

## DEVELOPMENT OF LOCAL ALLOMETRIC EQUATION TO ESTIMATE TOTAL ABOVEGROUND BIOMASS IN PAPUA TROPICAL FOREST

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DEVELOPMENT OF LOCAL ALLOMETRIC EQUATION TO ESTIMATE TOTAL ABOVEGROUND BIOMASS IN PAPUA TROPICAL FOREST. Recently, pantropical allometric equations have been commonly used across the globe to estimate the aboveground biomass of the forests, including in Indonesia. However, in relation to regional differences in diameter, height and wood density, the lack of data measured, particularly from eastern part of Indonesia, may raise the question on accuracy of pantropical allometric in such area. Hence, this paper examines the differences of local allometric equations of Papua Island with equations developed by Chave and his research groups. Measurements of biomass in this study were conducted directly based on weighing and destructive samplings. Results show that the most appropriate local equation to estimate total aboveground biomass in Papua tropical forest is  $\text{Log}(TAGB) = -0.267 + 2.23 \text{Log}(DBH) + 0.649 \text{Log}(WD)$  ( $CF=1.013$ ;  $VIF=1.6$ ;  $R^2= 95\%$ ;  $R^2\text{-adj}= 95.1\%$ ;  $RMSE= 0.149$ ;  $P<0.001$ ). This equation is also a better option in comparison to those of previously published pantropical equations with only 6.47% average deviation and 5.37 points of relative bias. This finding implies that the locally developed equation should be a better option to produce more accurate site specific total aboveground biomass estimation.

Keywords: Pantropical, local, allometric, biomass, Papua

*PENYUSUNAN PERSAMAAN ALOMETRIK LOKAL UNTUK MENDUGA BIOMASSA TOTAL DI ATAS PERMUKAAN TANAH DI KAWASAN HUTAN TROPIS PAPUA. Saat ini, persamaan alometrik pantropis telah umum digunakan untuk mendapatkan nilai dugaan biomassa di atas permukaan tanah di kawasan hutan, termasuk di Indonesia. Namun sehubungan dengan adanya perbedaan dalam karakteristik diameter, tinggi, dan berat jenis pohon, kurangnya pengukuran data, khususnya di daerah timur Indonesia menyebabkan adanya keraguan terkait besarnya simpangan dan bias yang dihasilkan oleh penggunaan persamaan pantropis di daerah tersebut. Oleh karena itu, tulisan ini mempelajari perbandingan persamaan alometrik yang dibangun secara spesifik terhadap lokasi dan persamaan alometrik pantropis yang telah dipublikasikan oleh Chave dan kelompok penelitiannya. Pendugaan biomassa pada penelitian ini dilaksanakan berdasarkan pengukuran dan penimbangan secara langsung secara destruktif. Hasil penelitian menunjukkan bahwa persamaan lokal yang sesuai untuk pendugaan nilai biomassa di atas permukaan tanah di kawasan hutan tropis Papua adalah  $\text{Log}(TAGB) = -0,267 + 2,23 \text{Log}(DBH) + 0,649 \text{Log}(WD)$  ( $CF=1,013$ ;  $VIF=1,6$ ;  $R^2= 95\%$ ;  $R^2\text{-adj}= 95,1\%$ ;  $RMSE= 0,149$ ;  $P<0,001$ ). Apabila dibandingkan dengan persamaan alometrik pantropis yang telah dipublikasikan sebelumnya, persamaan lokal tersebut menghasilkan nilai dugaan yang lebih baik dengan nilai simpangan rata-rata hanya 6,47% dan nilai bias relatif sebesar 5,37. Hasil ini mengindikasikan bahwa persamaan alometrik yang dibangun secara lokal sebaiknya dijadikan sebagai pertimbangan utama untuk mendapatkan nilai dugaan total biomassa di atas permukaan tanah yang lebih akurat.*

