

## INFLUENCE OF GROWTH PARAMETERS ON WOOD DENSITY OF

### *Acacia auriculiformis*

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#### ABSTRACT

Understanding the drivers of wood density variation both within a tree and between trees is important in predicting the quality of wood logs and improving this quality through adequate forestry management. This study examined the effect of the diameter growth of *Acacia auriculiformis* on its wood density variation. The study was conducted in the South of Benin in four plantations of *Acacia auriculiformis*. Near infrared spectroscopy (NIRS) method was used to predict the basic density of 225 tree wood cores of *Acacia auriculiformis*. A predicting model of the average tree density using the diameter as predictor was established. The relationship between wood density and tree diameter was best described by a linear mixed-effect model. The average wood density of trees increased with the diameter. The study concluded that the quality of the species logs can be improved through regular thinning and genetic selection.

28 **Keywords:** *Acacia auriculiformis*, log, NIRS, tree diameter, wood characteristics,

## 29 1- INTRODUCTION

30 Density is a major functional trait of wood (Hietz *et al.* 2013, Ducey 2012). It is related to the  
31 physico-mechanical characteristics and natural durability of the wood (Hietz *et al.* 2013,  
32 Pérez-Peña *et al.* 2020) and is correlated with the longevity of trees. Trees with low wood  
33 density usually have higher mortality risks (Hietz *et al.* 2013). Also, wood density represents  
34 a good indicator of forest biomass and the rate of carbon sequestered in wood (Morel *et al.*  
35 2018, Nabais *et al.* 2018). Wood density is among the most important parameters used in tree  
36 breeding programs (Alves *et al.* 2010). It varies inside species with the height and the radial  
37 position (Guller *et al.* 2012, Hietz *et al.* 2013). Hence, wood density is a combined effect of  
38 several intrinsic and extrinsic factors, including the environment, genetic factors, age and  
39 growth parameters (Nabais *et al.* 2018, Morel *et al.* 2018, Mevanarivo *et al.* 2020).

40 Tree growth is usually expressed in several ways, including height increment, and diameter  
41 growth (DeBell *et al.* 1994, Silva *et al.* 2019). Tree growth influence several characteristics of  
42 wood such as wood density. The effect of tree growth on wood density is not consistent across  
43 species. In general, fast-growing species have low density (DeBell *et al.* 1994). The increase  
44 in tree growth would lead to an increase in the ring's width, a low wood density (Wang *et al.*  
45 2000, Gapare *et al.* 2010) and an inter-annual variability in the wood density. On the contrary,  
46 DeBell *et al.* (2001) found that the wood density of *Eucalyptus saligna* increased with the  
47 diameter, particularly on nutrient rich soil whereas Jakubowski *et al.* (2020) found no  
48 significant effect of tree growth on wood density for *Betula pendula*. The relation between  
49 tree growth and wood density is apparently site-specific (e.g. climatic factors and soil water  
50 reserve) and species-specific (Bouriaud *et al.* 2005).

51 Several methods are used to estimate wood density. Direct measurements from felled trees,  
52 with wood density corresponding to the mass over the volume of a sample, provide quite

53 accurate results (Alves *et al.* 2010). Still, indirect measurements through near infrared  
54 spectroscopy (NIRS) could be used for predicting wood properties with high precision based  
55 on calibrated and validated Partial Least Squares (PLS) regression models (Alves *et al.* 2010,  
56 Cooper *et al.* 2011, Diesel *et al.* 2014). This method allows to evaluate a large amount of data  
57 very quickly and efficiently (Cooper *et al.* 2011). It also has the advantage of using samples  
58 from un-felled trees to determine wood characteristics.

59 Knowledge on the distribution of wood density in individual trees and its formation process is  
60 required to improve both the silvicultural processes and the wood production so as to obtain a  
61 wood of the desired quality (Guller *et al.* 2012, Mäkinen and Hynynen 2012). But the  
62 complexity of the wood density formation usually limits the interpretation of the models  
63 because a similar average density of two woods can result from different anatomical  
64 parameters and environmental factors. Thus, understanding the intraspecific variability of  
65 wood density can enable the identification of appropriate silvicultural practices to produce  
66 wood at a higher yield and of a better quality (Hai *et al.* 2010).

67 *Acacia auriculiformis* is native to Asia (Wickneswari and Norwati 1993) where its wood  
68 shows distinct rings. The number of rings may be inconsistent with the age of the tree  
69 (Chowdhury *et al.* 2009) and the determination of the radial variation of its wood is of definite  
70 interest. In West Africa, the species was introduced in 1980 (Tandjiekpon and Dah-Dovonon  
71 1997) mainly for firewood production but also currently include timber prospects (Tonouéwa  
72 *et al.* 2019). The species produces a wood with quite good characteristics (Hounlonon *et al.*  
73 2018, Tonouéwa *et al.* 2020). Determining the variability of its wood density within and  
74 between trees and understanding the influence of tree diameter on the wood density will allow  
75 for the improvement of the species growth parameters and wood physico-mechanical  
76 characteristics through appropriate silvicultural treatments. The main objective of this study is  
77 to identify patterns and drivers of wood density variation in *A. auriculiformis* grown in

