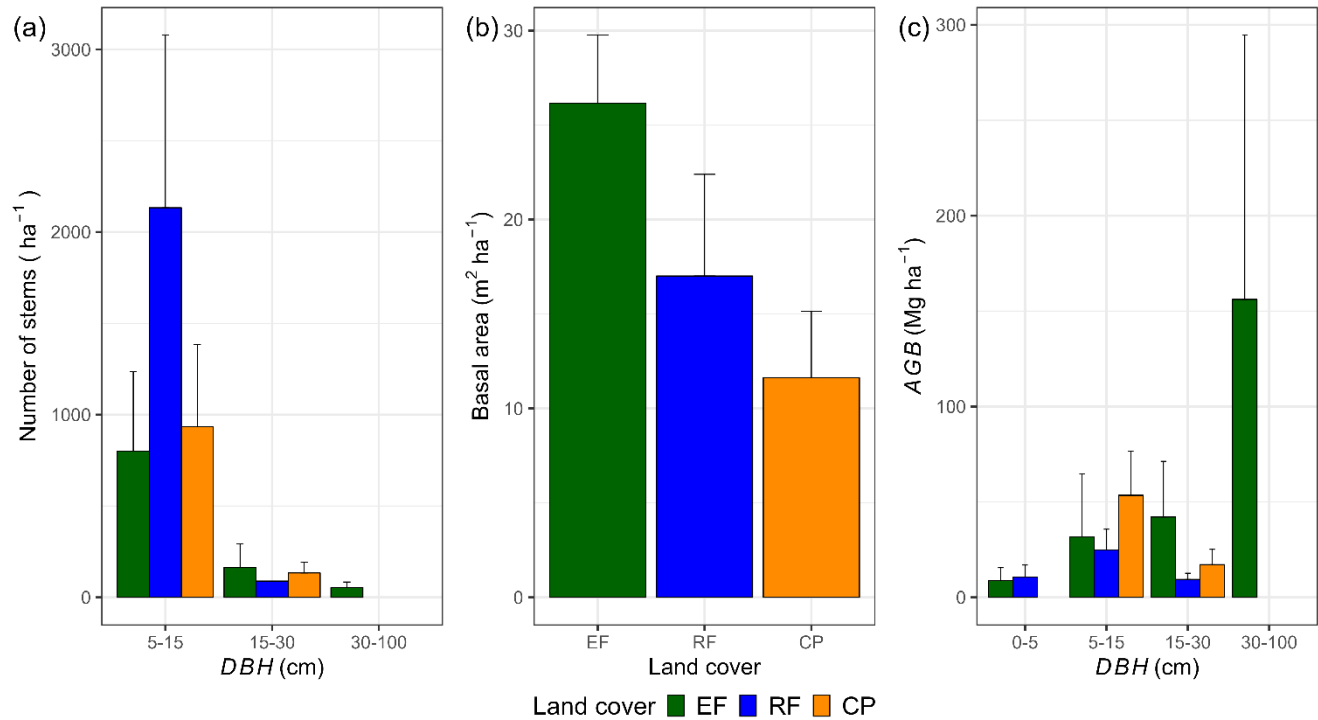


Figure 3. Estimations per land-cover class of a mean number of stems per hectare (a), basal area (*BA*, $\text{m}^2 \text{ ha}^{-1}$) (b), and mean aboveground biomass separated by the different diameters at breast height (*DBH*) classes (c). In (c), the contribution of different *DBH* classes to the mean aboveground biomass estimated by the *AGB_f* method was used in this calculation. The error bars in the figure represent one standard deviation.

LAI and *fPAR*

The mean total leaf area index values were $6.16 \pm 0.67 \text{ m}^2 \text{ m}^{-2}$ for EF, $5.57 \pm 0.76 \text{ m}^2 \text{ m}^{-2}$ for RF, and $3.07 \pm 0.61 \text{ m}^2 \text{ m}^{-2}$ for CP. The mean canopy *LAI* values were $4.62 \pm 0.5 \text{ m}^2 \text{ m}^{-2}$ for EF, $4.66 \pm 0.70 \text{ m}^2 \text{ m}^{-2}$ for RF, and $2.52 \pm 0.42 \text{ m}^2 \text{ m}^{-2}$ for CP. The ANOVA analysis revealed a significant difference in mean *LAI_T* and mean *LAI_C* among the three land-cover classes, while the Tukey HSD test did not find a significant difference in mean *LAI_T* and mean *LAI_C* between EF and RF (Table 3). The

phenology of both LAI_T and LAI_C revealed a similar pattern in EF and RF, with peak and base values in June and March, respectively (Fig. 4a–b, Table S7.3). The LAI_T and LAI_C patterns for CP resembled those of EF and RF but also had a strong decrease in April. Furthermore, the understory LAI (LAI_U , the difference between LAI_T and LAI_C) for the various land-cover classes indicates that the ground vegetation highly contributes to LAI_T for EF and RF, while the contribution was minor for CP (Fig. 4c). In particular, the LAI_U mean values within a year were approximately $1.54 \pm 0.57 \text{ m}^2 \text{ m}^{-2}$ for EF (25 %), $0.91 \pm 0.36 \text{ m}^2 \text{ m}^{-2}$ for RF (16 %), and $0.55 \pm 0.39 \text{ m}^2 \text{ m}^{-2}$ for CP (18 %). A general trend of high contribution LAI_U to LAI_T in June and low contribution in April was apparent for all land-cover classes.

