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1 **Amazon palm biomass and allometry**

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17

18 **Abstract**

19 Palms (family Arecaceae) are abundant in Amazonian forests, but the allometry of these  
20 monocotyledonous plants remains poorly quantified. Woody palm biomass is most commonly  
21 estimated with dicotyledonous tree models, which leaves substantial uncertainty as to their true  
22 biomass and productivity. We developed the first extensive dataset of directly-measured  
23 arborescent palm biomass: 136 individuals from nine species in *terra firme* and wetland forests  
24 —*Astrocaryum murumuru*, *Attalea phalerata*, *Bactris gasipaes*, *Euterpe precatoria*, *Iriartea*  
25 *deltoidea*, *Mauritia flexuosa*, *Mauritiella aculeata*, *Oenocarpus bataua*, and *Socratea exorrhiza*.  
26 We created single species ( $n = 8\text{--}21$ ) and family-level ( $n = 97\text{--}106$ ) allometric equations, using  
27 diameter, stem height, total height, and stem dry mass fraction, to estimate (i) total aboveground  
28 biomass for all species, (ii) belowground biomass for the two wetland species (*Mauritia* and  
29 *Mauritiella*), and (iii) leaf mass for all species. These new palm models were then applied to nine  
30 1-ha plots in the southwestern Amazon (Tambopata) to calculate the impact on forest biomass  
31 estimates once palm mass is estimated with palm-specific models, rather than from models  
32 created for dicot trees. We found that stem height was the best predictor variable for arborescent  
33 palm biomass, but the relationship between stem height and biomass differed among species.  
34 Most species showed weak biomass–diameter relationships, but a significant relationship could  
35 be identified across all species. The new palm models were better estimators of palm mass than  
36 existing dicot models. Using our species-level models increased estimates of palm biomass at our  
37 study site by 14–27 %, compared to using recently published pantropical biomass models for  
38 trees. In other forests, the effect of using these palm equations on biomass estimates will depend  
39 on palm sizes, abundance, and species composition.

40

41 **Key words:** Arecaceae, equations, *Iriartea*, *Mauritia*, leaf mass, productivity, tropical forest

## 42 1 Introduction

43 Palms (family Arecaceae or Palmae) are an ancient part of tropical ecosystems (Bremer et  
44 al., 2004) and one of the most widespread and ecologically diverse plant families (Tomlinson,  
45 2006; Eiserhardt et al., 2011). They play major roles in ecosystem processes (Peters et al., 2004;  
46 LaFrankie and Saw, 2005) and local livelihoods (May et al., 1985; Johnson, 1996; Runk, 1998).  
47 Arecaceae is one of the most heavily used plant families for non-timber forest products with  
48 multiple applications in indigenous and rural activities, mostly associated with food, fibres,  
49 animal fodder, and construction (Peters et al., 1989; Phillips and Gentry, 1993; Johnson, 1996;  
50 Zambrana et al., 2007).

51 Nearly 2,400 species of palms occur across the Neotropics, Africa, and Asia (Govaerts and  
52 Dransfield, 2005). Within the Neotropics, palms are most abundant in western Amazonia and  
53 Central America (Kahn et al., 1988; Terborgh and Andresen, 1998; Montufar and Pintaud, 2006;  
54 Eiserhardt et al., 2011). They are less prevalent in other regions but still occur, especially in  
55 forests with frequent inundation (Kahn et al., 1988; Terborgh and Andresen, 1998) and soils with  
56 poor physical properties, such as shallow rooting depth (Emilio et al., 2013). Arecaceae is the  
57 single most abundant arborescent plant family in western Amazonian forests, in both *terra firme*  
58 and flooded forests (Terborgh and Andresen, 1998). In some forests, palms have been found to  
59 represent over two-thirds of stems with diameter ( $D$ )  $\geq 10$  cm (Terborgh and Andresen, 1998) or  
60 nearly 100 % of stand biomass (Brown, 1997). Indeed, some species, such as *Mauritia flexuosa*,  
61 can establish nearly mono-dominant stands ('aguajales'; Kahn and Mejia, 1990) and are an  
62 integral part of many carbon-rich swamp ecosystems (Lahteenoja et al., 2009).

63 Despite their importance, there are no explicit studies of carbon stocks and dynamics of  
64 palms. While many models have been developed to estimate the biomass of dicotyledonous  
65 (dicot) trees (e.g., Brown et al., 1989; Baker et al., 2004; Chave et al., 2005; Basuki et al., 2009;  
66 Alvarez et al., 2012; Feldpausch et al., 2012), there are few available to estimate palm biomass.  
67 Thus, most stand-level and macro-ecological studies use dicot models to estimate palm mass  
68 (e.g., Malhi et al., 2004; Phillips et al., 2008; Baccini et al., 2012) or stem basal area to assess  
69 aboveground biomass (AGB) changes (e.g., Lewis et al., 2004; Malhi et al., 2004). The  
70 productivity of palms has also been poorly studied, and palms have even been described as a  
71 'missing term' in coarse woody productivity assessments (Malhi et al., 2009). For example, palm  
72 leaves are often not included in litterfall assessments even though they may contribute

73 substantially, and palm fruit productivity may be severely underestimated because they do not fit  
74 into standard leaf litter traps (Chave et al., 2010). Overall, these factors lead to substantial  
75 uncertainty in AGB stocks and productivity in areas where palms are prevalent.

76 Estimates of palm biomass and stem productivity made from dicot models are likely to be  
77 inaccurate, especially when using  $D$  and wood density ( $\rho$ ), because the two groups of plants have  
78 very distinct growth patterns and internal properties. Palms are monocotyledons which grow in  
79 height but lack secondary (diameter) growth (Rich et al., 1986; Tomlinson, 2006). Thus, many  
80 species have weak or no relationship between height and diameter (Rich et al., 1986). The  
81 internal stem structure is also very different in palms (Parthasarathy and Klotz, 1976), with  
82 higher density and stiffness towards the peripheries and base of the stem (Rich, 1987b), and  
83 ‘wood’ density in Arecaceae is generally lower than in dicot families (Chave et al., 2009; Zanne  
84 et al., 2009). Furthermore, palms lack branches, which can contribute substantially to dicot  
85 biomass (Goodman et al., in press). This suggests that palm biomass may be overestimated by  
86 dicot equations, but this has yet to be tested on directly-measured palm biomass data.

87 The lack of palm biomass equations is puzzling: there are still no broadly accepted or  
88 applicable equations to estimate their mass. The few existing palm models are created for a  
89 single species and often do not cover a wide range of sizes. Most palm models appear in  
90 technical reports or other unpublished works (Hughes, 1997; Delaney et al., 1999; Brown et al.,  
91 2001; Freitas et al., 2006; Sierra et al., 2007; Kumar and Russell, unpublished, cited in Kumar,  
92 2011). The only three peer-reviewed publications we could locate were each developed for a  
93 single species in a particular environment: *Prestoea montana* in sub-montane Puerto Rico  
94 (Frangi and Lugo, 1985), oil palm (*Elaeis guineensis*) under commercial cultivation (Thenkabail  
95 et al., 2004), and *Oenocarpus bataua* in a transition zone from lowland to premontane forests in  
96 Colombia (Sierra et al., 2007). One mixed-species model has been developed but only for very  
97 small individuals,  $1 \leq \text{height} \leq 1.5$  m (Sierra et al., 2007). There is clearly a strong need to  
98 develop more widely applicable equations to estimate the biomass and productivity of this  
99 prevalent and important plant group.

100 Similarly, there has been no rigorous examination of the most appropriate form of palm  
101 allometric relationships. Most models are built with the simple form: biomass =  $a + bx$  (Frangi  
102 and Lugo, 1985; Thenkabail et al., 2004; Kumar and Russell, unpublished, cited in Kumar,  
103 2011), but plant allometric relationships do not usually follow this simple linear relationship

104 (e.g., West et al., 1997; Chave et al., 2005). Stem height is the most commonly used predictor  
105 variable ( $x$ ) for palm mass (Frangi and Lugo, 1985; Delaney et al., 1999; Brown et al., 2001;  
106 Thenkabail et al., 2004; Sierra et al., 2007), but others have used total height (Frangi and Lugo,  
107 1985), diameter (Hughes, 1997), or age (Kumar and Russell, unpublished, cited in Kumar, 2011)  
108 to estimate palm biomass. Brown (1997) suggested that palm biomass could be estimated using  
109 height and  $D$  as if palms were cylinders (i.e.,  $D^2H$ ), multiplied by wood density, and added to a  
110 term accounting for leaves, but this approach has yet to be applied. Estimating AGB with  
111 compound variable  $\rho D^2H$  and a form factor to account for stem taper is common for dicots  
112 (Chave et al., 2005; Feldpausch et al., 2012) but may be particularly appropriate for palm  
113 allometry because they lack branches. Non-linear relationships between biomass and the  
114 predictor variable(s) —such as with a power-law, as has been suggested on theoretical grounds  
115 (West et al., 1997)— have also not been comprehensively tested.

116 Because palms exhibit primary (height) growth nearly independently of diameter and stems  
117 taper little, we expect that that (i) height will be highly predictive of palm biomass, and (ii) palm  
118 mass should be approximately proportional to its volume calculated as a cylindrical form with  $D^2$   
119 and total or stem height. Furthermore, because palm ‘wood’ density can vary 10-fold between  
120 species and even within individuals (Rich, 1987b), we expect that a variable accounting for  
121 density or moisture content will be necessary to include in mixed-species models. Our specific  
122 objectives are to (i) create single species and family-level models for arborescent palms using a  
123 variety of simple and compound predictor variables and model forms and then (ii) examine the  
124 impact of applying new palm models on forest biomass estimates in a well-studied western  
125 Amazonian site where arborescent palms are common.

## 126 **2 Materials and methods**

### 127 **2.1 Species selection and study area**

128 Species or genera were selected to include the six most dominant arborescent palm species  
129 in the Amazon —*Iriartea deltoidea*, *Attalea butyracea*, *Oenocarpus bataua*, *Euterpe precatoria*,  
130 *Socratea exorrhiza*, and *Astrocaryum murumuru* (Emilio et al., 2013)— and two prominent  
131 species in wetland forests, *Mauritia flexuosa* and *Mauritiella* spp. (Kahn, 1991; Roucoux et al.,  
132 2013). We focus on arborescent palms because these are included in most forest inventories ( $D \geq$   
133 10 cm).