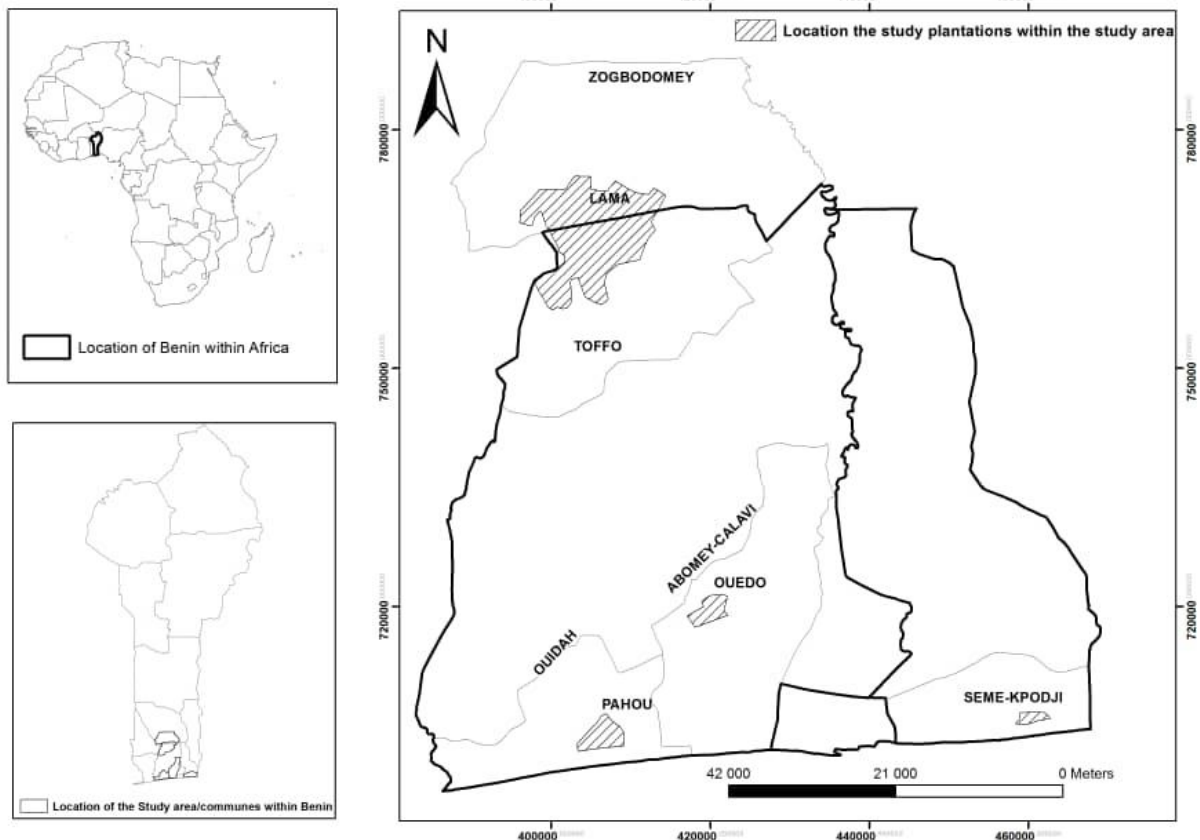
78 plantations in South Benin. More specifically it aims to (i) determine the relationship between  
79 the diameter and wood density of *A. auriculiformis*; and (ii) determine the radial variation of  
80 the wood density of *A. auriculiformis*.

81

## 82 2- MATERIALS AND METHODS

### 83 2.1 Study area

84 The study was conducted in Southern Benin around 6°22 'to 6°54'N latitude and 2°05 'to 2°8'E  
85 longitude (Figure 1) and precisely in twelve state-owned plantations of *A. auriculiformis*.  
86 These plantations were located at Lama (on vertisol), Pahou (on ferruginous soil), Ouedo (on  
87 ferralitic soil) and Sèmè-kpodji (on sandy soil). The climate is similar across the study sites  
88 (Amoussou *et al.* 2016). The average annual rainfall is 1100 mm and the average annual  
89 temperature 27 °C.



91 **Figure 1:** Map of study area and location of the study plantations.

## 92 **2.2 Data collection**

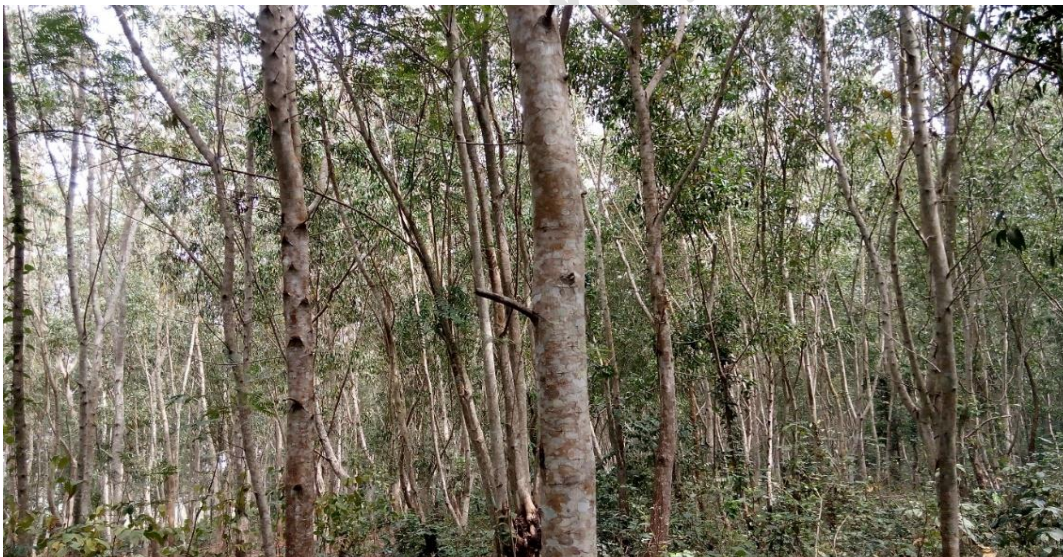
93 The data comes from twelve state-owned plantations of *A. auriculiformis* (Table 1; Figure 2).