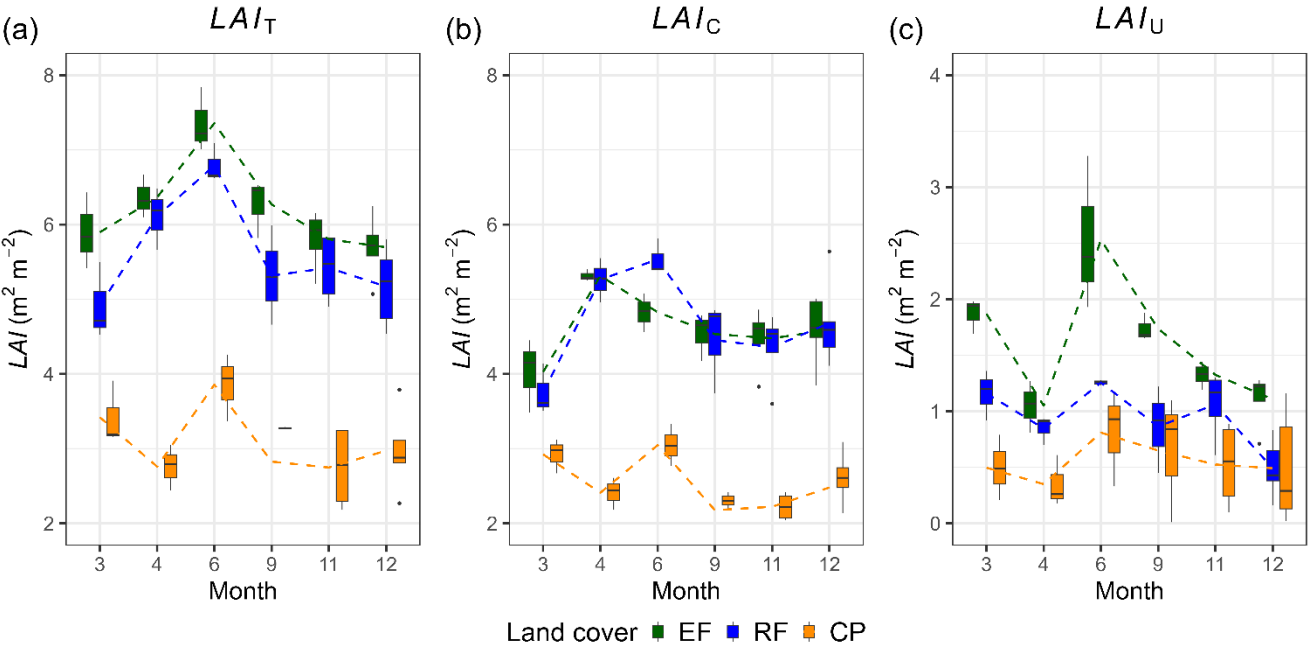


Figure 4. Total leaf area index (LAI_T , $\text{m}^2 \text{ m}^{-2}$), canopy leaf area index (LAI_C , $\text{m}^2 \text{ m}^{-2}$), and understory leaf area index (LAI_U , $\text{m}^2 \text{ m}^{-2}$), their variations across different months within a year for evergreen forests (EF), regrowth forests (RF), and cashew plantations (CP). The lines in the graph represent the connection between the mean LAI values from one month to another.

The observed mean annual $fPAR$ for EF, RF, and CP was high: 0.97 ± 0.01 , 0.96 ± 0.01 , and 0.76 ± 0.06 , respectively (Table 3). The values of EF and RF exhibited minimal fluctuations throughout the year, whereas the $fPAR$ of CP ranged between 0.55 and 0.93 (Fig. 5). Like LAI , the annual mean $fPAR$ among EF, RF, and CP were statistically significantly different according to both the ANOVA test and Tukey HSD's tests.

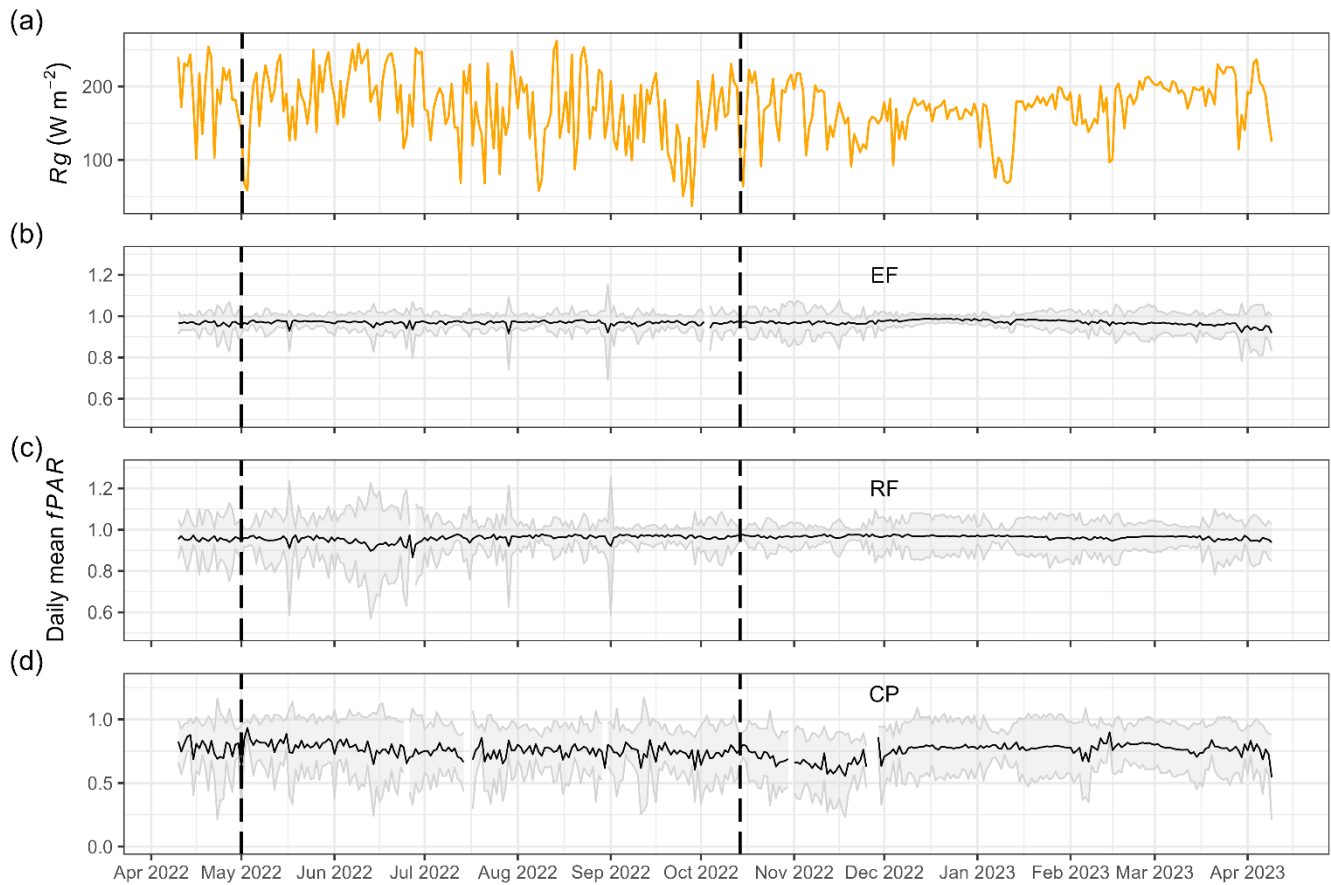
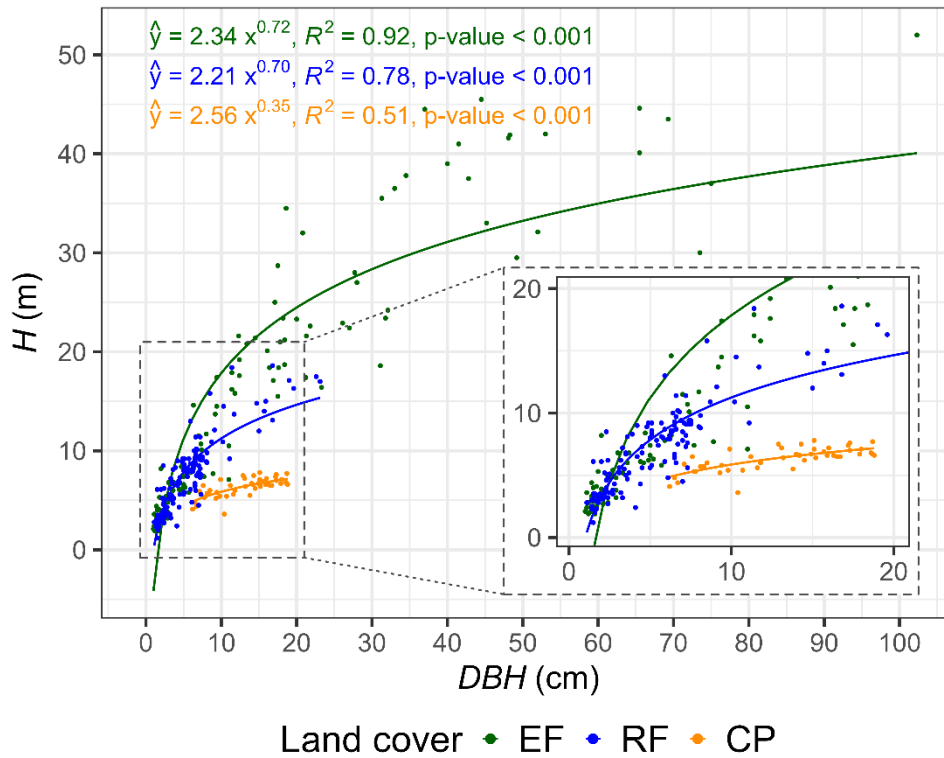


