

## 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.

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

## 28 1- INTRODUCTION

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

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

50 Several methods are used to estimate wood density. Direct measurements from felled trees,  
51 with wood density corresponding to the mass over the volume of a sample, provide quite  
52 accurate results (Alves *et al.* 2010). Still, indirect measurements through near infrared

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

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

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

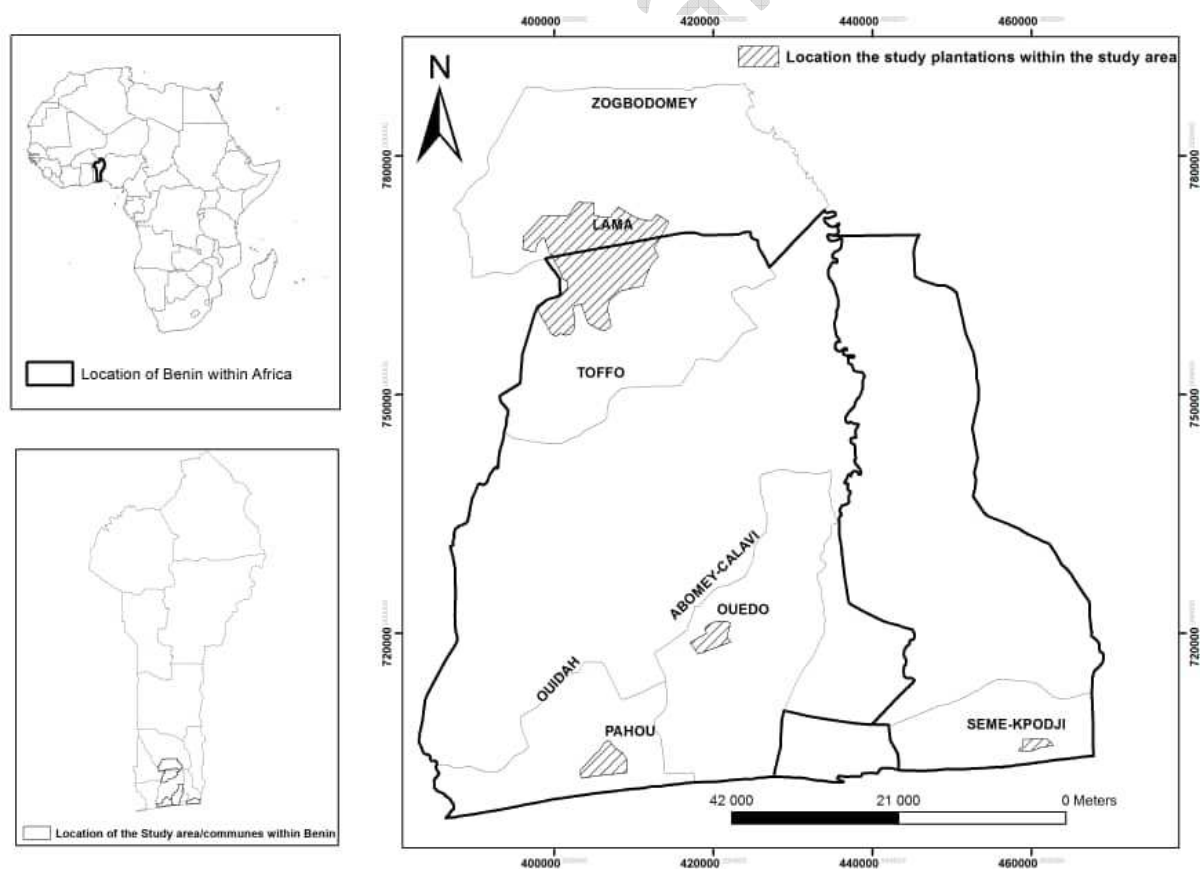
78 wood density of *A. auriculiformis*; and (ii) determine the radial variation of the wood density  
79 of *A. auriculiformis*.