134 Palms were harvested from mature forests in western Amazonia. In 2006, *Mauritia*  
135 *flexuosa* and *Mauritiella aculeata* were harvested and weighed in wetlands within the Pacaya–  
136 Samiria National Reserve in Loreto, Peru. In 2011, biomass data were collected from seven  
137 species in *terra firme*, moist tropical forests within a forestry concession in Madre de Dios, Peru  
138 (Table 1).

## 139 **2.2 Data collection**

140 In total, 136 arborescent palms from nine species were individually measured, harvested,  
141 and weighed in 2006 and 2011 (Table 1). Similar methods were used throughout. Sampling was  
142 designed to represent the entire range of stem heights exhibited by each species. In Madre de  
143 Dios, individuals from each species were selected within a 100-m radius of dicot trees harvested  
144 in a concurrent study (Goodman et al., in press), and the first individual encountered to fulfil the  
145 stem height criteria was selected so that there was no bias towards any certain form or structural  
146 integrity. Before harvesting,  $D$  was measured at 1.3 m or above the highest root and total height  
147 ( $H_{tot}$ ) was measured from the ground to the highest point of the highest leaf. After felling, stem  
148 height ( $H_{stem}$ ) was measured from ground level to the point where the first (lowest) leaf parted  
149 from the stem. All leaves were counted, and, in the Madre de Dios dataset, the length of three  
150 randomly-selected leaves was measured from the base of the rachis to the tip of the terminal  
151 leaflet.

152 Fresh mass of all plant parts was measured in the field immediately after felling.  
153 Aboveground parts were divided into aboveground roots, stem, leaves (petiole, rachis, and  
154 leaflets), and other parts (flowers, fruits, bracts, etc), and measured in the field with a 250 kg  
155 capacity scale with 0.1 kg precision. In Loreto, belowground roots were also sampled following  
156 Gallardo-Ordinola (2001). Fine roots were sampled from eight soil cores (10 cm diameter and 90  
157 cm deep). Four cores were excavated from each of two directions extending 80 cm from the base  
158 of the stem at 90°. The entire main root was then extracted using a 3-ton hand winch and  
159 weighed (Freitas et al., 2006).

160 Stem samples were collected from 3–4 individuals per species (except *Bactris*,  $n = 2$ ) to  
161 estimate moisture content. In Madre de Dios, samples were collected from individuals in the  
162 lower, middle, and upper height classes per species; and three samples were collected from each  
163 individual —at the base, middle, and top of each stem (Table A.1). In Loreto, three individuals  
164 were randomly selected, and one stem sample was collected from each individual. In Madre de

165 Dios, we collected a composite leaf sample consisting of one leaf sample from each species  
 166 including the rachis and attached leaflets. In Loreto, leaves were sampled from three individuals  
 167 per species. Fresh mass of each sample was measured immediately in the field. Samples were  
 168 then air-dried and transported to a drying oven. In this study, we did not measure  $\rho$  directly  
 169 because volume measurement errors would have been virtually unavoidable. Measuring volume  
 170 by water displacement would have relied upon doing so immediately, which was not possible in  
 171 the field, and estimating volume by calliper or ruler measurements would have been inaccurate  
 172 due to uneven edges and thickness of the sample cut.

173 [Table 1]

### 174 **2.3 Laboratory work and data preparation**

175 Stem samples were dried at 101 °C and leaf samples at 65 °C (Williamson and Wiemann,  
 176 2010), and dry mass was recorded after three consecutive days of constant mass with a digital  
 177 scale with 1 mg precision. Dry mass fraction (*dmf*) was calculated as the proportion of dry mass  
 178 per unit fresh mass (dry mass/ fresh mass or 1 – moisture content). Individual mean *dmf* was  
 179 calculated as the mean of three samples taken at different points along the stem (Table A.1), and  
 180 species mean *dmf* was calculated as the average of individual means (Table 2). Carbon content  
 181 was determined for *Mauritiella* and *Mauritia* by calorimetry (Segura-Madrigal, 1997) at  
 182 Universidad Nacional Agraria La Molina, Lima, Peru. To test whether *dmf* is a better  
 183 explanatory variable than wood density, we followed the established practice of assigning  $\rho$   
 184 values to each individual species to the finest taxonomy available according the Global Wood  
 185 Density Database (Chave et al., 2009; Zanne et al., 2009).

186 Stem, root, and leaf dry mass of every individual in the database was calculated as fresh  
 187 mass  $\times$  *dmf*, where *dmf* is mean *dmf* for each tissue for each species (Table 2). Mean individual  
 188 leaf mass was calculated by dividing total leaf mass by the number of leaves.

### 189 **2.4 Model development and evaluation**

190 All species were arborescent with a single stem and multiple leaves. Because of their  
 191 simple growth form (no or very little diameter growth and no branches), models were created  
 192 using  $H_{stem}$ ,  $H_{tot}$ , and two compound variables based on the premise that palms are nearly  
 193 cylindrical ( $D^2H_{tot}$  and  $D^2H_{stem}$ ). Single-species models to estimate AGB were created for each  
 194 species, except *Bactris gasipaes* ( $n = 3$ ), and to estimate belowground biomass for *Mauritia* and

195 *Mauritiella*. Given the ubiquity of forest inventories measuring  $D$  only, we also attempted to  
 196 create models to estimate AGB without any height variable.

197 To create family-level equations, data from all species were combined. A subset of data  
 198 —the individuals from which stem  $dmf$  samples were taken ( $n = 27$ )— were excluded to test the  
 199 developed models. We used the same five variables as the single-species equations, plus four  
 200 additional compound variables,  $dmfD^2H_{tot}$ , and  $dmfD^2H_{stem}$ ,  $\rho D^2H_{tot}$ ,  $\rho D^2H_{stem}$ , where  $dmf$  is the  
 201 species mean  $dmf$  of the stem determined in this study and  $\rho$  is ‘wood’ density for species, genus,  
 202 or family obtained from Global Wood Density Database (Chave et al., 2009; Zanne et al., 2009).  
 203 Finally, we created a mixed-species regression model to estimate mean leaf mass from leaf  
 204 length.

205 For each explanatory variable, we tested five model forms: simple linear, third-order  
 206 polynomial, exponential, logarithmic, and power. In several instances, variables in the linear  
 207 models had to be transformed to satisfy the assumption of equal variance and normality of the  
 208 residuals. Non-significant terms were removed via backwards elimination. All models were built  
 209 using the linear model function (`lm`) in R, version 2.15.1. For family-level models, we also  
 210 performed a generalised linear model (`glm`) analysis using the final linear model and species to  
 211 test whether the slope and intercept terms were significantly different between species (e.g.,  
 212  $AGB \sim D + \text{Species}$ ). However, because no *a priori* factor —such as habitat or phylogeny— could  
 213 explain species differences, we included all species in the final equations to make them the most  
 214 broadly-applicable possible. We evaluated models based on coefficient of determination ( $R^2$ ),  
 215 residual standard error (RSE), and Akaike information criterion (AIC), when comparable.

216 Next, all family-level models were evaluated against the test data ( $n = 27$ ) to examine their  
 217 suitability. For the test data, dry mass was calculated from the directly-measured  $dmf$  and fresh  
 218 mass and of each individual (Table A.1). Finally, we used the full directly measured palm  
 219 biomass dataset ( $n = 136$ ) to both further evaluate the recommended models and to assist the  
 220 interpretation of the forest plot analysis. A correction factor,  $\exp(\text{RSE}^2/2)$ , was applied to  
 221 biomass estimates from logarithmically transformed models (Baskerville, 1972). We examined  
 222 the errors produced by the recommended species-level models, selected family models, and two  
 223 dicot equations (Feldpausch et al., 2012). Errors (kg) were calculated on the original scale as  
 224  $\text{mass}_{\text{predicted}} - \text{mass}_{\text{observed}}$ , and relative errors (%) were calculated as  $\text{error} / \text{mass}_{\text{observed}} \times 100 \%$ .  
 225 We compared the equations based on mean error, mean % error (mean error / mean AGB  $\times 100$

226 %), and mean and standard deviation of relative errors. Overall predictability was assessed by  
 227 standard deviation of the relative errors (Chave et al., 2005), and  $R^2$  was calculated on the  
 228 original scale as  $1 - (SS_{\text{error}} / SS_{\text{total}})$ .

## 229 **2.5 Implications for forest biomass**

230 To explore the implications of using new palm models on palm and forest biomass  
 231 estimates in western Amazonia, we estimated stand level AGB density on nine, 1-ha permanent  
 232 plots within the Tambopata National Reserve in Madre de Dios, Peru (12.8° S, 69.3°W). Plots  
 233 were established between 1979 and 2010 and have been recensused every 2–3 years by  
 234 RAINFOR researchers (Malhi et al., 2002). All individuals with  $D \geq 10$  cm are included in the  
 235 inventories and have been botanically identified. In 2011,  $D$  of all individuals was remeasured,  
 236 and data were obtained from the ForestPlots.net database on 2 August 2012 (Lopez-Gonzalez et  
 237 al., 2011; Lopez-Gonzalez et al., 2012). For this study, total and stem heights were measured on  
 238 all palms with a laser hypsometer (Nikon Forestry 550) during the same year.

239 We estimated AGB using two published dicot and four new palm models. First, we used  
 240 two new pantropical biomass models (Feldpausch et al., 2012) using  $\rho$  and  $D$  only (Feld 1) and  $\rho$ ,  
 241  $D$ , and estimated  $H$  (Feld 2) to estimate AGB of all trees and palms. Total height was estimated  
 242 from  $D$  using the Weibull model for western Amazonia (Table 3 in Feldpausch et al., 2012).  
 243 Next, we recalculated AGB of all palms using the recommended species-level models and three  
 244 family-level models (Table 3). For species without a specific model (*Astrocaryum gratum*,  
 245 *Attalea butyracea*, *A. cephalotes*, *A. maripa*, and *O. mapora*; 7.4 % of all palms), we used the  
 246 model for the same genus. Each of the new palm estimates were compared to estimates made by  
 247 the two dicot models at the stand level.