94 A total of 255 cores of 5 mm diameter were sampled, at 1.30 m height from 255 tree  
95 individuals (1 core per tree), using a Pressler auger. The transverse surface of the cores was  
96 sanded with sandpaper (fine grits) to obtain a flat and smooth surface for the measurements.

97 The cores moisture was stabilized at 12%. The near infrared spectrometry (NIRS) method was  
98 used to determine the wood density on each core sample at 1 cm interval (Figure 3) and the  
99 radial variation of the wood density of *A. auriculiformis* was described. Near infrared spectra

100 were obtained with a Bruker Vector 22/N spectrometer run by OPUS 200 software version  
101 5.5.

102



103

104 **Figure 2:** *A. auriculiformis* plantations of 4 years old, South-Benin.

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111 **Table 1:** Wood samples distribution across soil types and age of the selected plantations.

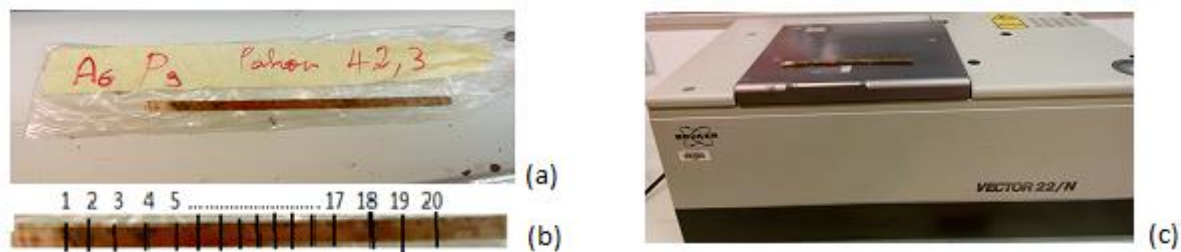
Type of soil	Ferrallitic			Ferruginous			Vertisol			Sandy			Total
Age of the selected plantation (years)	4	5	6	6	9	15	7	9	11	27	27	29	
Thinning regime						a				b	c	UH	
Number of trees/hectare (Tonouéwa <i>et al.</i> 2019)	970	910	827	790	407	323	707	530	420	190	293	187	
Number of plots	3	3	3	3	3	3	3	3	3	3	3	3	36
Number of cores extracted	21	21	21	21	21	21	22	22	21	21	21	21	253
Cores with validated measures	20	20	21	14	16	12	21	22	21	21	18	19	225

112 UH=Uncontrolled Harvesting; a= thinning at 9 years; b= thinning at 7 years+ Uncontrolled Harvesting; c=  
 113 Thinning at 14 years + Uncontrolled Harvesting

114 On each core, the number of measures in the heartwood was  $\leq 20$  (20<sup>th</sup> radial position, Figure  
 115 3 and 6).

116 After removing broken and poorly preserved cores from the samples and after removing  
 117 outliers, 225 cores translating into 2869 measurements (430 on sapwood and 2439 on  
 118 heartwood) were validated (Table 1). However, the wood density of *A. auriculiformis* was  
 119 predicted using only the 2439 measurements on heartwood because wood density is always  
 120 higher in heartwood than in sapwood (Githiomi and Kariuki 2010). This is due to the high  
 121 proportions of carbon, lignin and extractives in the heartwood (de Aza *et al.* 2011, Bertaud  
 122 and Holmbom 2004). In addition, heartwood is the most interesting part of the tree for use as  
 123 timber.

124 For the prediction of wood density, we used a pre-established NIRS model of the basic  
125 density of *A. auriculiformis* in South Benin (Tonouéwa *et al.* 2020). The model had a root-  
126 mean-square error (RMSE) of 50.33 kg/m<sup>3</sup> (R<sup>2</sup> = 0.75) for the prediction model and 45.88  
127 kg/m<sup>3</sup> (R<sup>2</sup> = 0.79) for the calibrated model. The Unscrambler software, version 9.7 CAMO  
128 (The Unscrambler 2007) was used to estimate the basic density and the standard deviation for  
129 each measured point. The value of the basic density at a given point on a wood core is an  
130 average of 16 measurements with the spectrometer at this point. The standard deviation of the  
131 16 measurements at each point was  $\geq 60 \text{ Kg/m}^3$ .



132  
133 **Figure 3:** *A. auriculiformis* wood core packed within the field with codes indicating the  
134 location, the tree and tree diameter (a); a wood core marked for measurements (b) and the  
135 spectrometer used to make the measurements (c).