80

## 81 2- MATERIALS AND METHODS

### 82 2.1 Study area

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

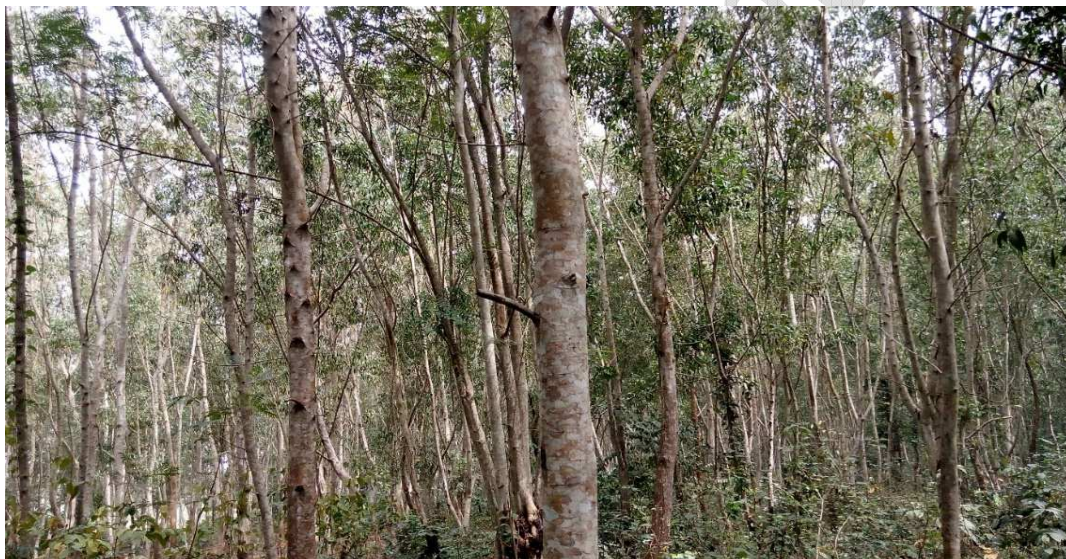


89 **Figure 1:** Map of study area and location of the study plantations.

90 **2.2 Data collection**

91 The data comes from twelve state-owned plantations of *A. auriculiformis* (Table 1; Figure 2).  
92 A total of 255 cores of 5 mm diameter were sampled, at 1.30 m height from 255 tree individuals  
93 (1 core per tree), using a Pressler auger. The transverse surface of the cores was sanded with  
94 sandpaper (fine grits) to obtain a flat and smooth surface for the measurements. The cores  
95 moisture was stabilized at 12%. The near infrared spectrometry (NIRS) method was used to  
96 determine the wood density on each core sample at 1 cm interval (Figure 3) and the radial  
97 variation of the wood density of *A. auriculiformis* was described. Near infrared spectra were  
98 obtained with a Bruker Vector 22/N spectrometer run by OPUS 200 software version 5.5.

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

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108 **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

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

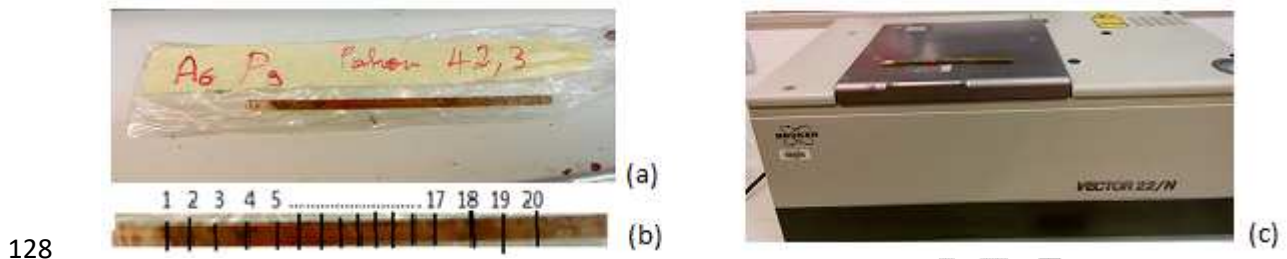
111 On each core, the number of measures in the heartwood was  $\leq 20$  (20<sup>th</sup> radial position, Figure  
 112 3 and 6).