*Kata kunci: Pantropis, lokal, alometrik, biomassa, Papua*

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## I. INTRODUCTION

Along with the progress of reducing emission from deforestation and forest degradation (REDD+) in Indonesia and the high possibility of benefits that might be achieved from conservation of forest carbon stocks, a verifiable and precise estimation of carbon stocks in the country's forestry sector is strongly needed. Estimating forest carbon stocks relies on certain approaches depending on the scales, starting from field weighing at local level to the application of geographical information system (GIS) at national level. However, all these approaches still rely on biomass measurement of the trees. (Clark & Kellner, 2012; Jaya et al., 2012; Achmad, Jaya, Saleh, & Kuncayho, 2013). Nowadays, there is a risk of environmental deterioration as a result from direct biomass measurements, combined with the cost of such destructive approach that tends to be very high. The alternative that has been generally used is an allometric equation (Lewis et al., 2013; Ngomanda et al., 2014). In general, allometric equation is a statistical model to estimate the biomass of the trees using their biometrical characteristics, like height or diameter, which are non-destructive and simpler to measure (Eggleston et al., 2006; Maulana, 2014; Ngomanda et al., 2014). Up to now, Chave's pantropical allometric equations are the most commonly used across the globe, including in Indonesia (Lewis et al., 2009; Ngomanda et al., 2014). Nevertheless, in relation to regional differences in diameter, height and wood density, the lack of data measured from eastern part of Indonesia may question the degree of deviation and bias produced from the use of pantropical allometric which were developed by Chave, Andalo, Brown, Cairns, Chambers, Eamus, ... Yamakura (2005) and Chave, Réjou-Méchain, Búrquez, Chidumayo, Colgan, Delitti, ... Vieilledent (2014) in such area as evidenced by Maulana (2014).

Furthermore, debates over the application of Chave's pantropical allometric are continuing since several regional studies have come out with different results. A study by Fayolle et al.

(2013) reported that pantropical equation is strongly justifiable to estimate the biomass of south-eastern Cameroon forests. Similar finding had also been contended by Vieilledent et al. (2012) for its validity over biomass estimation in Madagascar. In contrast, several studies also described that the use of Chave's pantropical equations might result in significant bias, as reported by Henry et al. (2010) in Ghana, Lima et al. (2012) in Amazonia, by Alvarez et al. (2012) in Columbia.

Hence, in order to answer the dilemma between the use of pantropical and locally developed equations, this paper studies the differences of local allometric equations for Papua Island with equations developed by Chave et al. (2005) and an improved pantropical allometric equation by Chave et al. (2014). Thus, the main objective of this study is to develop an improved allometric equation for mixed species in Papua Island. Considering this objective, this study produced local allometric equations for mixed species across Papua Island as an improvement to previously published equations by Maulana (2014) using new data that includes four additional genus, namely *Anthocephalus*, *Duabanga*, *Myristica* and *Syzygium*. Afterward, the study evaluated it against both Chave et al. (2005) and Chave et al. (2014) equations using actual (direct measurements) biomass data.

## II. MATERIAL AND METHOD

### A. Study Site

As depicted in Table 1 and Figure 1, this study was conducted at six regencies across Papua Island. Table 1 also contains the number of trees felled in this study, which were 83, with dbh (diameter at breast height/1.3 m) ranging from 5 to 48.5 cm, and consist of eight genera.

### B. Biomass Measurement

Concisely illustrated in Figure 2, a set of proper and prudence procedure has been adopted to obtain reliable data and minimize any source of bias. To the extent of greater accuracy, as advised by Basuki et al. (2009), measurements of biomass in this study were

Table 1. Study area, coordinates and number of trees felled per genera

Site	Location	Genera	Coordinates	Number of trees felled
1	Sorong	<i>Anthocephalus</i>	0°33' 42" – 1° 35' 29" S 30°40' 49" – 132°3'48" E	8
2	Mamberamo	<i>Duabanga</i>	01°28 - 3°50 S 137°46 - 140° 19 E	8
3	Fak-fak	<i>Intsia</i>	2°25'0" - 4°0'0" S 131°30'0" - 138°40'0" E	13
4	Bintuni	<i>Myristica</i>	1°57'50"-3°11'26" S	9
		<i>Palaquium</i>	132°44'59"-134°14'49" E	13
		<i>Syzygium</i>		9
5	Keerom	<i>Pometia</i>	140°15'0"-141°0'0" S 2°37'0" - 4°0'0" E	15
6	Raja Ampat	<i>Vatica</i>	0°10' S - 0°20' N 130°0' W- 132°0'55" E	8
Total number of trees felled				83