## 248 **3 Results**

### 249 **3.1 Architecture and internal properties**

250 For most species, height–diameter relationships were weak and height could not be  
 251 predicted from  $D$  (Figure 1A). Some species had a broad range of heights across a broad range of  
 252 diameters with very little relationship between the two (*Astrocaryum* and *Attalea*); the two  
 253 wetland species had a broad range of heights over a very narrow range of diameters (*Mauritia*  
 254 and *Mauritiella*); others were clustered with very narrow height and diameter ranges  
 255 (*Oenocarpus* and *Socratea*); while height and diameter were clearly related for *Euterpe* and  
 256 *Iriarteia*.

257 Dry mass fraction varied between species (Table 2), among individuals of the same  
 258 species, and along the stem of the same individual (Appendix, Table A.1). There appeared to be  
 259 a slight negative relationship between  $dmf$  and height at which the sample was collected (i.e.,  $dmf$   
 260 greatest at base) and a very slight positive relationship between mean stem  $dmf$  and the height of  
 261 the individual (i.e., mean  $dmf$  greater in taller individuals). However, no significant relationships  
 262 could be determined, so we calculated individual  $dmf$  as the mean of the three samples taken  
 263 along each stem and species-level  $dmf$  as the simple mean of the three individuals per species.  
 264 Mean  $dmf$  was consistently higher for leaf tissue than stem or root tissues (Table 2). Carbon  
 265 fraction was usually slightly  $< 50\%$  of dry mass in *Mauritia* and *Mauritiella* and similar  
 266 between the two species in stem and root tissues, but it was more variable and slightly higher and  
 267 in leaf tissue.

268 Individual AGB varied across more than three orders of magnitude, from as little as 0.7 kg  
 269 to as much as 1231 kg. AGB generally increases with stem diameter when all species are  
 270 combined (Figure 1B), but within a species AGB is more strongly related to stem height (Figure  
 271 1C). Belowground root biomass contributed 13–780 kg in *Mauritia* and *Mauritiella*, representing  
 272 13–47 % of total plant dry mass. Mean leaf mass varied by over an order of magnitude between  
 273 species, ranging from 0.2 kg leaf<sup>-1</sup> in *Bactris* to 14.2 kg leaf<sup>-1</sup> in *Oenocarpus* (Table 2).

274 [Table 2]

### 275 3.2 Species-level models

276 Single-species models performed well, with  $R^2 > 0.90$  for most species (Table 3, Figure 2).  
 277 Height was the key variable to estimate AGB, and including  $D$  added little to or even worsened  
 278 model performance. Models with  $H_{stem}$  alone were better than those with the compound variable  
 279  $D^2H_{stem}$  for most of species, and models with  $H_{tot}$  alone were always better than those with  $D^2H_{tot}$   
 280 (Table A.2). Models with only  $D$  were only significant for *Euterpe*, *Iriarteia*, and *Socratea*. For  
 281 all other species, AGB could not be estimated from  $D$  alone. The recommended models, one for  
 282 each species, are listed in Table 3. Other models, with different predictor variables, are available  
 283 in Table A.2.

284 [Table 3]

285 [Figure 2]

### 286 3.3 Family-level models

287 The transformed model with compound term  $dmfD^2H_{stem}$  best estimated AGB of all species  
 288 (Table 3). This model was selected as best from the metrics used to evaluate built models ( $R^2$ ,  
 289 RSE, AIC; Table A.3) and performed well against the test data (Table A.4). Separating these  
 290 variables in a logarithmically-transformed additive model did not improve model performance.  
 291 One individual with no stem ( $H_{stem} = 0$  m) had to be removed as an outlier; thus, these models are  
 292 only valid for individuals with  $H_{stem} \geq 1$  m.

293 The best family-level model ( $dmfD^2H_{stem}$ ) showed some differences between species, but  
 294 the glm analysis revealed that slope and intercept were only significantly different for only one  
 295 species each (*Astrocaryum* and *Mauritia*, respectively;  $P < 0.05$ ). This model generally  
 296 underestimated AGB for *Astrocaryum*, *Attalea*, *Mauritia*, and *Oenocarpus* and overestimated  
 297 mass for *Bactris*, *Euterpe*, *Iriarte*, *Mauritiella*, and *Socratea* (Figure 3A).

298 To permit palm biomass estimation from inventories that have not measured  $H_{stem}$ , we  
 299 explored the use of other predictor variables. Visible trends were observed between AGB and  $D$ ,  
 300  $H_{tot}$ , and  $D^2H_{tot}$ , but the relationships were subject to outliers or anomalies (Figure A.1). Thus,  
 301 we had to remove outliers, and the resulting equations are only valid within the given range  
 302 (Table 3). Models with just  $D$ , or  $D$  and  $dmf$ , performed reasonably well but are only valid for  
 303 individuals with diameters between 6 and 40 cm and stem heights  $> 3$  m. Prediction errors from  
 304 these models showed few differences between species, except that *Mauritia* was almost always  
 305 underestimated and *Iriarte* with  $D < 22$  cm was usually overestimated (Figure 3B,C). The  
 306 compound variable  $D^2H_{tot}$  was especially prone to producing outliers when individuals have very  
 307 short stems with tall leaves or short stems with large diameters: three *Attalea* and one  
 308 *Oenocarpus* were identified as such, all of which had  $H_{stem} \leq 3$  m and  $H_{tot} < 5$  m. Models with  
 309  $H_{tot}$  had a tendency to overestimate AGB of shorter individuals and underestimate taller  
 310 individuals (data not shown). For both pairs, the models with  $dmf$  ( $D+dmf$  and  $H_{tot}+dmf$ ) were  
 311 significantly better than the model with  $D$  or  $H_{tot}$  alone ( $P < 0.05$ ). We also tested models with  $\rho$   
 312 instead of  $dmf$ , but  $\rho$  was never significant. Likewise, models with a compound predictor variable  
 313 using  $\rho$  never performed as well as those with  $dmf$  (i.e.,  $\rho D^2H_{stem}$  vs.  $dmfD^2H_{stem}$ ), so we do not  
 314 report models with  $\rho$ .

315 Leaf mass can be estimated from leaf length (Table 3). We present a mixed species model,  
 316 created from *Astrocaryum*, *Attalea*, *Bactris*, *Euterpe*, *Oenocarpus*, and *Socratea*. However,

317 *Iriartea* leaves showed a very clear, and significantly different, relationship between leaf mass  
 318 and length, so we have reported separate results for this species (Figure A.2).

319 [Figure 3]

### 320 **3.4 Model evaluation**

321 Comparing model predictions to the subset of test data, the recommended family model  
 322 with the compound term  $dmfD^2H_{stem}$  had the lowest bias (mean % error = 0.2 %), but another  
 323 model using the log-transformed compound variable with total height ( $dmfD^2H_{tot}$ ) performed best  
 324 by all other criteria (Table A.4). All family-level models performed reasonably well, except the  
 325 models with  $H_{tot}$  and  $H_{tot}+dmf$  (Table A.4).

326 Testing model estimates against the full biomass dataset ( $n = 136$ ), the species model  
 327 estimates were always best, followed by the family-level  $dmfD^2H_{stem}$  model (Table A.5). The two  
 328 models without any height variable,  $D$  and  $D+dmf$ , and the two dicot models all overestimated  
 329 AGB and performed very poorly when applied to the full biomass dataset ( $-0.184 \leq R^2 \leq 0.145$ ).  
 330 However, when the dataset was reduced to only the individuals for which all models were valid  
 331 ( $H_{stem} > 3$  m and  $6 \leq D < 40$  cm;  $n = 125$ ), results for the recommended species and family model  
 332 estimates changed little, but estimates from the palm and dicot models without measured height  
 333 improved substantially ( $0.548 \leq R^2 \leq 0.615$ ; Table A.5). The species-level and Feld 1 models  
 334 slightly overestimated AGB (mean % error = 3 and 6 %, respectively), the recommended family-  
 335 level and Feld 2 models slightly underestimated AGB (mean % error = -5 and -7 %, respectively),  
 336 while the  $D$  and  $D+dmf$  model estimates were nearly neutral (mean % error = 0.6 and -0.03, respectively).  
 337 By nearly all metrics, all palm models were better estimators of palm  
 338 AGB than the dicot models.

339 The dicot models were poor estimators of individual palm AGB, with errors ranging from  
 340 -844 to +1651 kg. Whether each one over- or underestimates palm mass was largely dependent  
 341 upon species, diameter, and stem height. The magnitude of errors increased considerably with  
 342 diameter, but the direction of errors was largely dependent on stem height (Figure 4). The dicot  
 343 models typically overestimated AGB of palms with short stems but underestimated AGB of  
 344 taller stemmed individuals. This crossover occurred at  $H_{stem}$  c. 14 and 15 m for the Feld 1 and 2  
 345 models, respectively. Prediction errors between species are consistent between the two dicot  
 346 models, but the Feld 2 model estimates were usually lower. Both dicot models tended to  
 347 overestimate AGB of *Astrocaryum*, *Oenocarpus*, and any palm with  $D > 40$  cm and to

348 underestimate AGB of *Mauritia*, *Socratea*, most palms with  $H_{stem} > 15$  m, and all palms with  
 349  $H_{stem} > 22$  m.