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

120 For the prediction of wood density, we used a pre-established NIRS model of the basic density  
 121 of *A. auriculiformis* in South Benin (Tonouéwa *et al.* 2020). The model had a root-mean-square  
 122 error (RMSE) of 50.33 kg/m<sup>3</sup> ( $R^2 = 0.75$ ) for the prediction model and 45.88 kg/m<sup>3</sup> ( $R^2 = 0.79$ )



123 for the calibrated model. The Unscrambler software, version 9.7 CAMO (The Unscrambler  
124 2007) was used to estimate the basic density and the standard deviation for each measured point.  
125 The value of the basic density at a given point on a wood core is an average of 16 measurements  
126 with the spectrometer at this point. The standard deviation of the 16 measurements at each point  
127 was  $\geq 60 \text{ Kg/m}^3$ .



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

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

### 135 2.3 Statistical analysis

136 A first exploratory analysis of the predicted wood density of *A. auriculiformis* on cores was  
137 done. Samples with standard deviations greater than or equal to  $200 \text{ kg/m}^3$  (i.e. 1.5 times the  
138 interquartile range above the third quartile or below the first quartile; Crawley 2007) were  
139 considered as outliers and excluded from the batch for the subsequent analyses.

140 To assess the relationship between the wood density and tree diameter, several Linear and Non-  
141 Linear Mixed Effects Models (NLME) were tested in R.3.5.2 (R Core Team 2018) with the  
142 nlme package (Pinheiro *et al.* 2018). The random effects were: plots nested within zone of data  
143 collection and the fixed effect was tree diameter. Based on the inspection of the scatterplots and  
144 models previously used for explaining the relation between tree diameter or tree age and wood

145 density, a linear (Silva *et al.* 2019) and four nonlinear functions (Table 3) were assumed to  
146 potentially fit well the data. The nonlinear functions include the second-degree polynomial  
147 function (Githiomi and Kariuki 2010), the first exponential function (Oddi *et al.* 2019), the  
148 Michaelis-Menten asymptotic function (Oddi *et al.* 2019) and the second exponential function  
149 (Oddi *et al.* 2019).

150 The appropriate model was selected based on the Akaike Information Criterion (AIC) (Akaike  
151 1973) and the random effect. The best model is the one that minimizes the AIC value (Chave  
152 *et al.* 2005) and shows a high random effect (i.e. variance in the wood density explained by the  
153 random effect). The general equation of the NLME model developed in R.3.5.2 is showed  
154 below:

```
155 Model=nlme(ER,  
156     data=WDdata,  
157     fixed=a+b+c~1,  
158     random=a~1|zone/plot,  
159     start=c(a=u, b=v, c=w))
```

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

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

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

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



169 
$$\begin{cases} Y1 = ax1 + b \\ Y2 = ax2 + b \end{cases} \quad (1)$$

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

172 
$$A \begin{pmatrix} y1 \\ y2 \\ y3 \end{pmatrix} = B \begin{pmatrix} a \\ b \\ c \end{pmatrix} * C \begin{pmatrix} x1^2 & x1 & 1 \\ x2^2 & x2 & 1 \\ x3^2 & x2 & 1 \end{pmatrix} \quad (2)$$

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

174 
$$b = \log\left(\frac{y1.x2}{y2.x1}\right) / (x2 - x1) \text{ and } a = \frac{y1}{x1e^{-bx1}} \quad (3)$$

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

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

178 
$$b = \log\left(\frac{y2}{y1}\right) / \log\left(\frac{x2}{x1}\right) \text{ and } a = \frac{y1}{x1^b} \quad (4)$$