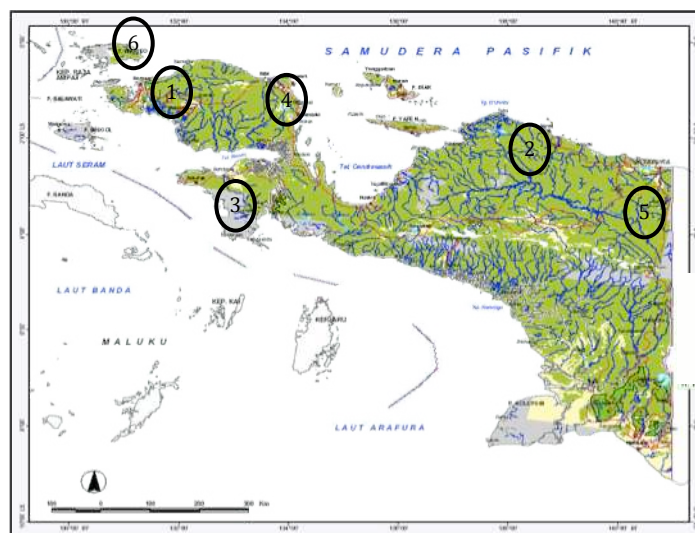


Figure 1. Study site map

Source: FWI (2004)

conducted directly based on weighing and destructive samplings. The dry biomass of the material of a pool of tree was measured using the aliquot approach, which is a piece of sample with a known mass as a fraction of the whole pool of the material. Based on this approach, dry biomass of a pool of material equals to the fresh biomass of the pool of the material divided by the fresh biomass of its aliquot times the dry biomass of the aliquot. This approach was logical if only the ratio of the dry over

fresh biomass was homogeneous for the whole pool. Therefore, as suggested by Ketterings et al. (2001), each tree felled was divided into five pools, namely leaves, twigs (diameter <3.2 cm), small branches (diameter 3.2–6.4 cm), large branches (diameter >6.4 cm) and stems.

Each tree was felled so that its crown fell on the most open ground possible in its area, which limited the destruction of its foliage to the lowest possible loss. Once a tree was felled, the volume of each section was calculated using