136 The basic density of *A. auriculiformis* trees was evaluated as a function of the radial position  
137 in the tree and as the ratio between the dry mass of a sample and its saturated volume  
138 (Rybníček *et al.* 2012, Diesel *et al.* 2014).

### 139 2.3 Statistical analysis

140 A first exploratory analysis of the predicted wood density of *A. auriculiformis* on cores was  
141 done. Samples with standard deviations greater than or equal to 200 kg/m<sup>3</sup> (i.e. 1.5 times the  
142 interquartile range above the third quartile or below the first quartile; Crawley 2007) were  
143 considered as outliers and excluded from the batch for the subsequent analyses.

144 To assess the relationship between the wood density and tree diameter, several Linear and  
145 Non-Linear Mixed Effects Models (NLME) were tested in R.3.5.2 (R Core Team 2018) with

146 the nlme package (Pinheiro *et al.* 2018). The random effects were: plots nested within zone of  
147 data collection and the fixed effect was tree diameter. Based on the inspection of the  
148 scatterplots and models previously used for explaining the relation between tree diameter or  
149 tree age and wood density, a linear (Silva *et al.* 2019) and four nonlinear functions (Table 3)  
150 were assumed to potentially fit well the data. The nonlinear functions include the second-  
151 degree polynomial function (Githiomi and Kariuki 2010), the first exponential function (Oddi  
152 *et al.* 2019), the Michaelis-Menten asymptotic function (Oddi *et al.* 2019) and the second  
153 exponential function (Oddi *et al.* 2019).

154 The appropriate model was selected based on the Akaike Information Criterion (AIC) (Akaike  
155 1973) and the random effect. The best model is the one that minimizes the AIC value (Chave  
156 *et al.* 2005) and shows a high random effect (i.e. variance in the wood density explained by  
157 the random effect). The general equation of the NLME model developed in R.3.5.2 is showed  
158 below:

```
159 Model=nlme(ER,  
160     data=WDdata,  
161     fixed=a+b+c~1,  
162     random=a~1|zone/plot,  
163     start=c(a=u, b=v, c=w))
```

164 ER: the linear or nonlinear regression equation used (Table 2); WDdata: the database  
165 containing the dependant (wood density) and the explanatory variables;  $a + b + c \sim 1$ : the  
166 parameters of the function (Table 2) used for the fixed effect of the model;  $a \sim 1 | \text{zone} / \text{plot}$ :  
167 the random effect;  $u, v, w$ : the list of initial estimates for the values of  $a, b$ , and  $c$  respectively.

168 When the function has two parameters ( $a, b$ ), the fixed effect takes the form  $a + b \sim 1$ .



169 Regarding the type of linear or nonlinear relationship, parameters  $a$ ,  $b$  and  $c$  were calculated  
170 with an adjustment in R (R.3.5.2).

171 Concerning the linear function, parameters  $a$  and  $b$  were determined by solving in R the  
172 following matrix equation (1):

$$\begin{cases} Y1 = ax1 + b \\ Y2 = ax2 + b \end{cases} \quad (1)$$

173 Considering the second-degree polynomial function, parameters  $a$ ,  $b$ , and  $c$  were calculated by  
174 solving in R the following matrix equation (2):

$$A \begin{pmatrix} y1 \\ y2 \\ y3 \end{pmatrix} = B \begin{pmatrix} a \\ b \\ c \end{pmatrix} * C \begin{pmatrix} x1^2 & x1 & 1 \\ x2^2 & x2 & 1 \\ x3^2 & x2 & 1 \end{pmatrix} \quad (2)$$

175 Regarding the first exponential function (3),  $a$  and  $b$  were calculated as follow:

$$176 \quad b = \log\left(\frac{y1 \cdot x2}{y2 \cdot x1}\right) / (x2 - x1) \quad \text{and} \quad a = \frac{y1}{x1 e^{-bx1}} \quad (3)$$

177 For the Michaelis-Menten asymptotic function,  $a$  and  $b$  were calculated using the `SSmicmen`  
178 function of the package `stats`.

179 For the second exponential function (4),  $a$  and  $b$  were calculated as follow:

$$180 \quad b = \log\left(\frac{y2}{y1}\right) / \log\left(\frac{x2}{x1}\right) \quad \text{and} \quad a = \frac{y1}{x1^b} \quad (4)$$

181 After the selection of the best model, normality quantile, modelling variance, correlation  
182 structure, residuals wood density as function of tree diameter, and a plot of the predicted vs  
183 observed wood density were analysed to evaluate the model assumptions.

184 In addition, the relations between the wood density and the age of the plantation, and the  
185 radial position in the tree were determined using a mixed effect model in R.3.5.2 (R Core  
186 Team 2018) with the `lmerTest` package. The average values of the wood density were  
187 eventually regressed against the radial position for each age of plantation.

188 **3- RESULTS**

189 **3.1- Variation in wood density of *A. auriculiformis* in relation to tree diameter**

190 Among the models tested (Table 2) for expressing the wood density as a function of tree  
 191 diameter, the linear mixed effect model was the most suitable for *A. auriculiformis* (low AIC,  
 192 high random effect). This model validity is restricted to *A. auriculiformis* trees with at least 10  
 193 cm diameter at breast height and growing in conditions similar to those of south Benin. The  
 194 model is in the form of:  $WD = a * D + b$ , with D = tree diameter, WD = wood density,  
 195  $a = 1.2335$ ,  $b = 537.6931$  from solving Equation (1). With a Root-Mean-Square Error (RMSE)  
 196 of 51.79 Kg/m<sup>3</sup> and a Bias of 0.2%, the quality of the model is suitable.