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

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

### 186 3- RESULTS

#### 187 3.1- Variation in wood density of *A. auriculiformis* in relation to tree diameter

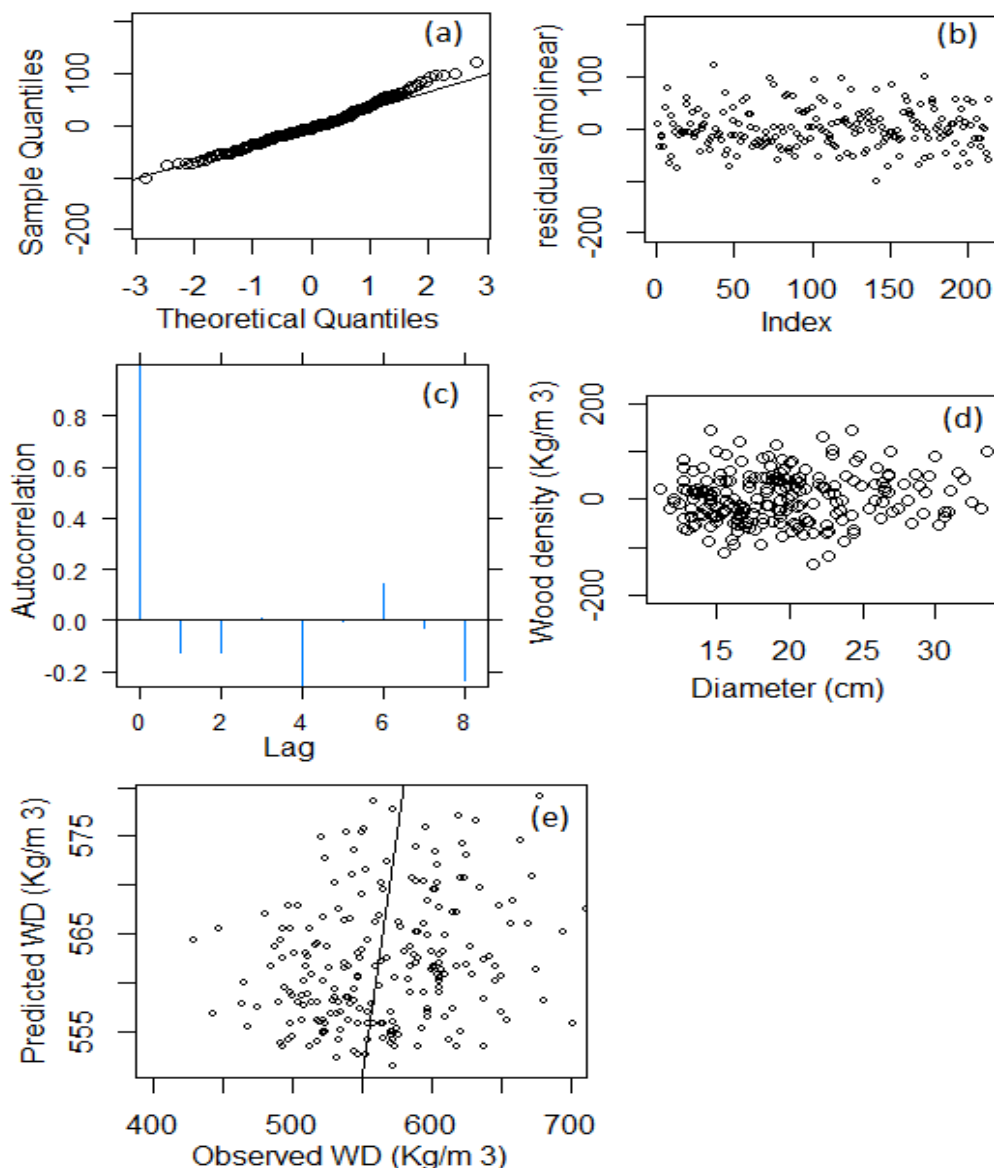
188 Among the models tested (Table 2) for expressing the wood density as a function of tree  
 189 diameter, the linear mixed effect model was the most suitable for *A. auriculiformis* (low AIC,  
 190 high random effect). This model validity is restricted to *A. auriculiformis* trees with at least 10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