Figure 5. Daily mean global radiation (R_g , W m^{-2}) (a) and daily mean $f\text{PAR}$ for evergreen forests (EF) (b), regrowth forests (RF) (c), cashew plantations (CP) (d) from April 11, 2022, to April 9, 2023 at Kulen. The shaded area represents one standard deviation from the mean, computed using the ten PAR sensors installed in each land-cover class.

3.5 Estimated Aboveground biomass based on DBH-H relationship

DBH-H relationship

Land-cover change weakens tree allometry, reducing the consistency of DBH-H relationships in human-impact forest and agricultural ecosystems. Strong positive relationships between DBH and H were observed in both EF and RF. For EF, 92 % of the variation in H can be explained by the variation in DBH , whereas for RF and CP, it was 78 % and 51 %, respectively (Fig. 6, Table S7.4). The power-law relationships between DBH and H further indicated that the K_1 and K_2 values for EF and RF were similar, whereas the values for CP were much lower. For a plot-level analysis of relationships between $\ln(\text{DBH})$ and $\ln(H)$, see Fig. S7.2 and Table S7.5.



380 **Figure 6. Relationship between diameter at breast height (DBH) (cm) and tree height (H) (m) for evergreen forests (EF), regrowth forests (RF), and cashew plantations (CP) in Kulen. Figure shows the derived power-law intercept (K_1) and slope (K_2) values for EF, RF, and CP.**

Comparison of AGB estimation methods

Our results indicate that locally calibrated *DBH-H* relationships and species-specific wood density substantially affected
 385 aboveground biomass estimates compared to generalized models (Fig. 7). The AGB_{wd} method consistently produced higher values than AGB_f across all land-cover classes, reflecting the influence of wood density and the dominance of high-density tree species at our study site. In EF and RF, where *DBH-H* relationships were strong, AGB_h estimates were markedly higher than AGB_f (EF: 312 ± 184 vs. 239 ± 92 Mg ha⁻¹, RF: 54 ± 14 vs. 42 ± 10 Mg ha⁻¹), consistent with plot-level regression results (Fig. S7.2, Table S7.5). In contrast, in CP, AGB_h yielded much lower values than AGB_f (17 ± 5 vs. 71 ± 22 Mg ha⁻¹),
 390 highlighting the limited reliability of this method under weak *DBH-H* relationship conditions. The differences between AGB_h and AGB_f estimates across land covers are illustrated in 1:1 comparison plots and plot-level summaries (Figs. S7.3–S7.5).

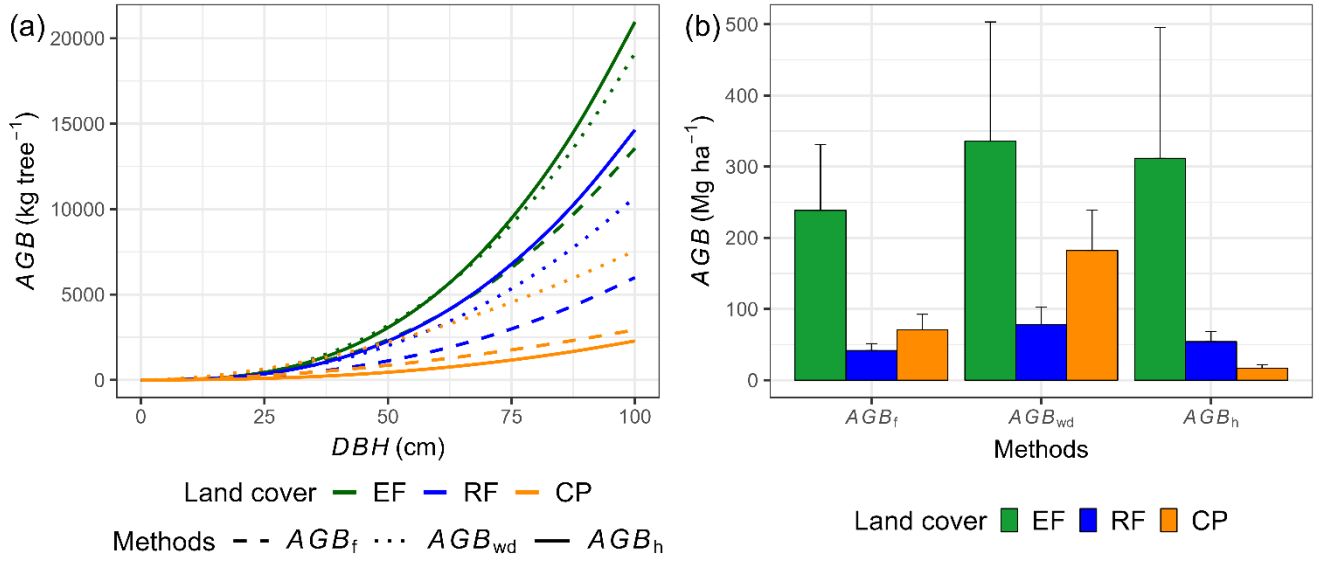


Figure 7. Power-law relationships between aboveground biomass (AGB) of AGB_f , AGB_h , and AGB_{wd} and diameter at breast height (DBH) for each land-cover class (a), along with the corresponding results of AGB estimation (b). AGB_f represents aboveground biomass estimated by adopted functions, AGB_{wd} represents aboveground biomass estimated by adopted functions utilizing species-specific wood density, and AGB_h represents aboveground biomass estimated by the DBH and tree height (H) relationship, in conjunction with species-specific wood density, for the study site. The error bar in (b) represents a standard deviation.