197 **Table 2:** Results from the five linear and nonlinear models tested for the prediction of *A.*  
 198 *auriculiformis* wood density (WD) in kilograms per cubic meter (Kg/m<sup>3</sup>) with tree diameter  
 199 (D) in centimetre (cm) as the predictor.

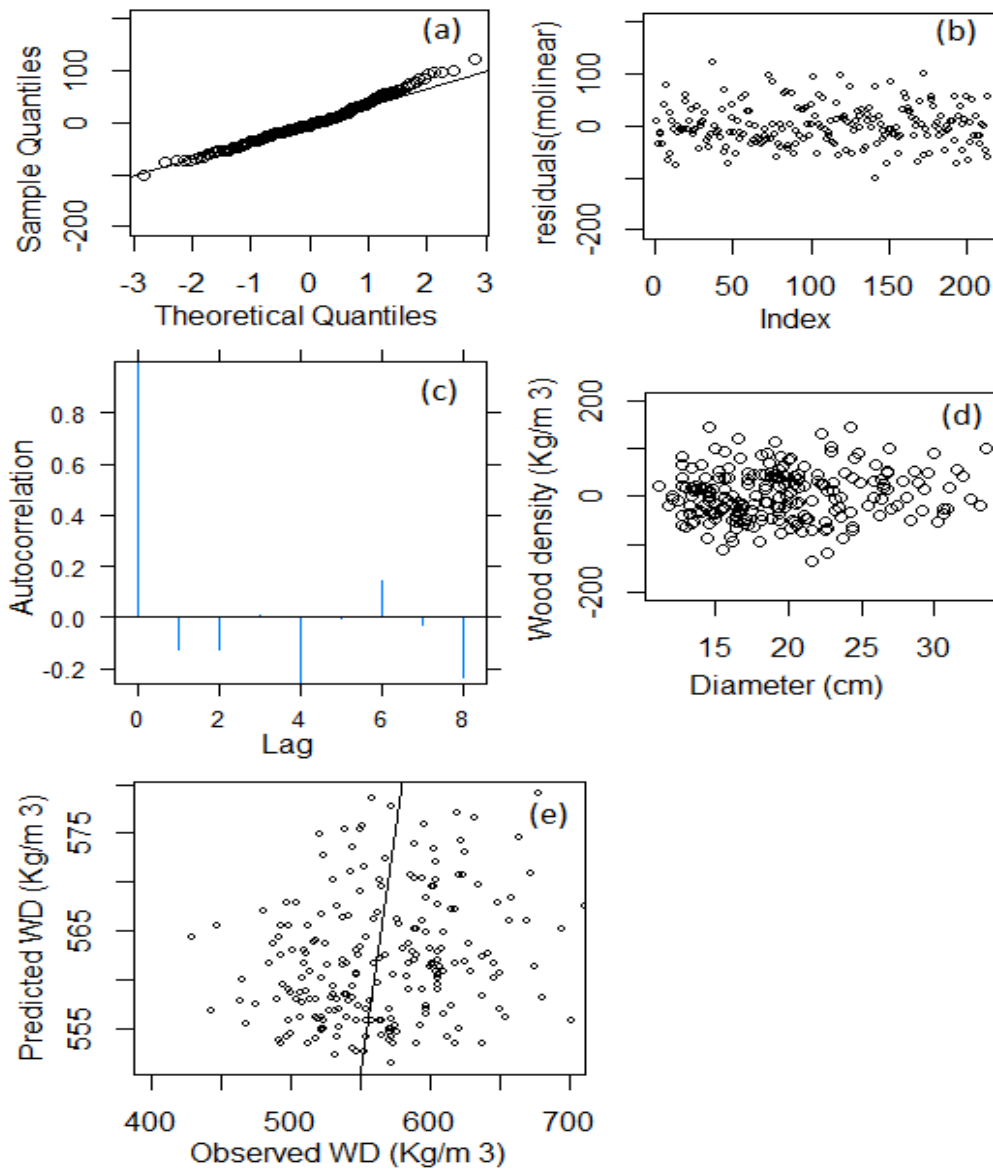
Model	Equation	Fixed coefficient	Random effect	AIC	log-Likelihood	ΔAIC
Linear	$WD = a * D + b$	a=1,23 b= 537,69	38,21	2218,89	-1104,447	0
First exponential	$WD = a * D * e^{(-b*D)}$	a=71,48 b=0,05	4,57	2251,77	-1120,887	32,88
Polynomiale	$WD = a * D^2 + b * D + c$	a= -0,20 b=9,63 c= 455,68	0,06	2251,15	-1119,575	32,26
Asymptotique	$WD = a * \frac{D}{1 + (b * D)}$	a= 586,02 b= 0,79	34,44	2226,91	-1108,453	8,02
Second exponential	$WD = e^{b*D}$	a=533,51 b= 0,003	30,73	2228,1	-1109,05	9,21