350 [Figure 4]

### 351 **3.5 Implications for forest biomass**

352 Across the nine plots in Tambopata, palms represented between 3 and 32 % of all stems ( $D$   
 353  $\geq 10$  cm) and, based on species-level equations, contributed from 5 to 43 Mg of above ground  
 354 biomass per hectare. The two dicot models yielded different AGB estimates, for both trees and  
 355 palms the Feld 2 (with estimated height) estimates were lower than the Feld 1 estimates (Table  
 356 4). Using estimates from the recommended species models (Table 3), palm AGB density in the  
 357 nine plots was on average 29 or 40 % greater than would have been estimated with the Feld 1  
 358 and 2 dicot models, respectively. However, plot means may be artificially large due to large  
 359 relative differences in two plots with very low palm presence (TAM05 and TAM07; Table 4).  
 360 Thus, if palm biomass on all plots is combined as one unit, the overall difference in palm mass  
 361 across all nine plots is 14 or 27 % greater than Feld 1 and 2 estimates, respectively. Total AGB  
 362 density estimates of the whole stand (i.e., dicot trees and palms combined) were between 1 and 2  
 363 % greater than the Feld models (Table 4).

364 Stand-level palm AGB estimated from the other three palm models was usually lower than  
 365 the species-level palm model estimates. Among the family-level equations, the model with  $D$   
 366 only yielded the most similar results to the species-level estimates, followed by the  $dmfD^2H_{stem}$   
 367 model, but the  $D+dmf$  model produced much lower estimates. In some plots the family-level  
 368 models gave lower AGB estimates than did the dicot models, but across all plots palm AGB is  
 369 still higher than would have been estimating using dicot models (Table 4).

370 [Table 4]

## 371 **4 Discussion**

### 372 **4.1 Architecture and intrinsic properties**

373 Our data appear to have captured several different growth patterns of arborescent palms, as  
 374 demonstrated by the differing relationships between diameter, height, and AGB among species  
 375 (Figures 1 and A.1). These differing allometries have implications for the best single- and mixed-  
 376 species biomass models. For example, *Iriartea* (Rich, 1987a), *Socratea* (Rich et al., 1986), and  
 377 *Euterpe* (Avalos and Fernandez Otarola, 2010) can continue to increase in diameter via sustained  
 378 cell expansion (Rich, 1987a; Renninger and Phillips, 2012), and as a result these are the same

379 three genera for which we were able to estimate AGB from  $D$  alone. For other genera, such as  
380 *Mauritiella*, stem diameter is virtually the same at every height, so including  $D$  in allometric  
381 equations adds little or no explanatory value.

382 Variable moisture content between species, between individuals of the same species, and  
383 within a single stem make it difficult to determine the exact dry mass of palms. As with stem  
384 density (Rich, 1987b), dry mass fraction tends to be greatest at the base and decreases along the  
385 stem. Mean stem  $dmf$  may also increase with increasing stem height, as reported by Rich  
386 (1987b), perhaps because cell walls thicken with age (Rich, 1987a; Tomlinson, 2006).

387 Nonetheless, our data show that using species mean  $dmf$  values works well. Using the test data,  
388 the differences in dry mass estimates, when calculated from fresh mass and either species mean  
389  $dmf$  or  $dmf$  measured for each individual, are very small: the mean difference between the two  
390 calculations was only 2.5 kg or 1.9 %. Therefore, we consider the results reported here to be  
391 reliable and to represent an advance in improving palm biomass estimates. When utilising these  
392 models, values for  $dmf$  can be found in Table 2. For the genera included in this study, we  
393 recommend using mean stem  $dmf$  for the respective taxon and the overall mean (0.370) for all  
394 other genera.

#### 395 **4.2 New models**

396 As hypothesised, height was always a very important parameter to consider in palm  
397 biomass equations. Total height was sometimes a better estimator than stem height. However,  
398 total height is likely to be a less reliable measurement than stem height, as it can be subjective  
399 and often difficult to measure if only one leaf is extending upwards. Thus, we recommended the  
400 second best model for each of these species, which includes  $H_{stem}$  in all cases. Estimating palm  
401 mass using compound variables ( $D^2H_{stem}$  or  $D^2H_{tot}$ ) was often not necessary for single-species  
402 models but was best for family-level models because  $H-D$  relationships differed between  
403 species. As expected, accounting for dry mass fraction also improved mixed-species model  
404 estimates, and  $dmf$  was a better variable than  $\rho$  to account for internal species differences,  
405 perhaps because of the difficulty of measuring palm  $\rho$  accurately.

406 There proved to be a reasonable relationship between these two variables when all species  
407 were combined, and family-level models with  $D$  alone or  $D+dmf$  performed remarkably well  
408 given the weak relationship of diameter with biomass at the species-level. These models  
409 appeared to be unbiased when tested on the full biomass dataset and in plot estimates. However,

410 these models should be applied with caution, as estimates made outside the diameter and height  
411 ranges used to build the models can be erroneous (Table A.5). We provide species- and family-  
412 level palm models with a variety of input data to accommodate existing inventories, and these  
413 models could be used to create new palm biomass estimates from existing forest inventory data  
414 (i.e.,  $D$  and species). These estimates would be more accurate than estimates from dicot  
415 equations, but estimates from palm models including height would be far more accurate (Tables  
416 A.4, A.5). Likewise, because AGB could not usually be predicted from  $D$  within a single species,  
417 it is likely that the relationship between AGB and  $D$  within an individual is also unreliable and,  
418 therefore, that productivity of individual palm trees should preferably also be estimated from  
419 models that include stem height or total height.

### 420 **4.3 Implications for forest biomass**

421 Contrary to our expectations, palm biomass estimates were greater in each of the nine plots  
422 examined when applying our most reliable palm equations, compared to palm biomass estimated  
423 from standard dicot models. Although palms do not have branches or relatively dense stems, they  
424 often weigh more than dicot trees at small diameters because they can be much taller (Rich et al.,  
425 1986). As a result, AGB of many tall palms with small diameters can be underestimated by the  
426 dicot models (Figure 4). Evaluating model estimates on our directly-measured biomass dataset  
427 could not fully reflect this because the destructive dataset was designed to create reliable  
428 regression models across a broad range of sizes, and, thus, includes a higher proportion of short  
429 stemmed individuals than inventoried in mature forests in Tambopata ( $D \geq 10$  cm). Mean  $H_{stem}$  of  
430 palms in the forest plots (15.3 m) was slightly greater than in the biomass dataset (12.7 m) and  
431 above the limit at which dicot models underestimate AGB of most palms (14–15 m). The effect  
432 of using new palm biomass models will also likely be determined by the species composition and  
433 the interaction between size and composition. In Tambopata, for example, *Iriartea deltoidea*  
434 makes up 54% of all registered palms with  $D \geq 10$  cm, followed by *Euterpe precatorea* (14 %),  
435 *Socratea exorrhiza* (12 %), and *Mauritia flexuosa* (7 %). In the destructive biomass dataset, the  
436 dicot models estimated AGB of *Iriartea* moderately well but consistently underestimated AGB  
437 of *Socratea* and *Mauritia* (Figure 4). Thus, palm AGB estimates may be considerably higher  
438 than previously estimated in stands where these species, especially *Mauritia flexuosa*, are  
439 dominant. Conversely, new palm equations may slightly reduce AGB estimates, as compared to  
440 dicot model estimates, in forests where *Oenocarpus bataua* is the major palm component, such

441 as in central Amazonia (Emilio et al., 2013), or in stands where the palm population is dominated  
442 by shorter individuals (Kahn and Mejia, 1990).

443 Our results show higher palm AGB density estimates in plots than would have been  
444 estimated from dicot equations, but the magnitude of this increase depends on the palm equation  
445 used to estimate AGB. When tested on the directly-measured biomass dataset, the species  
446 models slightly overestimated AGB but the recommended family-level model ( $dmfD^2H_{stem}$ ;  
447 Table 3) underestimated AGB by a greater amount. Likewise, the Tambopata plot estimates were  
448 greater using the species-level models than this family model. Thus, the true ‘increase’ in palm  
449 AGB is likely to be in between these estimates but closer to the species model estimates. The  
450 overall differences (when all plots are combined) in palm AGB estimates between the two palm  
451 estimates discussed above are similar. Thus, despite some sensitivity to the palm model used,  
452 true palm AGB in Tambopata is greater than would have been estimated by dicot equations.

453 This dataset and new models do not, however, represent small palms, nor do they capture  
454 the reproductive parts of mature individuals. Though stemless and juvenile palms can be  
455 abundant in some ecosystems (Kahn and Mejia, 1990), they are not generally included in forest  
456 inventories and contribute little to forest biomass (Nascimento and Laurance, 2002). We also  
457 found that although palm leaves are large, they weigh little compared to the woody tissues.  
458 Because no individuals in the biomass dataset were fruiting at the time of harvest, these estimates  
459 do not accurately account for reproductive parts.

460 These new palm biomass equations should have multiple applications and facilitate more  
461 accurate estimates of carbon stocks and cycling in tropical forests. Though increases in whole  
462 forest estimates are locally small (0.9–1.8 % at Tambopata), this increase could be expected to  
463 impact total carbon stock estimates in tropical forests more broadly, particularly in forests with  
464 hydromorphic soils. These models may also finally assist accurate quantification of above- and  
465 belowground carbon stocks of the palm community in the extensive, carbon-rich peat ecosystems  
466 which cover *c.* 150,000 km<sup>2</sup> in Amazonia (Lahteenoja et al., 2009).

467 These new equations can also be used to improve palm productivity estimates. Palm stem  
468 productivity may also be greater than previously estimated by dicot models using diameter —as  
469 commonly measured in permanent plot inventories— because palms grow in height with little or  
470 no corresponding increase in stem diameter (Rich et al., 1986). Estimating leaf mass by either the  
471 species mean or leaf length will allow researchers to account for litterfall from palms, which is

472 usually ignored due to the technical difficulties of sampling palm leaf fall (Chave et al., 2010).  
473 Though other structures, such as inflorescences, bracts, and fruits, can contribute substantially to  
474 forest productivity (Phillips, 1993), they still remain poorly quantified or ignored in  
475 comprehensive studies (Chave et al., 2010; Malhi et al., in press). These ‘missing terms’ in forest  
476 productivity estimates (Malhi et al., 2009) need to be incorporated, particularly as they can be  
477 expected to vary substantially in space and time. For example, because palms are more abundant  
478 in the western Amazon (Kahn et al., 1988; Terborgh and Andresen, 1998; Eiserhardt et al.,  
479 2011), it is possible that the magnitude of increase in aboveground forest productivity from east  
480 to west across Amazonia may be even greater than previously considered (Malhi et al., 2004;  
481 Aragão et al., 2009; Quesada et al., 2012).