3.6 AGB_h relationships with LAI_T , SLA_{cwm} , and S_R

We observed positive relationships between aboveground biomass and three pivotal ecosystem characteristics: LAI_T , S_R , and SLA_{cwm} determining 76 %, 72 %, and 68 % of the variability in AGB, respectively (Fig. 8, Table S8.1 for statistical regression tables). LAI_T exhibited strong positive correlations with SLA_{cwm} , S_R , and AGB, with the Pearson correlation coefficient in the range of 0.67–0.85. SLA_{cwm} had a positive correlation with S_R and AGB. Furthermore, additional insights regarding the Pearson correlation matrix depicting relationships among various ecosystem characteristics are presented in Fig. S8.1.

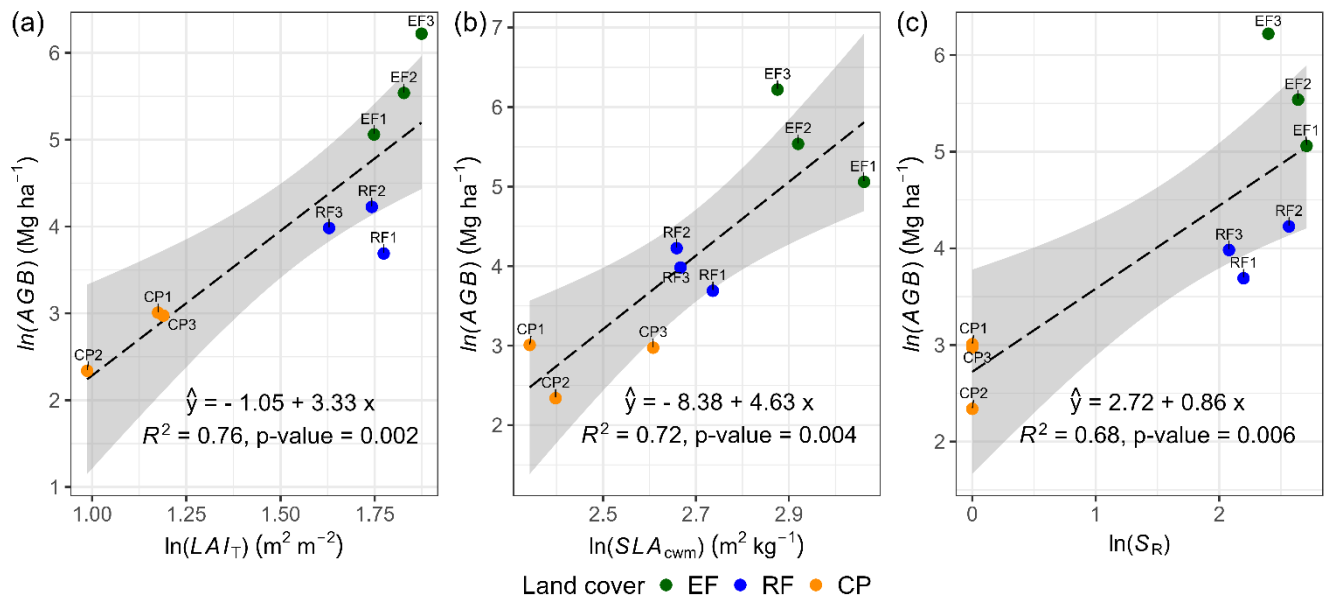


Figure 8. Ordinary least squares regression showing the effect of mean total LAI (LAI_T , m² m⁻²), mean SLA_{cwm} (m² kg⁻¹), and species richness (S_R , count per plot) on AGB . Mean LAI_T is a mean ground LAI measurement, S_R is a woody species count excluding seedlings in a plot, and AGB is AGB_h whose estimation was based on the $DBH-H$ relationship.

4 Discussions

4.1 Soil conditions

The observations of soil temperature, soil water content and soil electrical conductivity all support our hypothesis that land-cover conversion changes soil conditions in these tropical forest ecosystems. The difference in soil temperature between the forested land-cover classes (EF and RF) and the cashew plantations (Table 3) aligns with prior studies by van Haren et al. (2013) and Geng et al. (2022) and can be explained by the substantial difference in interception of incoming radiation between these ecosystems (Fig. 4). The multi-layered canopies and the dense layer of deadwood and litterfall, effectively prevent direct sunlight from reaching the ground. This natural shield reduces the impact of solar irradiance, thereby maintaining cooler soil surface temperatures (Senior et al., 2018). Conversely, CP has a simpler canopy structure, predominantly featuring a single layer of cashew trees of similar age. The understory in these areas is sparser, and the reduction in deadwood, due to management, facilitates greater penetration of solar irradiance and elevates soil temperatures.

Our observed annual mean soil water content across the three land-cover classes (0.14–0.23 m³ m⁻³) is consistent with earlier findings (Rodell et al., 2004; Wang et al., 2012; Horel et al., 2022). Variations in SWC among these classes may stem from differences in their stand structural complexity (vegetation cover and root system) and soil properties (organic matter content and texture) (Pickering et al., 2021; Tang et al., 2021). The higher SWC in evergreen forests compared to regrowth forests is attributed to their dense and multilayered vegetation cover, which reduces penetration of solar irradiance and temperature at the forest floor, thereby reducing evaporation and maintaining topsoil moisture (Fig. 4). In addition, the complex root systems

of primary forests enhance water retention by creating channels and pores in the soil, while organic matter from deadwood and litterfall further enhances soil water retention, particularly during arid conditions (Luo et al., 2023). Another explanation could be the soil texture, as our field investigation observed that cashew plantations are all on sandier soils with lower water-holding capacity, leading to decreased SWC (Ibrahim and Alghamdi, 2021). Nevertheless, further examination of soil samples is
430 necessary to accurately measure the specific soil properties in each land-cover class.