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

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

201  
 202 The characteristics of the linear model obtained indicate a distribution of the data close to  
 203 normal (Figure 4-a), a homogeneity of the variance (Figure 4-b), a weak autocorrelation  
 204 structure (Figure 4-c), and a coherent and uniform distribution of wood density residues as a  
 205 function of the diameter of the trees (Figure 4-d). The scatter plot between the real data and  
 206 those predicted is not perfect (Figure 4-e), however we observe that the predicted values are  
 207 averages of the real values.



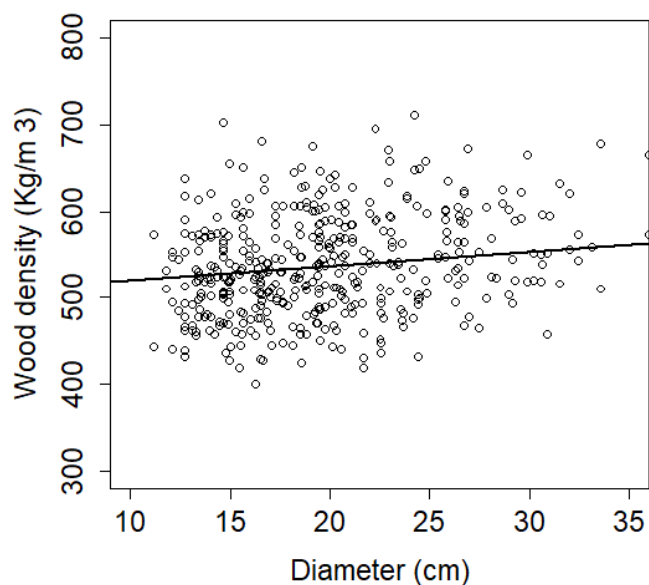
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209 **Figure 4:** Characteristics of the linear mixed effect model of wood density as a function of tree  
210 diameter for *A. auriculiformis* in Benin: a- Normality quantile, b-modelling variance, c-  
211 correlation structure, d- Residual wood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

213 In general, wood density of *A. auriculiformis* increased with the diameter (Figure 5). There is

214 also a significant and positive correlation between tree diameter and wood density of *A.*

215 *auriculiformis* ( $r = 0.23$ ;  $p\text{-value} = 0.0006^{**}$ ).



216

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

218 **3.2- Intra-species and intra-tree variations in the wood density of *A. auriculiformis***

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

221 **Table 3:** Results of the mixed effect model predicting wood density as a function of the  
 222 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

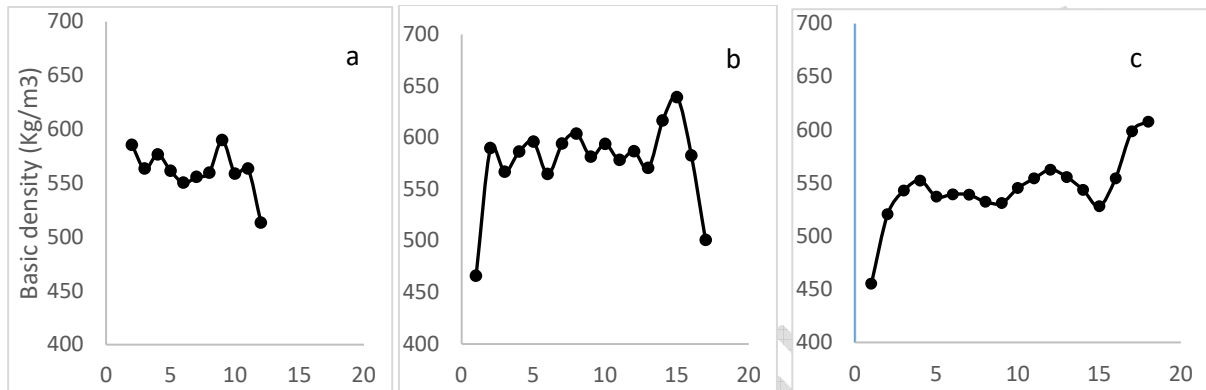
223 \*\*\*Significant difference at 0,1% level; ns = no significant