## 482 **5 Conclusions and future directions**

483 This study is the first to create a comprehensive dataset of arborescent palms and family-  
484 level allometric equations to estimate aboveground biomass. We report both single- and mixed-  
485 species models with a variety of input variables to accommodate different forest inventory  
486 methods. Single species models estimated palm biomass very well, as does the recommended  
487 family-level model with  $dmfD^2H_{stem}$ . The family-level models without a height variable provide  
488 unbiased estimates of palm AGB, but should be applied with caution. With these new models,  
489 we can finally estimate palm biomass and productivity more reliably. When equations were  
490 applied to forest plots at one location in western Amazonia, palm biomass density was on  
491 average 14 or 27 % greater than would have been estimated using two pantropical biomass  
492 models for dicot trees. In other forests, the effect of new palm equations on plot biomass  
493 estimates will depend on palm sizes, abundance, and species composition. The magnitude of  
494 palm productivity and carbon cycling fluxes will also likely be greater than previously estimated  
495 by dicot models, but the magnitude of this effect has yet to be formally explored. We recommend  
496 that palm stem height should be measured in future inventories to accurately estimate palm  
497 biomass and, especially, biomass changes in this important forest component.

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721

722 **Tables**

723 Table 1. Directly measured biomass data analysed in this study from Madre de Dios (MdD) and  
 724 Loreto, Peru: number of individuals ( $n$ ), diameter at 1.3 m or above roots ( $D$ ), stem height  
 725 ( $H_{stem}$ ), and total height ( $H_{tot}$ ).

<b>Location</b>	<b>Species</b>	<b><math>n</math></b>	<b>D</b> (cm)	<b><math>H_{stem}</math></b> (m)	<b><math>H_{tot}</math></b> (m)
MdD	<i>Astrocaryum murumuru</i>	19	15-29	1.5-9.0	7.1-14.7
MdD	<i>Attalea phalerata</i>	21	17-50	0-20.1	7.1-25.6
MdD	<i>Bactris gasipaes</i>	3	11-15	9.3-18.1	13.0-20.8
MdD	<i>Euterpe precatoria</i>	8	12-19	10.2-20.4	13.3-22.8
MdD	<i>Iriartea deltoidea</i>	21	6-33	3.3-21.8	5.6-25.1
Loreto	<i>Mauritia flexuosa</i>	16	19-36	5.1-30.5	9.1-38.4
Loreto	<i>Mauritiella aculeata</i>	18	8-15	3.5-20.6	5.3-26.1
MdD	<i>Oenocarpus bataua</i>	10	21-41	2.9-14.5	14.2-25.9
MdD	<i>Socratea exorrhiza</i>	20	4-24	2.0-21.9	3.2-23.9
<b>Total</b>	<b>9 species</b>	<b>136</b>	<b>4-50</b>	<b>0-30.5</b>	<b>3.2-28.4</b>

726

727 Table 2. Dry mass fraction (dry mass / fresh mass) and carbon fraction (dry mass<sub>carbon</sub> / dry  
 728 mass<sub>total</sub>) in stem, leaf, and root tissue, and mean and standard deviation of individual leaf dry  
 729 mass of the nine species sampled.

Species	Stem	Leaf	Root	Leaf dry mass (kg)	
	<i>Dry mass fraction</i>			mean	SD
<i>Astrocaryum murumuru</i>	0.400			2.687	1.057
<i>Attalea phalerata</i>	0.357			2.649	0.938
<i>Bactris gasipaes</i>	0.619			0.471	0.132
<i>Euterpe precatorea</i>	0.398			0.620	0.269
<i>Iriartea deltoidea</i>	0.244			4.065	3.787
<i>Oenocarpus bataua</i>	0.338			9.315	1.683
<i>Socratea exorrhiza</i>	0.339			1.764	1.629
mixed species (above)		0.463			
<i>Mauritia flexuosa</i>	0.367	0.517	0.402	11.444	5.845
<i>Mauritiella aculeata</i>	0.269	0.320	0.297	0.951	0.447
<b>mean</b>	<b>0.370</b>	<b>0.433</b>	<b>0.350</b>	<b>3.774</b>	<b>1.754</b>
	<i>Carbon fraction</i>				
<i>Mauritia flexuosa</i>	0.481	0.494	0.491		
<i>Mauritiella aculeata</i>	0.480	0.529	0.485		
<b>mean</b>	<b>0.481</b>	<b>0.512</b>	<b>0.488</b>		

730 Table 3. Recommended models for each genus and mixed-species to estimate aboveground biomass (AGB; kg dry mass) or  
 731 belowground root biomass (BGB; kg dry mass) from stem height ( $H_{stem}$ ; m), diameter ( $D$ ; cm), and dry mass fraction ( $dmf$ ;  $g\ g^{-1}$ ), and  
 732 leaf dry mass (kg) from mean leaf length (m). All models follow the form  $y = a + bx_1 + cx_2$ . The family-level model with  $dmfD^2H_{stem}$   
 733 is only valid for individuals with  $H_{stem} \geq 1$  m; and the family-level models without a height term are only valid for individuals with  
 734  $H_{stem} > 3$  m and  $6 \leq D < 40$  cm.

Genus or group	y	x <sub>1</sub>	x <sub>2</sub>	a	b	c	R <sup>2</sup>	n	RSE	F	dfe	Pr < F	AIC
<b>Aboveground biomass</b>													
<i>Astrocaryum</i>	AGB	$H_{stem}$		.	21.302		0.957	18	26.1	379	17	<0.0001	171.4
<i>Attalea</i>	ln(AGB)	$\ln(H_{stem}+1)$		3.2579	1.1249		0.858	21	0.371	115	19	<0.0001	21.9
<i>Euterpe</i>	AGB	$H_{stem}$		-108.81	13.589		0.973	8	8.37	215	6	<0.0001	60.4
<i>Iriarte</i>	ln(AGB)	$\ln(D^2H_{stem})$		-3.483	0.94371		0.967	21	0.311	560	19	<0.0001	14.5
<i>Mauritia</i>	ln(AGB)	$\ln(H_{stem})$		2.4647	1.3777		0.897	16	0.273	121	14	<0.0001	7.7
<i>Mauritiella</i>	AGB	$H_{stem}$		.	2.8662		0.972	18	8.21	591	17	<0.0001	129.9
<i>Oenocarpus</i>	ln(AGB)	$H_{stem}$		4.5496	0.1387		0.784	10	0.237	29	8	0.000658	3.4
<i>Socratea</i>	ln(AGB)	$\ln(D^2H_{stem})$		-3.7965	1.0029		0.976	20	0.227	740	18	<0.0001	1.3
	AGB <sup>0.25</sup>	$(dmf \times D^2 H_{stem})^{0.25}$		.	0.55512		0.990	106	0.367	10410	105	<0.0001	91.5
Family-level	ln(AGB)	$\ln(D)$		-3.3488	2.7483		0.802	97	0.588	384	95	<0.0001	176.1
	ln(AGB)	$\ln(D)$	$\ln(dmf)$	-2.0752	2.6401	0.8426	0.815	97	0.570	208	94	<0.0001	171.1
<b>Belowground root biomass</b>													
<i>Mauritia</i>	ln(BGB)	$\ln(H_{stem})$		-0.3688	2.0106		0.929	16	0.323	184	14	<0.0001	13.1
<i>Mauritiella</i>	ln(BGB)	$H_{stem}$		1.0945	0.11086		0.951	18	0.132	310	16	<0.0001	-18.0
<b>Leaf dry mass</b>													
Family-level	mass <sup>0.3</sup>	length		0.66020	0.10896		0.732	76	0.171	202	74	<0.0001	-48.5
<i>Iriarte</i>	ln(mass)	length		-5.1751	1.4547		0.803	21	0.649	78	19	<0.0001	45.4

735

736 Table 4. Palm and dicot tree aboveground biomass density estimates for nine RAINFOR plots in Tambopata National Reserve, Peru.  
737 Estimates were made using two recent pantropical dicot models (Feldpausch et al., 2012) based on diameter and wood density (F 1) or  
738 diameter, wood density, and estimated height (F 2); three family-level palm equations ( $D$ ,  $D+dmf$ , and  $dmfD^2H_{stem}$ ; Table 3); and the  
739 recommended species-level model for each species or genera (Table 3). Percent differences are shown for each palm model compared  
740 to each dicot model. Results are summarised as mean of all plots, sum of all plots, and overall percent difference.