The analysis of soil electrical conductivity categorized the soils as non-saline across the land-cover classes. Evergreen forests had higher *ECs* than cashew and regrowth forests, potentially indicating larger nutrient availability (Omuto et al., 2020). This higher nutrient availability in evergreen forests may be linked to greater organic matter decomposition, species richness, higher soil moisture content, and no history of being clear-cut, which could lead to nutrient losses via run-off during the phase without
435 vegetation (Austin et al., 2004; Vestin et al., 2020; Guo et al., 2023b).

4.2 Species diversity

Species richness and Shannon-Wiener index clearly declined from evergreen forests to regrowth forests and were lowest in cashew plantations, supporting our hypothesis that land-cover conversion reduces species diversity. While EF and RF showed similar mean values to other evergreen forests in the mainland of Southeast Asia and India (Zin and Mitlöhner, 2020; Theilade
440 et al., 2022; Tynsong et al., 2022), S_R was lower compared to the most diverse rainforests in South America and equatorial Southeast Asia, where often > 250 species ha^{-1} have been reported (Mohd Nazip, 2012; ter Steege et al., 2023) and S_H was lower than for some moist evergreen and humid lowland forests in the region (Mohd Nazip, 2012; Zin and Mitlöhner, 2020). These tropical rainforests may have more species because of their larger forest patch sizes and higher rainfall, compared to the relatively isolated monsoon forest at the top of Kulen, surrounded by agricultural areas (Galanes and Thomlinson, 2009). The
445 relatively low S_H may also be explained by the high proportion of the top five dominant species in each land cover, accounting for over 50 % of total stems in their communities. Another possible reason could be the limited number of sample plots, which may not fully capture the overall species composition and distribution in these forests. Tropical tree species composition is markedly influenced by biogeography and disturbance history, showing significant local variations even over short distances (Whitmore, 1998; Van and Cochard, 2017). This emphasizes the necessity for comprehensive field data sampling to accurately
450 assess the species richness and evenness of these highly diverse plant communities. The comparison between S_R and S_H of EF and RF with previous studies is presented in Table S9.1– S9.2.

4.3 Leaf functional traits

Specific leaf area, leaf dry matter content, and chlorophyll content are all key leaf traits in the leaf economic spectrum, and carry diverse implications for understanding carbon sequestration, resource availability, successional stages, and
455 environmental responses (Wright et al., 2004; Gao et al., 2022). Our results support the hypothesis that changing from pristine evergreen forests to regrowth forests and cashew plantations leads to a substantial decline in key leaf functional traits, particularly specific leaf area and chlorophyll content, indicating reduced ecosystem productivity and resource-use efficiency.

Our observations emphasize the significant consequences of transitioning from EF to RF or CP, resulting in a substantial reduction in actual values and diversity in SLA_{cwm} , reflecting a reduction in both ecosystem productivity and resilience to disturbances (Liu et al., 2023). The higher SLA_{cwm} in EF, suggests higher photosynthetic capacity, especially in shaded environments, due to its dense canopy cover and abundant resource availability (water and nutrients) for plant growth (Green et al., 2020). High SLA_{cwm} values also link to faster turnover and promote nutrient cycles, carbon sequestration, and nutrient use efficiency in forest ecosystems (Guerrieri et al., 2021). The lower SLA_{cwm} values of RF and CP may be attributed to limited water and nutrient availability in the soil because of high competition in those ecosystems. The notable reduction in SLA_{cwm} caused by the shift from EF to RF or CP underlines the profound impact land-cover change has on ecosystem productivity, resilience, and overall functioning. Our SLA_{cwm} of EF exceeded the mean values of tropical forests in Bolivia, Brazil, Costa Rica, and China (Finegan et al., 2015; Wang et al., 2016). The SLA_{cwm} of RF was somewhat higher than the mean of neotropical regrowth forests, but still within the range (Poorter, 2021). SLA_{cwm} in CP was greater than the range value in Parakou, Benin, but fell within the range reported for 15 cashew varieties in Karnataka, India (Akossou et al., 2016; Mog and Nayak, 2018).

Chlorophyll is essential for photosynthesis and serves as a crucial indicator of a plant's photosynthetic capacity, profoundly influencing overall growth (Stirbet et al., 2020). The elevated Chl_{cwm} seen in EF can be attributed to the well-developed and dense canopy structure, which creates a light-shaded environment. This prompts plants to invest more in chlorophyll production, enhancing light harvesting efficiency (Niinemets, 2010). Meanwhile, RF, experiencing intense competition for light in early successional stages, may exhibit lower chlorophyll levels as resources prioritize vertical growth over chlorophyll production (Laurans et al., 2014). Our CP had lower Chl_{cwm} than EF and RF due to less light competition and higher temperatures, which could lead to photoinhibition and lowered leaf chlorophyll content (Rosa et al., 2020). Our Chl_{cwm} of EF and RF falls within the range, but surpasses the mean Chl_{cwm} observed in Chinese forest ecosystems (Li et al., 2018).

Leaf dry matter content is a measure of construction cost per fresh weight mass unit, and it serves as a metric for a plant's resource use strategy and resilience to environmental stresses (Guo et al., 2023a). The higher $LDMC_{cwm}$ in EF compared to that of RF indicates a conservative resource usage, longer leaf lifespan, and increased carbon sequestration, implying higher ecosystem stability and function for EF (Rawat et al., 2021). Conversely, the highest $LDMC_{cwm}$ in CP, is attributed to cashew monoculture and the species' high resilience to environmental stress, especially in nutrient-poor soils and water-stressed conditions (Bezerra et al., 2007). This study emphasises EF's increased stress tolerance, conservative resource utilisation and greater carbon sequestration compared to RF, while also emphasizing cashew as a highly proficient species in environmental stress tolerance.

4.4 Stand structure attributes

DBH and tree height

Our findings confirm significant differences in mean *DBH* and tree height resulting from the conversion of pristine evergreen forests to young regrowth forests and cashew plantations following human disturbance. The observed reduction in large-

490 diameter and tall trees in regrowth forests and cashew plantations compared to the evergreen forests (Fig. 3) provides clear evidence of structural degradation, which negatively affects crucial key ecosystem functions such as carbon storage, nutrient cycling, and biodiversity (Díaz et al., 2007; Lutz et al., 2018; Thiel et al., 2021). Observed species in our evergreen forests, such as *Dipterocarpus costatus*, *Sandoricum indicum*, *Mesua ferrea*, *Nageia wallichiana*, and *Litchi chinensis* reach heights of 40–52 m, similar to those found in Cambodia's central evergreen forests (Theilade et al., 2022). Our mean *DBH* of evergreen
495 forests is comparable to mature tropical forests in Vietnam and falls within the pantropical range, while regrowth forests have a slightly higher mean *DBH* than tropical secondary forests in Sarawak, Malaysia (Brown, 1997; Kenzo et al., 2009; Yen and Cochard, 2017). In contrast, cashew plantations show a significantly lower mean *DBH* compared to older counterparts in Kampong Cham, Cambodia (Avtar et al., 2013).