200 *WD = wood density; a, b and c are the parameters of the function used for the fixed effect of*  
 201 *each model (see Table 2); AIC = Akaike Information Criterion; ΔAIC = difference between*  
 202 *the AIC of the best model (smallest AIC) and each of the alternate models, for ease of*  
 203 *comparison.*

204

205 The characteristics of the linear model obtained indicate a distribution of the data close to  
 206 normal (Figure 4-a), a homogeneity of the variance (Figure 4-b), a weak autocorrelation  
 207 structure (Figure 4-c), and a coherent and uniform distribution of wood density residues as a

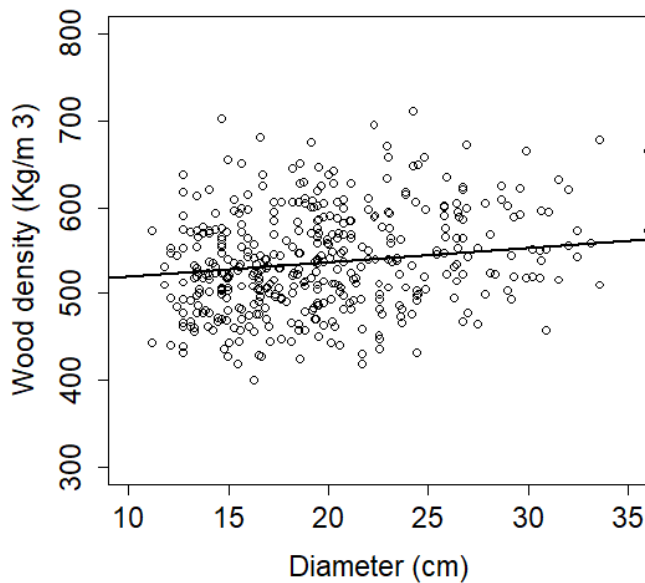
208 function of the diameter of the trees (Figure 4-d). The scatter plot between the real data and  
209 those predicted is not perfect (Figure 4-e), however we observe that the predicted values are  
210 averages of the real values.



211

212 **Figure 4:** Characteristics of the linear mixed effect model of wood density as a function of  
213 tree diameter for *A. auriculiformis* in Benin: a- Normality quantile, b-modelling variance, c-  
214 correlation structure, d- Residual wood density as a function of tree diameter, e- Predicted  
215 Wood density vs observed wood density, 45° line.

216 In general, wood density of *A. auriculiformis* increased with the diameter (Figure 5). There is  
217 also a significant and positive correlation between tree diameter and wood density of *A.*  
218 *auriculiformis* ( $r = 0.23$ ;  $p\text{-value} = 0.0006^{**}$ ).



219

220 **Figure 5:** Wood density of *A. auriculiformis* as a function of the tree diameter.

221 **3.2- Intra-species and intra-tree variations in the wood density of *A. auriculiformis***

222 Wood density varies from one tree to another (Table 3), and the random effect of this  
 223 parameter explains 25.7% of the variation of *A. auriculiformis* wood density in the study  
 224 plantations.

225 **Table 3:** Results of the mixed effect model predicting wood density as a function of the  
 226 plantation age and the radial position of the wood in the tree.

Random effects		Variance	Std.Dev.		
	Tree	917,9	30,3		
	Residual	3562,7	59,69		
Fixed effects		Estimate	Std.Error	t.value	p.z
	(Intercept)	566,42	5,27	107,57	0,00***
	Age of plantation	26,11	1,32	19,3	0,00***
	Radial position	-1,58	1,23	-1,28	0,20ns
	Age of plantation:Radial position	0,20	1,22	0,17	0,87ns

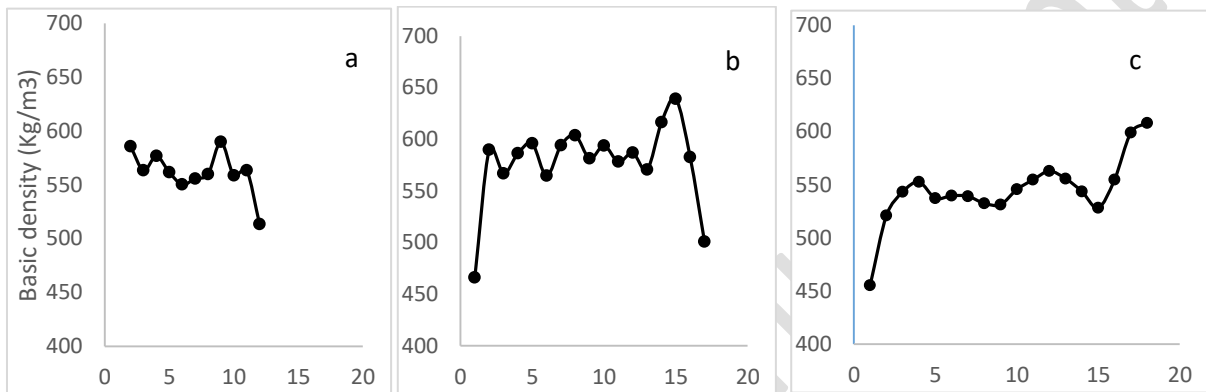
227 \*\*\*Significant difference at 0,1% level; ns = no significant