224 In general, the relation between the wood density of *A. auriculiformis* and the radial position of  
 225 the wood in a tree was not significant (Table 3, Figure 6). The radial pattern of the wood density  
 226 of the species did not show the demarcation between juvenile and mature wood. Still, we noted

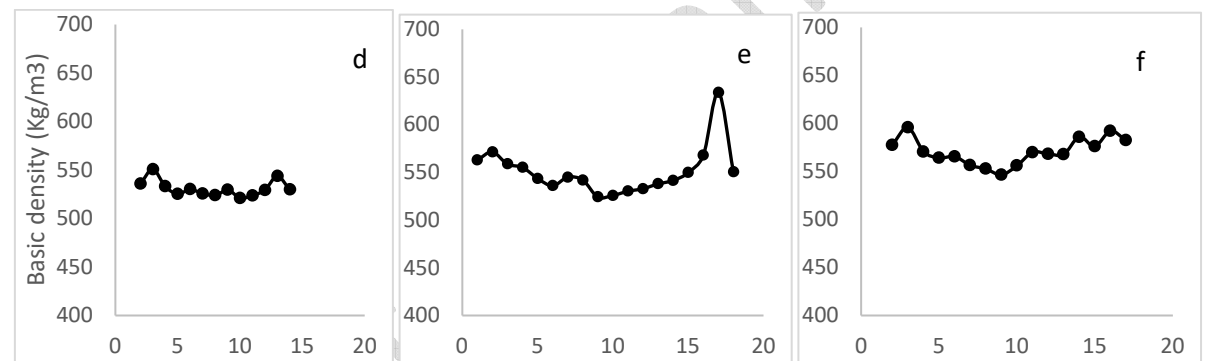
227 a sawtooth radial variability of wood density, an increase in the density of the wood outwards  
228 until a wood density peak is obtained, and a decrease in the wood density towards the ends of  
229 the tree. On the contrary, the age of the plantation has a significant and positive effect on wood  
230 density (Table 3).

231

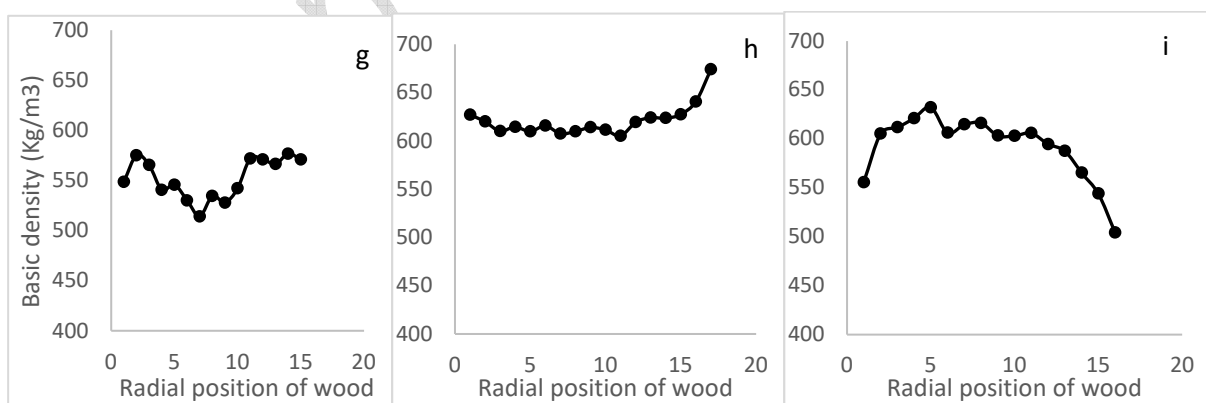
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233