741

Plot	No. stems	% palm stems	Aboveground biomass estimate (Mg)						% Difference: (Palm - Dicot)/Dicot × 100 %							
			Dicot		Palm equation				D		D+dmf		dmfD <sup>2</sup> H <sub>st</sub>		Species	
			F 1	F 2	D	D+dmf	dmfD <sup>2</sup> H <sub>st</sub>	Species	F 1	F 2	F 1	F 2	F 1	F 2	F 1	F 2
TAM01	598	31.9	279	233					3.4	5.1	0.9	2.1	2.7	4.2	1.4	2.7
	191		27	25	36.5	29.3	34.4	30.9	35.4	48.6	8.9	19.5	27.6	40.1	14.9	26.1
TAM02	659	28.8	262	223					2.6	3.9	0.7	1.7	2.8	4.1	1.8	2.9
	190		24	22	31.1	26.0	31.5	28.8	28.4	39.3	7.3	16.4	29.9	40.9	18.9	29.0
TAM03	617	15.9	372	312					-0.2	1.6	0.4	2.3	-0.7	0.9	0.3	2.2
	98		42	36	41.1	43.3	39.0	42.9	-1.5	13.6	3.7	19.7	-6.6	7.7	2.8	18.6
TAM04	714	9.0	354	299					-0.3	0.1	-0.6	-0.3	-0.6	-0.3	1.1	1.7
	64		14	13	12.8	11.7	11.6	17.7	-7.7	1.6	-	-7.2	-	-7.9	27.3	40.1
TAM05	526	3.8	316	262					-0.1	0.0	-0.1	-0.1	0.1	0.2	0.6	0.8
	20		3	3	2.6	2.4	3.2	4.8	-8.0	-2.8	-	-9.7	13.1	19.5	70.3	79.9
TAM06	660	31.8	359	297					2.7	4.4	0.8	2.2	1.7	3.2	0.4	1.7
	210		34	30	43.6	36.9	40.0	35.4	28.8	43.0	9.0	21.0	18.2	31.2	4.6	16.1
TAM07	507	3.4	267	224					-0.4	-0.4	-0.4	-0.4	-0.3	-0.4	0.8	1.0
	17		3	3	1.7	1.8	1.9	4.9	-	-	-	-	-	-	74.9	81.8
TAM08	513	12.3	266	222					0.6	1.0	-0.1	0.2	0.5	0.9	1.1	1.6
	63		9	9	10.9	9.1	10.6	12.3	15.6	25.0	-3.8	4.1	12.8	22.0	30.2	40.8
TAM09	552	15.2	271	228					1.5	2.3	0.4	0.9	1.0	1.7	0.6	1.3
	84		13	11	16.5	13.5	15.3	14.2	32.0	45.4	7.6	18.6	22.2	34.7	13.6	25.1
Mean of all plots	594		305	256					1.1	2.0	0.2	0.9	0.8	1.6	0.9	1.8
	104	16.9	19	17	21.9	19.3	20.8	21.3	9.3	19.7	-3.6	5.5	7.5	17.5	<b>28.6</b>	<b>39.7</b>
Sum or Overall % diff.	Sum of all plots (Mg)								Overall % Difference ( $\sum \text{Diff} / \sum \text{Estim} \times 100 \%$ )							
	4673		2745	2300					1.0	2.0	0.2	1.0	0.7	1.6	0.9	1.8
	938	20.1	168	152	197	174	187	192	17.0	29.8	3.4	14.8	11.4	23.7	<b>14.1</b>	<b>26.6</b>

742 **Figure captions**

743

744 Figure 1. Raw data showing the relationships between (A) stem height and diameter and (B)  
745 aboveground biomass and diameter, and (C) aboveground biomass and stem height for each  
746 species.

747

748 Figure 2. Aboveground biomass (AGB) data and recommended model (line) for each genus to  
749 estimate AGB (kg dry mass) from stem height ( $H_{stem}$ ; m) and diameter at 1.3 m or above stilt  
750 roots ( $D$ ; cm). Equations are given in Table 3.

751

752 Figure 3. Aboveground biomass (AGB) data for all species (points) and three family-level model  
753 estimates (lines) using (A)  $dmfD^2H_{stem}$ , (B)  $D$ , and (C)  $D+dmf$ . Equations are given in Table 3.

754

755 Figure 4. Errors ( $AGB_{estimated}-AGB_{observed}$ ) for harvested palm aboveground biomass (AGB)  
756 when estimated by dicot models using diameter and wood density (Feld 1) and diameter,  
757 wood density, and estimated height (Feld 2; Feldpausch et al., 2012) compared with diameter  
758 and stem height.

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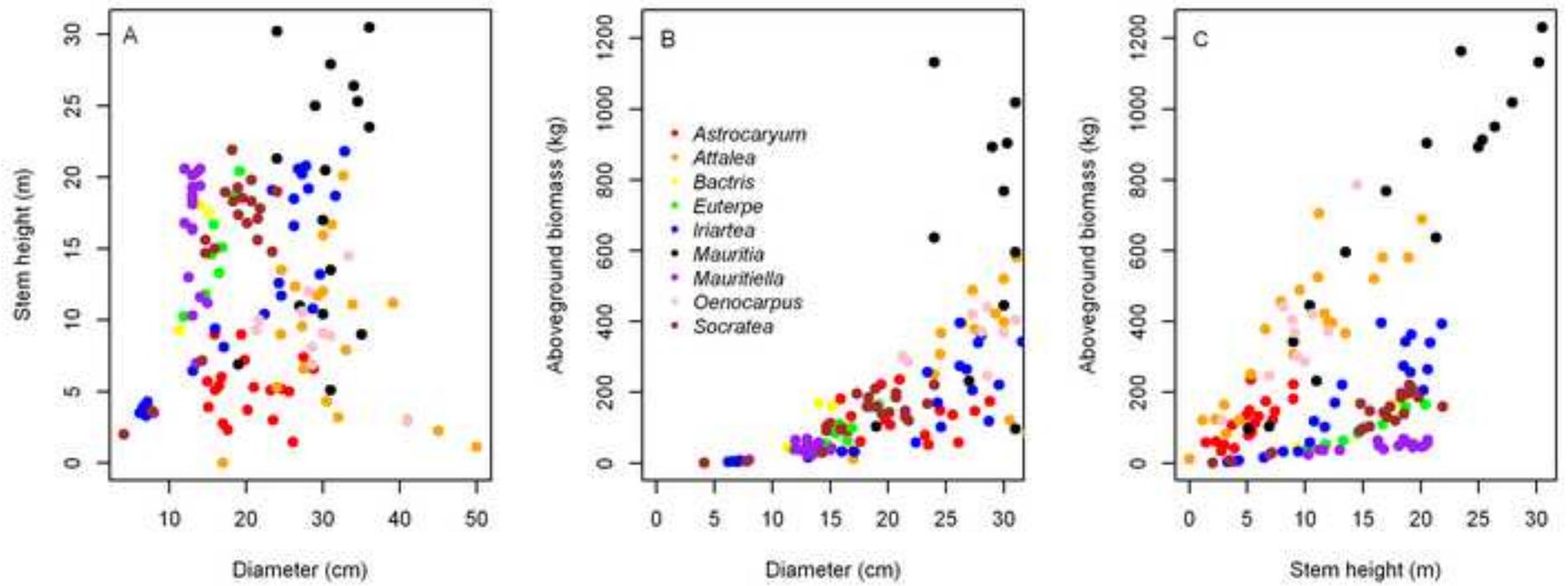


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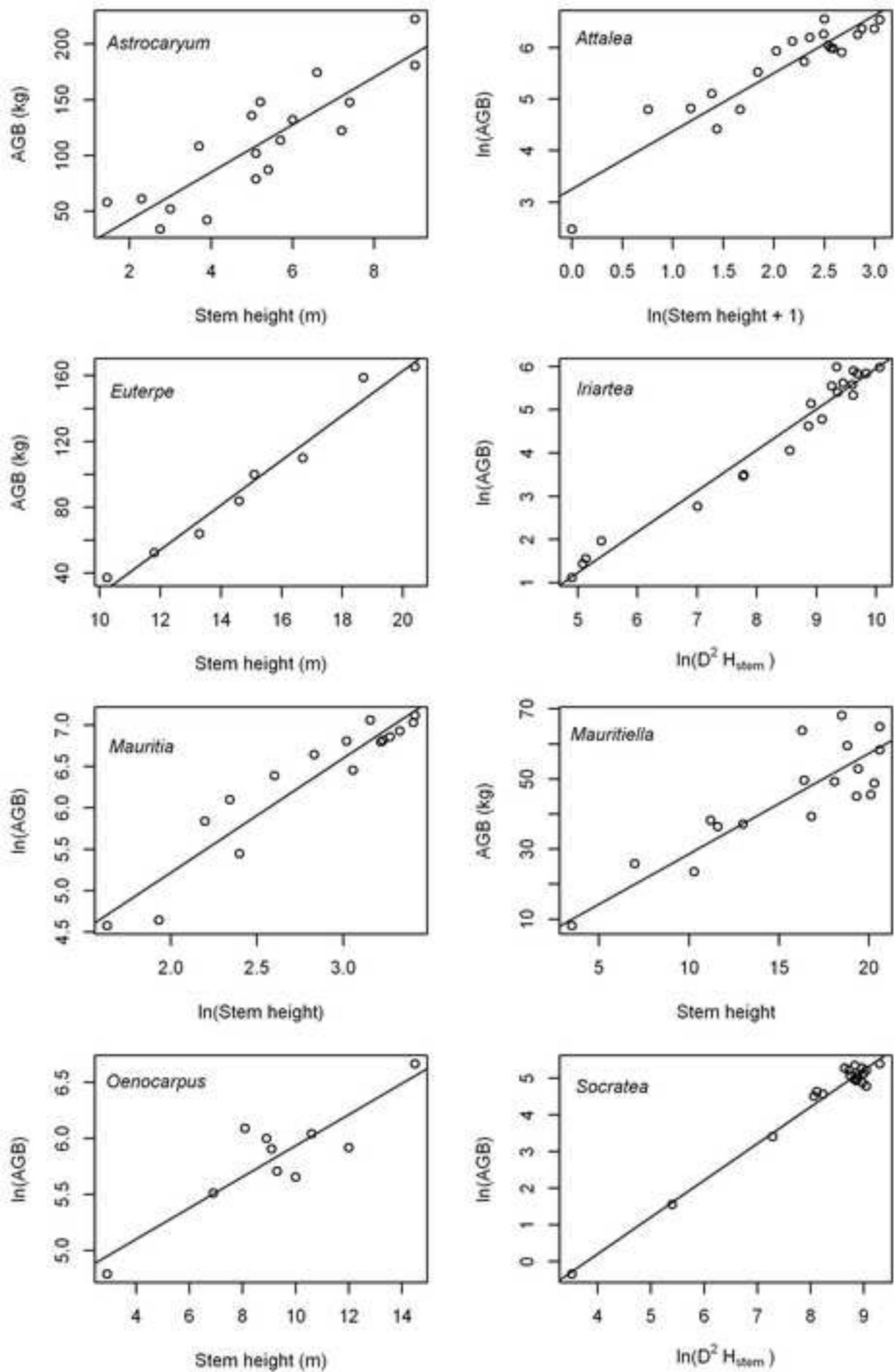