228 In general, the relation between the wood density of *A. auriculiformis* and the radial position  
 229 of the wood in a tree was not significant (Table 3, Figure 6). The radial pattern of the wood

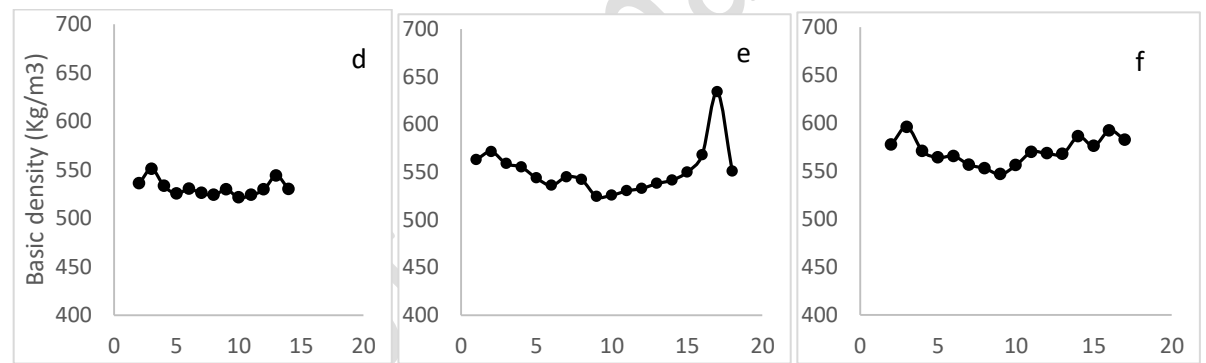
230 density of the species did not show the demarcation between juvenile and mature wood. Still,  
231 we noted a sawtooth radial variability of wood density, an increase in the density of the wood  
232 outwards until a wood density peak is obtained, and a decrease in the wood density towards  
233 the ends of the tree. On the contrary, the age of the plantation has a significant and positive  
234 effect on wood density (Table 3).

235

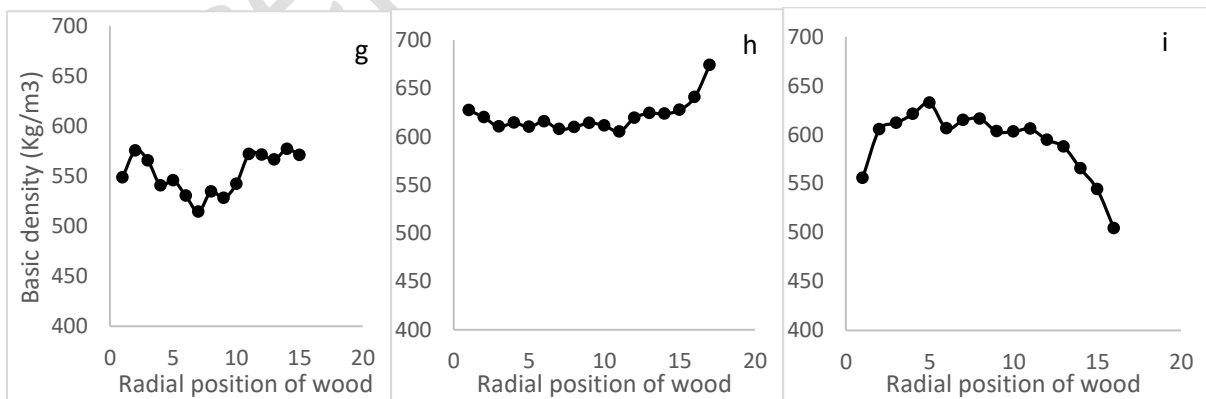
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238



239 **Figure 6:** Variations in the wood density of *A. auriculiformis* as a function of the radial  
240 position of the wood and the age of the plantation: a = 4 years old plantation; b = 5 years old  
241 plantation; c = 6 years old plantation; d= 7 years old plantation; e = 9 years old plantation; f =

242 11 years old plantation; g = 15 years old plantation; h = 27 years old plantation; i = 29 years  
243 old plantation.

244

#### 245 4- DISCUSSION

246 The best model for predicting the wood density of *A. auriculiformis* as a function of diameter  
247 is a linear function, and wood density increased with tree diameter within the range of  
248 diameters considered for the sample trees in this study (10-35 cm diameter at 1.30 cm). Linear  
249 models have been widely applied for the estimation of wood density in various settings and  
250 for various species. For example, an estimation of the wood density of Cerrado species (e.g.  
251 *Luehea paniculata*, *Terminalia fagifolia*) was obtained in Brazil with a lineal mixed model  
252 (Silva *et al.* 2019) and for several species in Madagascar (Ramananantoandro *et al.* 2016).  
253 Similarly, a linear mixed-effect model was used to estimate the wood density of *Quercus*  
254 *petraea* as a function of growth parameters and site quality (Guilley *et al.* 2004). Wood  
255 density increased with diameter. Forestry focused on accelerating the growth of *A.*  
256 *auriculiformis* trees is generally recommended for the rapid obtaining of good wood quality  
257 and large diameters. As such, regular thinning could be effective in increasing the value of *A.*  
258 *auriculiformis* plantations as shown elsewhere (e.g. Huong *et al.* 2020, Wiersum and Ramlan  
259 1982). However, climate and soil conditions can alter trees response to thinning and post-  
260 thinning diameter growth. Thus, further experiments are needed to evaluate the response of  
261 the species to frequent thinning practices in the specific conditions of South Benin.