234



235 **Figure 6:** Variations in the wood density of *A. auriculiformis* as a function of the radial position  
236 of the wood and the age of the plantation: a = 4 years old plantation; b = 5 years old plantation;  
237 c = 6 years old plantation; d = 7 years old plantation; e = 9 years old plantation; f = 11 years old  
238 plantation; g = 15 years old plantation; h = 27 years old plantation; i = 29 years old plantation.

239

#### 240 4- DISCUSSION

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

257 The correlation between the diameter of the trees and wood density, although significant, was  
258 moderate in this study and suggests that the diameter of the tree alone does not explain entirely  
259 the variation in wood density of *A. auriculiformis*. Indeed, the relationship between wood  
260 density and growth rate in a species is generally influenced by environmental, silvicultural and  
261 genetic factors (Zobel and van Buijtenen 1989, Zhang and Morgenstern 1995). Consequently,  
262 contrasting relations between tree growth and wood density are found in the literature. For  
263 example, in Madagascar, Ramananantoandro *et al.* (2016) found that tree diameter did not  
264 significantly affect wood density for native hardwood species. Similarly, a lack of correlation

265 between tree growth and wood density was observed for *Eucalyptus globulus* in Portugal  
266 (Quilhó and Pereira 2001) and for black spruce (*Picea mariana*) in Canada (Hall 1984), while  
267 DeBell *et al.* (2001) found a negligible influence of growth rate on wood density for *Eucalyptus*  
268 *saligna* in Hawaii. In contrast, Roque and Fo (2007) and Boyle *et al.* (1988) reported negative  
269 correlations between wood density and growth traits, respectively for *Gmelina arborea* in Costa  
270 Rica and for black spruce (*Picea mariana*) in Canada. In the present study, the moderate  
271 positive correlation between *A. auriculiformis* wood density and tree diameter suggests that  
272 diameter growth can be improved with a small gain in the species wood density. This is  
273 interesting as most wood mechanical properties are closely related to wood density.

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

283 Regarding the intra-tree variation of wood density, we found no significant relation between  
284 the radial positions of the wood and the wood density for *A. auriculiformis*. Wood density being  
285 the main technological parameter of wood, its variation within trees and the magnitude of the  
286 variation provide information on the quality of the logs produced (Guilley *et al.* 2004). As such,  
287 our results indicate that the characteristics of *A. auriculiformis* logs are heterogeneous.  
288 Bouriaud *et al.* (2005) linked the variability of the radial density of wood to the differences in  
289 radial growth of trees. The latter is influenced by climatic variations, thinning and soil fertility



290 (Mäkinen and Hynynen 2012, Hietz *et al.* 2013, Miranda and Pereira 2015, Nabais *et al.* 2018).  
291 For *A. auriculiformis* these site-specific constraints could relate to climatic conditions and tree  
292 spacing.

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

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

## 307 **5- CONCLUSIONS**

308 In this study, the best model for predicting the wood density of *A. auriculiformis* as a function  
309 of diameter is a linear function, and wood density increased with tree diameter. The moderate  
310 positive correlation between *A. auriculiformis* wood density and tree diameter suggests that  
311 diameter growth can be improved with a small gain in the species wood density. The suggested  
312 opportunities for improvement include the selection of good quality parent and the practice of  
313 frequent and regular thinning to reach desirable diameters and produce high-density timber.

314 However, further experiments should evaluate the response of the species to frequent thinning  
315 in the specific climate and soil conditions of *A. auriculiformis* grown in South Benin. The study  
316 also highlights the need for reducing the intra-tree variability in wood quality, which could also  
317 be achieved through improved tree breeding.

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