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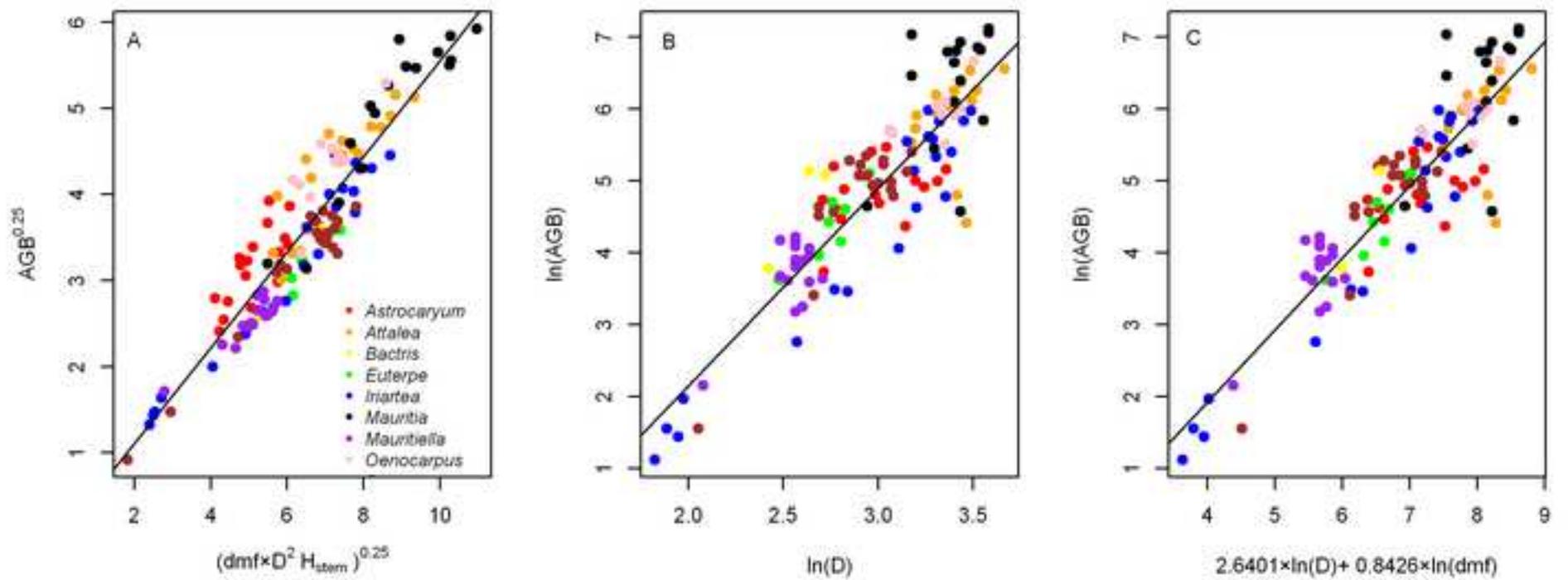


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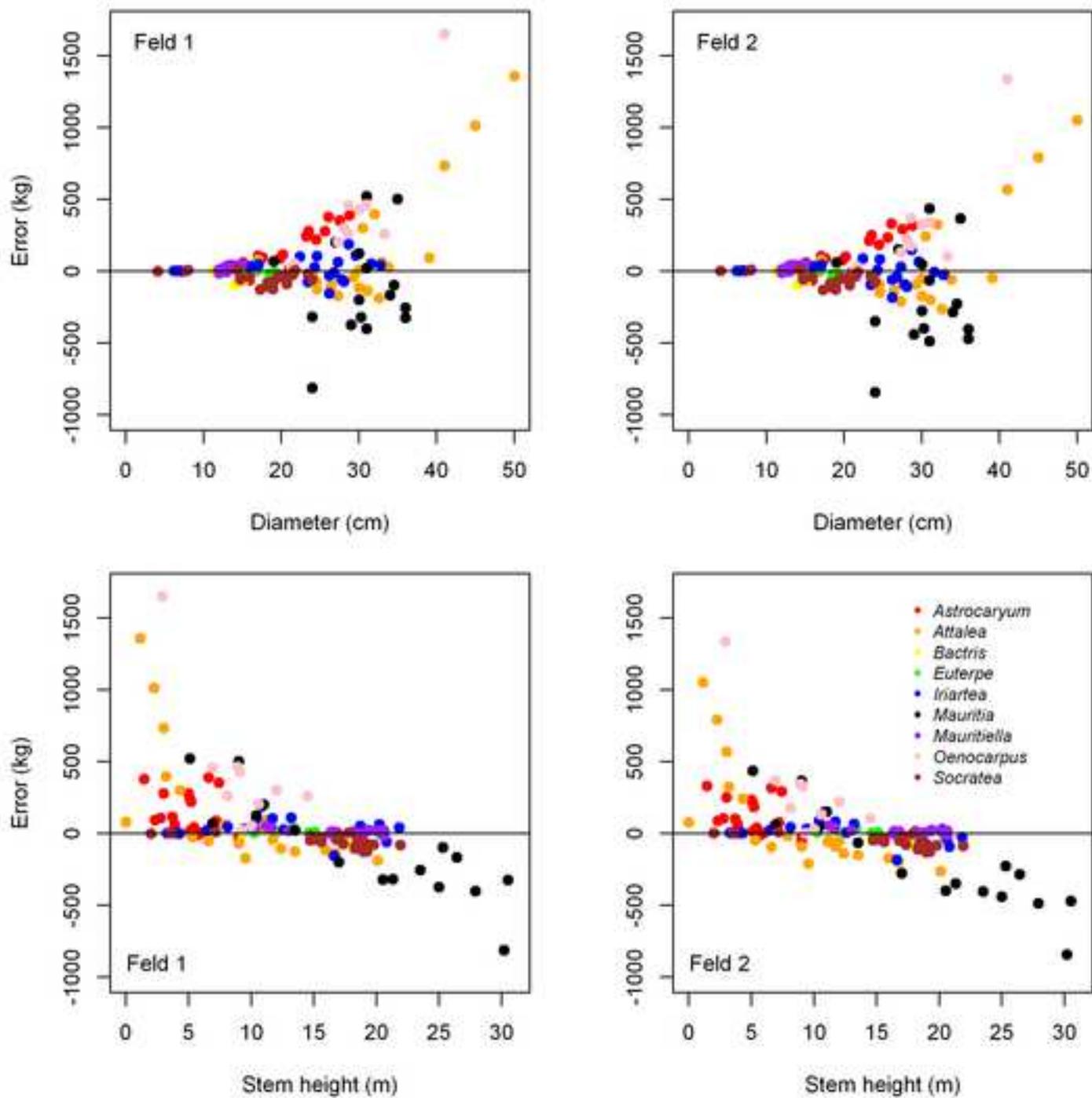


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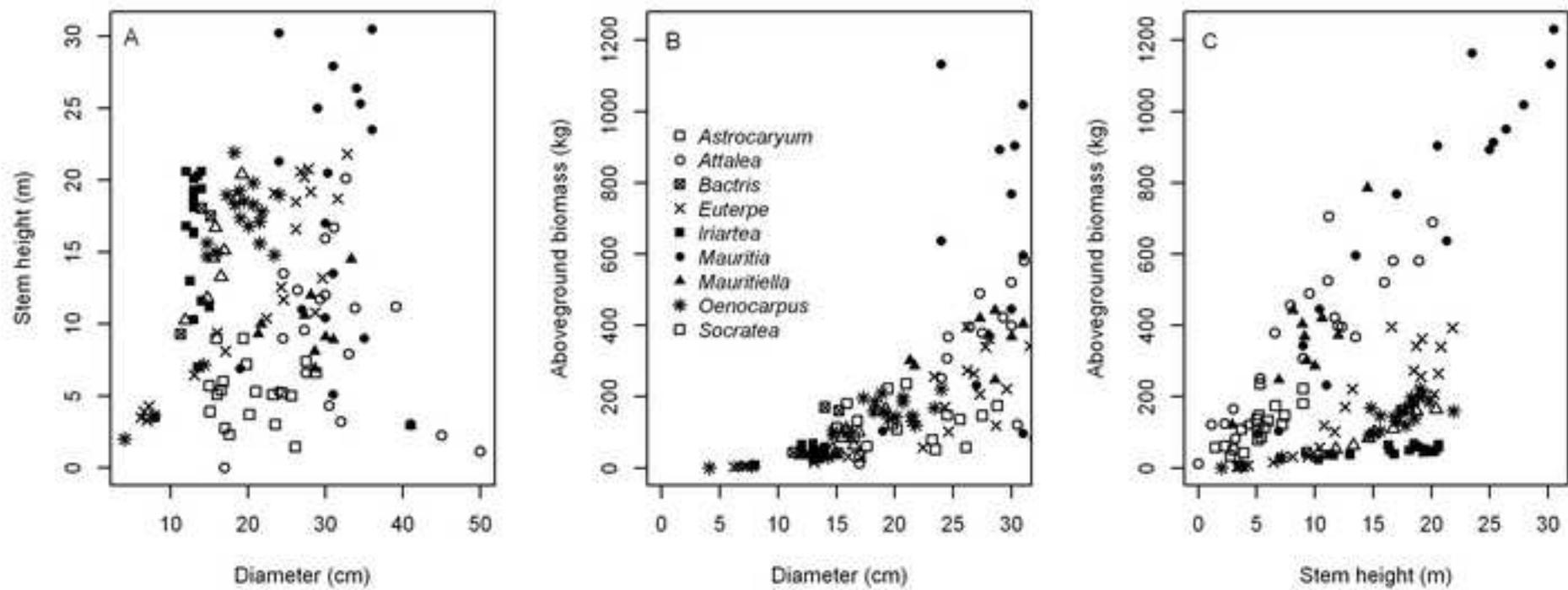