Aboveground and deadwood biomass

500 Our results support that land-cover conversion reduces in aboveground and deadwood biomass in regrowth forests and cashew plantations compared to evergreen forests. The substantial decline in aboveground biomass following conversion from EF to RF or CP is primarily driven by historical human disturbance, particularly clear-cutting and the removal of large trees, as evidenced by reduced *DBH* and tree height in this study. Similarly, *DWB* decreased as EF were replaced by RF and CP, reflecting the impacts of land-cover change on forest biodiversity and ecosystem health. *DWB* is a key indicator of biodiversity
505 and ecosystem health, supporting various species and ecosystem processes like carbon and nitrogen cycling, soil fertility enhancement, pollination, and erosion control (Parisi et al., 2018a; Santopuoli et al., 2021; Tláškal et al., 2021). Variations in total *DWB* values could result from the degree of disturbances within the studied forests (Baker et al., 2007). The higher *DWB* in EF is due to its old stand age, long-term accumulation of *DWB*, and absence of slash-and-burn practice as observed in RF and CP (van Galen et al., 2019). In CP, some farmers periodically cut and burn dead branches of cashew trees to promote
510 growth. Consistent with these trends, *DWB* in our EF and RF was comparable to previous studies in Cambodia and Malaysia (Saner et al., 2012; Kiyono et al., 2018), whereas our CP has less *DWB* than plantations in Cameroon (Victor et al., 2021).

Stem density and basal area

The land-cover change alters stem density across EF, RF, and CP. Our mean stem density per hectare of evergreen forests is consistent with previous studies in Cambodia, Vietnam, and in Borneo, while regrowth forests show lower densities compared
515 to those in the Yucatan Peninsula, Mexico (Slik et al., 2010; Con et al., 2013; Román-Dañobeytia et al., 2014; Chheng et al., 2016; Theilade et al., 2022). Additionally, our stem density in cashew plantations is similar to that of Isuochi, Nigeria, but significantly greater than that of Casamance, Senegal, due to their differences in planting distance and management practices (Nzegbule et al., 2013; Ndiaye et al., 2020). The variation in stem density between evergreen forests and regrowth forests reflects distinctive stages of succession. In the early succession stage following clearance, open niches and resource abundance
520 create a favourable environment for fast-growing and highly reproductive early-succession species, resulting in higher stem density and heightened interspecies competition (Zhang et al., 2020). As the forest matures, stem density naturally decreases

as larger trees occupy more space, and take more of the light, water, and nutrient resources. This competition ultimately leads to the mortality of smaller trees, aligning with the power-law relationship between stem density and biomass commonly observed in mature forests (Mrad et al., 2020). This natural process also alters species composition, stand structure, habitat
525 heterogeneity, and biomass of forests (Forrester et al., 2021). In cashew plantations, stem density is controlled by humans to enhance cashew yield. This alteration in stand structure complexity influences interspecies competition. These modifications also affect stand structure and interspecies competition, ultimately influencing the biodiversity and functioning of the ecosystem.

Also basal area decreases significantly when EF is replaced with RF or CP, impacting biomass, productivity, stand structure,
530 and structural complexity (Gea-Izquierdo and Sánchez-González, 2022). RF has lower *BA* than EF, indicating early succession and disturbance (Ziegler, 2000). Despite tropical forests possess natural regenerative capabilities, RF may require several decades to achieve *BA* levels comparable to EF, highlighting the critical importance of conserving EF to maintain their ecological integrity and ecosystem services. In addition, the basal area of evergreen forests in our study aligns with those in northeast Cambodia and Pahang National Park, Malaysia, but falls below values reported for Laos, Cambodia's central plains
535 and Vietnam's lowlands (Rundel, 1999; Sovu et al., 2009; Mohd Nazip, 2012; Chheng et al., 2016; Theilade et al., 2022). Our regrowth forest's *BA* exceeds that of regrowth forest in Laos, while cashew plantations surpass plantations in Tanzania's (Sovu et al., 2009; Malimbwi et al., 2016).

LAI and fPAR

LAI and *fPAR* differed markedly among the three land-cover classes, reflecting structural changes in canopy structure
540 associated with land-cover conversion and consistent with our hypothesis. In our study, canopy leaf area index in evergreen forests surpasses that of dry evergreen forests in Kampong Thom, Cambodia, while regrowth forests lie between those of 18–35-year tropical secondary forests in Costa Rica; however, cashew plantations exceed reported values in India (Ito et al., 2007; Clark et al., 2021; Kumaresh et al., 2023). The *LAI_C* difference between the forests (EF and RF) and CP was significant due to CP management practices, resulting in a thin canopy with low *LAI_C*. In contrast, natural forests with their densely developed
545 canopy have a high *LAI_C*. Additionally, *LAI_C* phenology followed the rainy and dry seasons, with peak values during the rainy season and low values during the dry season (Ito et al., 2007). During the dry season, reduced rainfall leads to less water availability for plant growth, causing plants to adapt to water stress by shedding their leaves, resulting in low *LAI_C* in the ecosystem (Maréchaux et al., 2018). The comparison between *LAI_C* and *LAI_T* of EF and RF with previous studies is presented in Table S9.3.

550 Our mean fraction of photosynthetically active radiation for EF and RF marginally exceeded the global range for broadleaf forests and the monthly range observed in the Amazon tropical forest in Santarém, Brazil (Senna et al., 2005; Pastorello et al., 2020). The *fPAR* for CP, on the other hand, is within the range values reported for broadleaf crops (Xiao et al., 2015). Despite annual variations in *LAI_C* (24 % for EF, 32 % for RF, 29 % for CP) and incoming solar irradiance, *fPAR* remained remarkably stable throughout the year in the forest ecosystems (EF and RF, Fig. 5). This stability can be attributed to the exponential

555 relationship between $fPAR$ and LAI , which typically saturates at LAI above 3 (Dawson et al., 2003). Our recorded lowest LAI for EF and RF was 3.48, likely contributing to this saturation and explaining the lack of phenology displayed in $fPAR$. The exclusion of reflected PAR above the canopy in the $fPAR$ estimation may also contribute to the stability; however, previous studies have shown that the difference between intercepted (what we measured) and absorbed PAR (including the reflected component) is minimal (Olofsson and Eklundh, 2007).

560 Despite an increased stem density in RF compared to EF, similar canopy LAI and $fPAR$, our observations of significant differences in mean DBH , tree height and basal area, the reduction in aboveground and deadwood biomass, altered stem density, and reduction in contribution of understory LAI to total LAI , supports our hypothesis that the land-cover change causes a decreased complexity in stand structure.