262 The correlation between the diameter of the trees and wood density, although significant, was  
263 moderate in this study and suggests that the diameter of the tree alone does not explain  
264 entirely the variation in wood density of *A. auriculiformis*. Indeed, the relationship between  
265 wood density and growth rate in a species is generally influenced by environmental,  
266 silvicultural and genetic factors (Zobel and van Buijtenen 1989, Zhang and Morgenstern  
267 1995). Consequently, contrasting relations between tree growth and wood density are found in

268 the literature. For example, in Madagascar, Ramananantoandro *et al.* (2016) found that tree  
269 diameter did not significantly affect wood density for native hardwood species. Similarly, a  
270 lack of correlation between tree growth and wood density was observed for *Eucalyptus*  
271 *globulus* in Portugal (Quilhó and Pereira 2001) and for black spruce (*Picea mariana*) in  
272 Canada (Hall 1984), while DeBell *et al.* (2001) found a negligible influence of growth rate on  
273 wood density for *Eucalyptus saligna* in Hawaii. In contrast, Roque and Fo (2007) and Boyle *et*  
274 *al.* (1988) reported negative correlations between wood density and growth traits, respectively  
275 for *Gmelina arborea* in Costa Rica and for black spruce (*Picea mariana*) in Canada. In the  
276 present study, the moderate positive correlation between *A. auriculiformis* wood density and  
277 tree diameter suggests that diameter growth can be improved with a small gain in the species  
278 wood density. This is interesting as most wood mechanical properties are closely related to  
279 wood density.

280 In addition, wood density varies from one tree to another within same plot. This variability of  
281 the wood density can find an explanation in the silvicultural practices in Benin. *A.*  
282 *auriculiformis* plantations in Benin are set up from seeds from mother plants chosen mostly  
283 randomly, or based on the availability and accessibility of the trees (survey in South Benin).  
284 The probability of a large genetic variability between trees on the same plot is thus high. This  
285 stress the importance of genetic selection, as the structural characteristics of *A. auriculiformis*  
286 wood, such as the wood density, are heritable. The selection of good quality parent material  
287 are particularly critical for improving the properties of new plantations (Hai *et al.* 2010,  
288 Chowdhury *et al.* 2012, Nabais *et al.* 2018).

289 Regarding the intra-tree variation of wood density, we found no significant relation between  
290 the radial positions of the wood and the wood density for *A. auriculiformis*. Wood density  
291 being the main technological parameter of wood, its variation within trees and the magnitude  
292 of the variation provide information on the quality of the logs produced (Guilley *et al.* 2004).

293 As such, our results indicate that the characteristics of *A. auriculiformis* logs are  
294 heterogeneous. Bouriaud *et al.* (2005) linked the variability of the radial density of wood to  
295 the differences in radial growth of trees. The latter is influenced by climatic variations,  
296 thinning and soil fertility (Mäkinen and Hynynen 2012, Hietz *et al.* 2013, Miranda and Pereira  
297 2015, Nabais *et al.* 2018). For *A. auriculiformis* these site-specific constraints could relate to  
298 climatic conditions and tree spacing.

299 Although the intra-tree variation of wood density was not significant in this study, the  
300 following patterns could be drawn from the data: (i) a sawtooth radial variability of wood  
301 density that possibly reflect the succession of rainy and dry seasons at the study sites; (ii) an  
302 increase in the density of the wood outwards until a wood density peak is obtained, suggesting  
303 that the wood formed during the last years of the tree's life is of higher density; (iii) a decrease  
304 in the wood density from pith to bark, which is potentially linked to the presence of sapwood.  
305 Similar variations of radial characteristics were also recorded on *A. auriculiformis* produced  
306 in Asia (Chowdhury *et al.* 2012) as well as for other timber species (Hietz *et al.* 2013).

307 The low radial variability in the wood density indicates that for estimation of wood biomass  
308 of *A. auriculiformis* and carbon content in the wood, the samples wood can be taken at any  
309 radial position in the tree (Chave *et al.* 2006, Hietz *et al.* 2013). This low radial variability in  
310 the wood density also makes it possible to predict low constraints growing of the tree (Curran  
311 *et al.* 2008, Nock *et al.* 2009, Nabais *et al.* 2018) predicting good ecological and biological  
312 conditions of growing and better quality of the log.

## 313 **5- CONCLUSIONS**

314 In this study, the best model for predicting the wood density of *A. auriculiformis* as a function  
315 of diameter is a linear function, and wood density increased with tree diameter. The moderate  
316 positive correlation between *A. auriculiformis* wood density and tree diameter suggests that



317 diameter growth can be improved with a small gain in the species wood density. The  
318 suggested opportunities for improvement include the selection of good quality parent and the  
319 practice of frequent and regular thinning to reach desirable diameters and produce high-  
320 density timber. However, further experiments should evaluate the response of the species to  
321 frequent thinning in the specific climate and soil conditions of *A. auriculiformis* grown in  
322 South Benin. The study also highlights the need for reducing the intra-tree variability in wood  
323 quality, which could also be achieved through improved tree breeding.

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