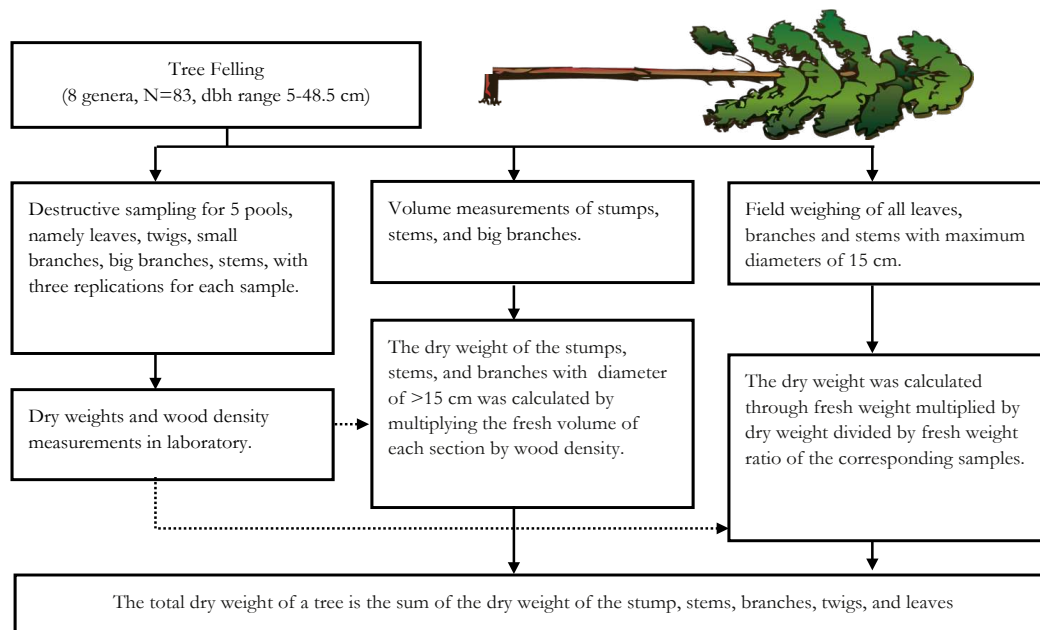


Figure 2. Flow chart of biomass measurement

Smalian’s formula as cited by De Gier (2003), so that the total volume is the sum of the volumes of each section. Meanwhile, branches and stems with maximum diameters of 15 cm were measured directly in the field using hang-up balance of 50 kg capacity with an accuracy of 1%. Moreover, the smaller samples were weighed using a 1000 gr table scale with an accuracy of 0.5%. Three replications were taken for the samples from the partitioned trees and put into sealed plastic bags, and then brought to the laboratory to measure their moisture content. From that point, an analytical balance with maximum capacity of 500 gr and an accuracy of 0.001 gr was utilized to weigh those samples. Dry weights were obtained by drying the samples at 105°C temperature until the constant value was obtained (Stewart et al., 1992; Ketterings et al., 2001).

In order to measure the wood density at the laboratory, samples were taken from the lower and upper parts of the main trunk sections with 2 meters interval. To include the inner and outer parts of the trunks with their bark, the samples were taken as a pie shape or cylinder (Nelson et al., 1999). Water replacement method was used in measuring the wood density. The

samples were saturated at first to prevent size contraction during volume measurement. This was conducted through 48 hours rehydration. Each sample’s volume was obtained from the displaced water volume when submerged. Finally, the wood density was equal to the oven dry weight divided by the saturated volume. The dry weight of the stumps, stems, and branches with diameter of >15 cm was calculated by multiplying the fresh volume of each section by wood density. For the other partitioned trees, the dry weight was calculated through fresh weight multiplied by dry weight divided by fresh weight ratio of the corresponding samples. The total dry weight of a tree is the sum of the dry weight of the stump, stem, branches, twigs, and leaves (Stewart et al., 1992).

### C. Allometric Equations

#### 1. Analysis of variance

Based on findings by Maulana (2014), in this study locally developed allometric for mixed species was established using two predictors, namely diameter at breast height (*DBH*) and wood density (*WD*). Hence, the equations to estimate total aboveground biomass (*TAGB*) were established according to the following

basic models:

$$\text{Log}(TAGB) = c + \alpha \text{Log}(DBH) + \beta \text{Log}(WD) \quad (1)$$

$$TAGB = c + \alpha DBH + \beta WD^2 \quad (2)$$

$$TAGB = c + \alpha WD + \beta DBH^2 \quad (3)$$

Subsequently, in order to fulfill the assumptions in the regression establishment, four tests were conducted, namely Variance Influential Factors (*VIF*) for multicollinearity test, Pearson's correlation coefficient test, normal distribution of residuals test, and test for constant variance of residuals. Allometric model comparison and selection was analyzed using the values of standard error of the coefficient,  $R^2$ ,  $R^2$  (adj), and root mean square error (*RMSE*) based on Minitab 14.0 software. The chosen model would be the one with the highest values for  $R^2$  and  $R^2$  (adj), while having the lowest values of standard error of the coefficient and *RMSE*. Additionally, in order to enhance the reliability of the established log model (Equation 1), a correction factor (*CF*; Equation 4) for back transforming the model was calculated from the standard error of estimate (*SEE*; Equation 5) as defined in Sprugel (1983).

$$CF = \exp(SEE^2/2) \quad (4)$$

$$SEE = \sqrt{\sum \left( \frac{\text{Log} Y_i - \text{Log} \hat{Y}_i}{N - 2} \right)^2} \quad (5)$$

Where:

$\text{Log} Y_i$  = values of dependent variable

$\text{Log} \hat{Y}_i$  = corresponding predicted values calculated from the equation

$N$  = total number of observations

Afterwards, using data from actual biomass measurement in Papua Island, the chosen equation in this study was evaluated against Chave et al. (2005) equation, which were Equation 6 and Equation 7, as well as Chave et al. (2014)'s pantropical allometric as depicted in Equation 8. Meanwhile, as suggested by Basuki et al. (2009); Ngomanda et al. (2014) and Tedeschi (2006), criteria for this evaluation included average deviation (Eq. 9) and relative

bias (Eq. 10).