4.5 Estimated Aboveground biomass based on $DBH-H$ relationship

565 **$DBH-H$ relationship**

Variation in the strength of $DBH-H$ relationships reflects different disturbance histories in EF and RF, and the influence of management practices in CP. The $DBH-H$ relationship is crucial for understanding variations in tree growth rates, successional stage, aboveground biomass, and forest health (Kramer et al., 2023). Finding a strong positive $DBH-H$ relationship may indicate disturbances within the ecosystem, as these by initiating gaps in the canopy provide opportunities for fast-growing species to establish and utilize increased light availability and resources within the ecosystem (Senf et al., 2020). Hence, the observed relationships between EF and RF suggest a composition of fast-growing species and indicate that EF may have experienced past disturbances. Indeed, a windthrow in EF1 is reflected in its lowest LAI_C among EF plots and a smaller mean DBH (Table 1, Fig. S7.1a).

The lower $DBH-H$ relationship in cashew plantations results from the growth strategy of the single species and management practices. In monocultures with uniformly aged cashew plants, competition for light and resources is comparable, resulting in a consistent resource distribution. Cashew's natural growth characteristics, with the species reaching up to 15 m in height and a DBH of 100 cm under favourable conditions (Avtar et al., 2014), indicate a preference for investing resources in branches and stems over height, especially in low-light competition environments. However, our observations indicate significant variation in the $DBH-H$ relationship among CP plots (low R^2 value in Fig. 6, Fig. S7.2g–i) which may have been influenced by their different management practices, such as spacing, pruning, and thinning. These practices impact the $DBH-H$ relationship by minimizing light competition, resulting in a higher $DBH-H$ ratio which also affects the relationship (Deng et al., 2019; Bhandari et al., 2021).

Comparison of AGB estimation methods

Our results suggest that locally calibrated $DBH-H$ relationships and wood density substantially affect AGB estimates compared to generalized models, supporting the feasibility of site-specific calibration, particularly for natural forest ecosystems. Recent

studies have emphasized the significant uncertainty in estimating plot-level aboveground biomass when directly applying a generic AGB allometric equation (AGB_f) due to variations in species composition and stand structure between the study site and the equation's origin (Feldpausch et al., 2011; Burt et al., 2020). To address this challenge, our study proposes an allometric approach (AGB_h) using local species-specific wood density and the $DBH-H$ relationship at the study site. This approach captures the unique characteristics of the site's species composition and stand structure (Ketterings et al., 2001a; Nyirambangutse et al., 2017). Our locally adopted AGB_h method produced estimates $\sim 30\%$ higher than the generic AGB_f for both EF and RF (Table 3, Fig. 7b). This is likely due to the combined effects of higher mean wood density and a stronger DBH relationship, resulting in a more pronounced exponential growth response in AGB (Fig. 7a). Still, these $\sim 30\%$ higher values align with the range reported in previous studies (Tables A2–A3). In contrast, in the CP case, our AGB_h method produced estimates less than a quarter of the generic AGB_f method. The reason is that the AGB_h method is less reliable when a weak $DBH-H$ relationship is detected because it fails to accurately capture the overall tree size and volume. This is also reflected in the substantially larger uncertainty as indicated by the standardized errors of the parameters within the $DBH-H$ relationship (Table A4, Table S7.4). The substantial difference between AGB_{wd} and AGB_f is primarily due to the wood density values used: 0.45 g cm^{-3} from Zanne et al. (2009) in this study versus 0.18 g cm^{-3} in the original AGB_f equation (Mlagalila, 2016), likely reflecting variation in cashew wood properties or wood density measurement protocols among the two studies. Despite clear differences in Fig. 7b, formal statistical comparisons were not conducted due to the limited number of plots per class ($n = 3$), which restricts statistical power. However, to fully validate the AGB allometric equations, destructive field-observed data would be necessary. Therefore, future research should include direct field measurements of AGB to more accurately validate the methods for these land-cover classes.

4.6 AGB_h relationships with LAI_T , SLA_{cwm} and S_R

Exploring the relationship between aboveground biomass and key ecosystem characteristics such as leaf area index, specific leaf area, and species richness is vital for comprehending the complexity of ecosystem dynamics and informing ecosystem modelling. We observed a strong positive relationship between LAI_T and AGB_h , supporting prior findings (He et al., 2021; Zhao et al., 2021). Higher LAI_T enhances light interception and results in higher biomass. Elevated AGB_h levels stimulate LAI_T expansion by providing resources for robust leaf growth, leading to a denser canopy and greater leaf coverage. Similarly, our findings support a positive relationship between SLA_{cwm} and AGB_h (Finegan et al., 2015; Ali et al., 2017; Gao et al., 2021). Higher SLA_{cwm} values indicate a plant community with improved photosynthetic capacity, nutrient uptake, and leaf turnover, which is essential for nutrient cycling (Reich et al., 1991). An increase in AGB_h has a reinforced effect on SLA_{cwm} values, suggesting enrichment of the soil nutrient pool and providing structural support for plant growth. This influences light availability and competition dynamics, affecting leaf morphology and SLA_{cwm} . Furthermore, the positive relationship between AGB_h and S_R is widely observed and explained by the niche complementarity hypothesis (Waide et al., 1999; Jactel et al., 2018; Steur et al., 2022). This concept suggests that an ecosystem with high species diversity has a greater variation in functional traits and resource-use strategies, lowering competition for scarce resources, and thus promoting productivity.

(Tilman et al., 1997). In return, an increase in AGB_h fosters the coexistence of diverse species by providing more available
620 resources and habitat complexity in an ecosystem, thereby increasing species richness.

5 Conclusions

In response to growing concerns over the ecological impacts of forest conversion in tropical Southeast Asia, we investigated how land-cover change from pristine evergreen forests to regrowth forests and cashew plantations alters stand structure, species diversity, leaf functional traits, and soil conditions, and evaluated the feasibility of locally calibrated $DBH-H$ allometries for
625 improving aboveground biomass estimation. Our findings confirm our hypotheses that land-cover change reduces stand structural complexity, species composition, and leaf functional traits, and causes a substantial change in soil conditions. We further demonstrate the utility of our novel dataset for improving aboveground biomass estimation through the application of an allometric function based on locally specific WD and the $DBH-H$ relationship. This approach has great potential for improving carbon stock estimations and promoting informed forest management practices. However, as we lack direct
630 destructive samples of aboveground biomass, we can neither reject nor support our second hypothesis that locally calibrated $DBH-H$ relationships would substantially improve aboveground biomass estimates compared to generalized models. Moreover, our analysis of relationships between leaf area index, specific leaf area, species richness, and aboveground biomass, underlines land-cover change’s profound impact on ecosystem productivity and functioning in these tropical forest regions. To strengthen and extend these findings, future studies should incorporate destructive sampling to validate our locally
635 calibrated aboveground biomass allometric equations based on $DBH-H$ relationships and WD . Expanding field data collection by increasing the number and spatial distribution of plots across a broader range of land-use classes in tropical Southeast Asia and promoting open data sharing will be critical for improving our understanding of ecosystem responses to forest conversion and supporting sustainable forest management under global change in the region.