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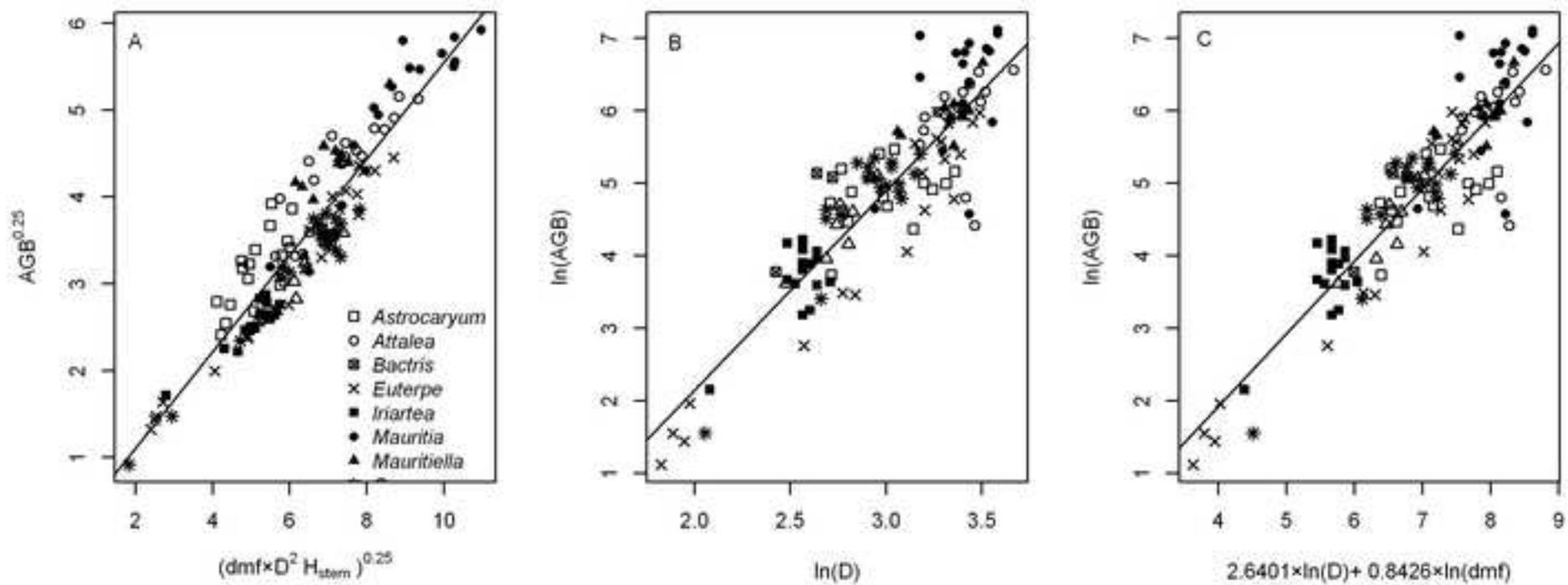
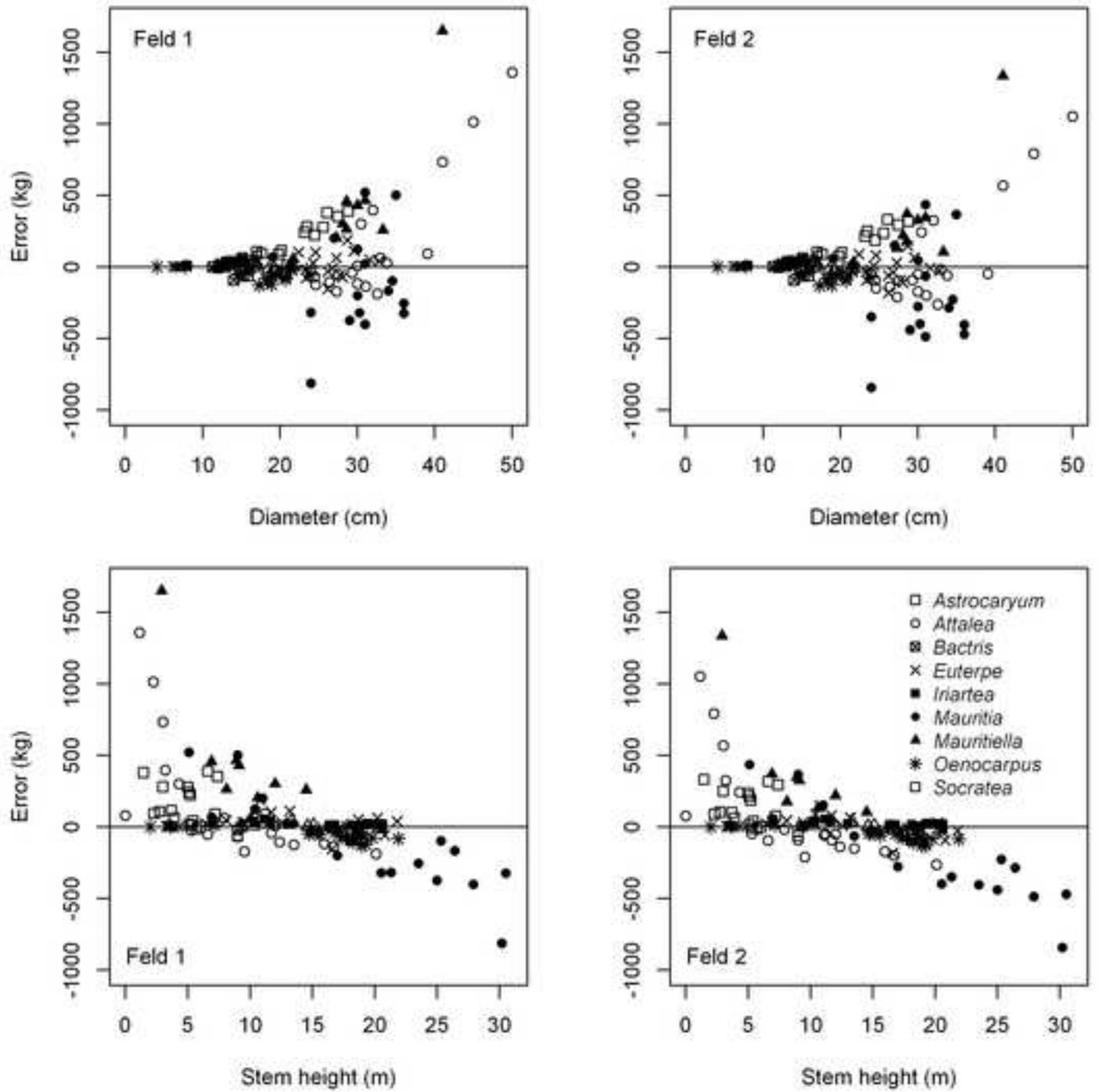


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Figure A1

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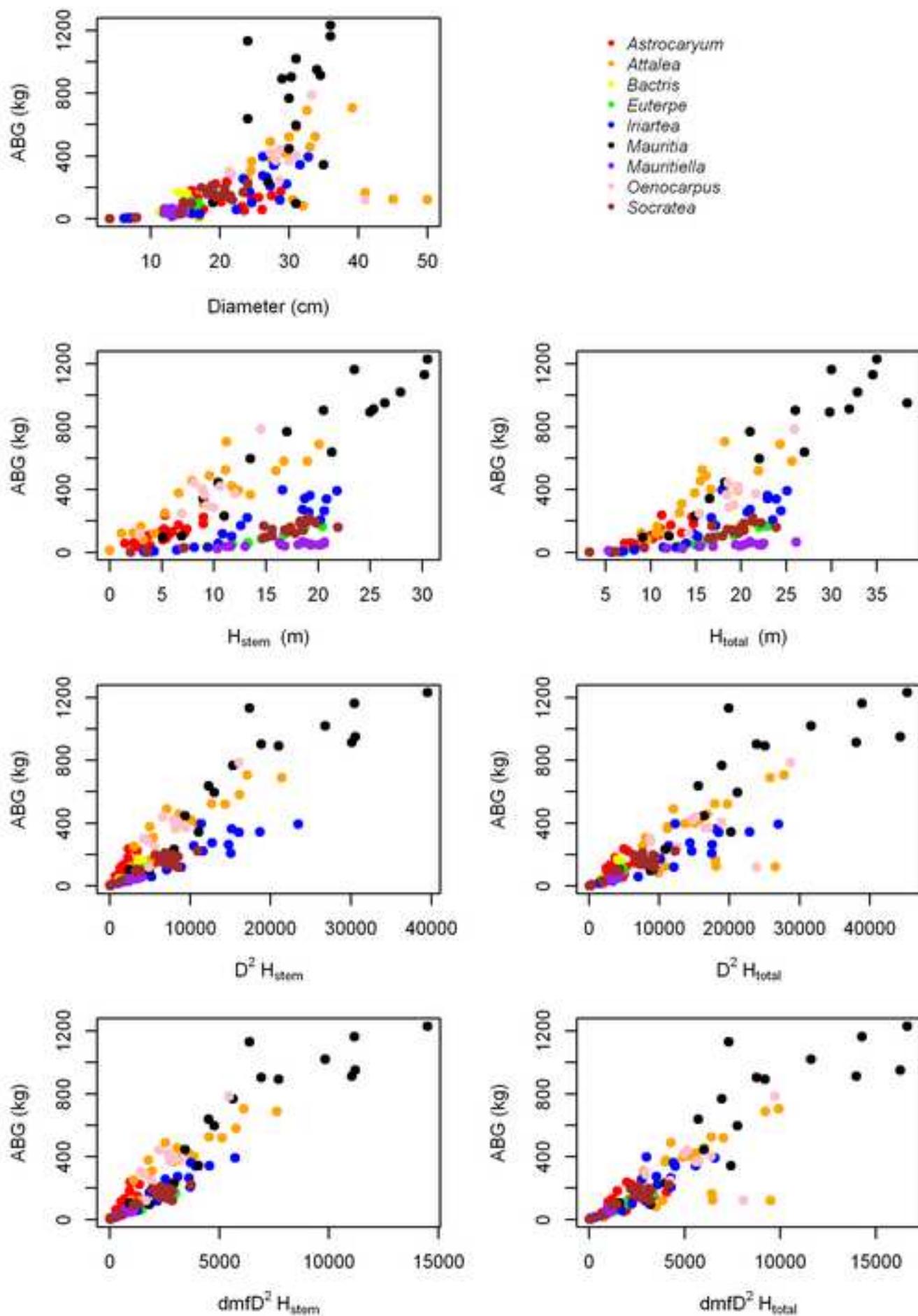


Figure A2  
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