Appendix A

640 **Table A1. Estimated lying deadwood biomass ($Mg\ ha^{-1}$), standing deadwood biomass ($Mg\ ha^{-1}$), and total deadwood biomass ($Mg\ ha^{-1}$) by different land-cover classes in Kulen. Mean $\pm$ SD is a mean plus or minus a standard deviation.**

Land cover	Lying deadwood biomass ($Mg\ ha^{-1}$)		Standing deadwood biomass ($Mg\ ha^{-1}$)		Total deadwood biomass ($Mg\ ha^{-1}$)	
	Mean $\pm$ SD	Range	Mean $\pm$ SD	Range	Mean $\pm$ SD	Range
EF (n = 3)	17.74 $\pm$ 19.93	1.64–40.03	9.74 $\pm$ 8.49	0–15.56	27.48 $\pm$ 12.37	15.31–40.03
RF (n = 3)	3.65 $\pm$ 5.32	0.48–9.79	1.16 $\pm$ 1.66	0–3.06	4.81 $\pm$ 6.97	0.48–12.85
CP (n = 3)	0.40 $\pm$ 0.19	0.28–0.62	0	0	0.40 $\pm$ 0.19	0.28–0.62

Table A2. Comparing estimated aboveground biomass (AGB, Mg ha⁻¹) in evergreen forests (EF) using adopted allometric equations (AGB_f), diameter at breast height (DBH) and tree height (H) power-law relationship (AGB_h), and previous AGB reported in previous studies. Mean ± SD is a mean plus or minus a standard deviation.

No.	Region	Vegetation type	AGB (Mg ha ⁻¹)		References
			Mean ± SD	Range	
1	Kulen, Cambodia	Tropical evergreen forest	311.66 ± 183.88	147.53–510.57	AGB _h in this study
2	Kulen, Cambodia	Tropical evergreen forest	238.53 ± 92.41	161.83–341.13	AGB _f in this study
3	Global	Tropical forest	379.02 ± 187.40	230.58–589.58	Chave et al. (2014)
4	Gia Lai, Vietnam	Tropical evergreen forest	273.24 ± 112.22	189.53–400.76	Nam et al. (2016)
5	Mondulkiri, Cambodia	Tropical moist evergreen forest	333.00 ± 137.00	78.00–837.00	Sola et al., (2014)
6	Borneo (Brunei, Malaysia, Indonesia)	Tropical lowland evergreen forest	458.16 ± 123.62	196.30–778.50	Slik et al. (2010))
7	Thanh Hoa, Vietnam	Tropical evergreen broadleaf forest	251.81 ± 125.43	40.88–543.88	Nguyen and Kappas (2020)
8	Africa	Tropical evergreen forest	429.00	114.00–749.00	Lewis et al. (2013)
9	Cambodia	Evergreen forest	243.00 ± 128.00	11.00–837.00	Sola et al., (2014)
10	Kampong Thom, Cambodia	Evergreen forest	294.00 ± 65.00	176.00–398.00	Ota et al. (2015)
11	Vietnam	Tropical evergreen broadleaf forests in various ecoregions	230.10 ± 8.60	199.00–320.20	Van Do et al. (2019)

Table A3. Comparing estimated aboveground biomass (AGB, Mg ha⁻¹) in regrowth forests (RF) using adopted allometric equations (AGB_f), diameter at breast height (DBH) and tree height (H) power-law relationship (AGB_h), and previous AGB reported in previous studies. Mean ± SD is a mean plus or minus a standard deviation.

No.	Region	Vegetation type	AGB (Mg ha ⁻¹)		References
			Mean ± SD	Range	
1	Kulen, Cambodia	Natural regrowth evergreen forest	54.19 ± 14.09	38.26–65.04	AGB _h in this study
2	Kulen, Cambodia	Natural regrowth evergreen forest	41.66 ± 9.82	31.60–51.21	AGB _f in this study
3	Sumatra, Indonesia	Mixed secondary forest	59.04 ± 17.15	39.26–69.79	Ketterings et al. (2001a)
4	Kampong Thom, Cambodia	Regrowth forest	42.00 ± 21.00	22.00–90.00	Ota et al. (2015)
5	Malaysia	Young forests aged 8.5–17 years	63.60 ± 34.93	34.00–118.00	Kho and Jepsen (2015)

Table A4. Comparing estimated aboveground biomass (AGB, Mg ha⁻¹) in cashew plantations (CP) using adopted allometric equations (AGB_f), diameter at breast height (DBH) and tree height (H) power-law relationship (AGB_h), and previous AGB reported in previous studies. Mean ± SD is a mean plus or minus a standard deviation.

No.	Region	Vegetation type	AGB (Mg ha ⁻¹)		References
			Mean ± SD	Range	
1	Kulen, Cambodia	Family-scale cashew plantation	16.70 ± 4.80	11.23–20.23	AGB _h in this study
2	Kulen, Cambodia	Family-scale cashew plantation	70.60 ± 22.01	46.16–88.87	AGB _f in this study
3	Benin	Cashew agroforestry farming	18.07 ± 2.14	-	Biah et al. (2019)
4	Guinean, Cote d'Ivoire	Cashew plantation	13.78 ± 0.98	-	Kanmegne Tamga et al. (2022)
5	Kampong Cham, Cambodia	Large-scale and intensively managed cashew plantation (10–16 years of age)	104.30 ± 19.65	72.00–143.00	Avtar et al. (2013)

Data availability

- 655
- All the collected data used in this study are publicly available via the links as follows:
1. The datasets of the forest inventory, leaf area index, and leaf functional traits across various land-cover classes are available at <https://doi.org/10.5281/zenodo.10146582> (Sovann et al., 2024a).
 2. The daily data, including *f*PAR, soil conditions, and meteorological conditions from April 10, 2022, to April 9, 2023, can be downloaded from <https://doi.org/10.5281/zenodo.10159726> (Sovann et al., 2024b).
 3. Future data from the field site will be uploaded to https://zenodo.org/communities/cambodia_ecosystem_data on a regular basis.
- 660

Author contributions

CS led field data collection, analysis, and manuscript writing. TT and SO contributed to conceptualization, manuscript review, editing, and supervision. SK and SS provided administrative support and supervised fieldwork in Cambodia. PV offered

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technical guidance and support in equipment installation and maintenance. SB managed field data collection. All authors contributed to editing the manuscript.

Competing interests

The authors declare no conflict of interest.

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