$$TAGB = WD \exp [-1.499 + 2.148 \text{Ln}(DBH) + 0.207 (\text{Ln}(DBH))^2 - 0.0281 (\text{Ln}(DBH))^3] \quad (6)$$

$$TAGB = WD \exp [-1.239 + 1.98 \text{Ln}(DBH) + 0.207 (\text{Ln}(DBH))^2 - 0.0281 (\text{Ln}(DBH))^3] \quad (7)$$

$$TAGB = 0.0673 \times (DBH^2 \times WD \times H)^{0.976} \quad (8)$$

$$\bar{S} = \frac{100}{\eta} \sum_{i=1}^{\eta} \frac{|B_i - b_i|}{b_i} \quad (9)$$

$$RB = \frac{1}{\eta} \sum_{i=1}^{\eta} (B_i - b_i) / b_i \quad (10)$$

Where:

$\bar{S}$  = average deviation

$RB$  = relative bias

$B_i$  = actual aboveground biomass for tree-*i*

$b_i$  = its estimation based on the model

$\eta$  = number of observations

### III. RESULT AND DISCUSSION

#### A. Local Equation

Compared to wood density values from published literature, as shown in Table 2, result of measurements in this study illustrate a highly rational wood density for each genera. However, it should be kept in mind that although samples for wood density measurements originated from both upper and lower parts (with 2m interval) of each main trunk, these data were also used to estimate the weight of the material of the trees that were difficult or even impossible to measure, such as big branches. As reported by Basuki et al. (2009), this approach might result in over-estimation in regard to total weight of the tree. Meanwhile, Nogueira et al. (2007) also noted that wood density of a tree tends to be higher at breast height than in the upper part of the bole, and also higher at the bottom of the trunk of the tree than at the living crown's base. Afterwards, following the previously determined model, three local equations were established as listed in Table 3. In this study, multicollinearity test was conducted by harnessing the value of Variance Influential

Table 2. Result of wood density measurements

Genus	Number of trees (N)	Number of wood density sample (n)	Wood density range (gr/cm <sup>3</sup> )		Average (gr/cm <sup>3</sup> )	Standard deviation	Coefficient of variation	PROSEA (2007)* (gr/cm <sup>3</sup> )
			min	max				
<i>Anthocephalus</i>	8	36	0.30	0.56	0.43	0.09	0.21	0.29-0.56
<i>Duabanga</i>	8	42	0.28	0.48	0.38	0.07	0.18	0.27-0.51
<i>Intsia</i>	13	92	0.43	0.86	0.64	0.15	0.23	0.50-1.04
<i>Myristica</i>	9	87	0.41	0.63	0.52	0.08	0.15	0.40-0.65
<i>Palaquium</i>	13	86	0.33	0.56	0.44	0.07	0.16	0.45-0.51
<i>Pometia</i>	15	98	0.37	0.75	0.56	0.12	0.21	0.39-0.77
<i>Syzygium</i>	9	74	0.54	0.80	0.67	0.09	0.13	0.56-0.83
<i>Vatica</i>	8	44	0.54	0.67	0.60	0.05	0.08	0.60-0.76
Total	83	559	-	-	-	-	-	-

\*published wood density

Factors (*VIF*). In short, according to Fahrmeir et al. (2013), the aim of such test is to investigate whether there is an indication of high correlation among predictors (input variables, x) or not. Evidently, none of the established models indicated the presence of significant multicollinearity among its predictors, since the *VIF* value only ranged from 1.4 to 1.6 point. As described by Fahrmeir et al. (2013), this means that there is only a very low degree of correlation among predictors used in the model, and that values of *VIF* were insufficient to be overly concerned.

Meanwhile, to strengthen the result of the multicollinearity test, analysis of correlation among variables was performed. The first step of this correlation test was to use a

graphical approach or scatter plot to explore the appearance of correlation among involved variables (Chaturvedi & Raghubanshi, 2015). As shown in Figure 3, and in line with the result of multicollinearity test, apparently it is clear that there was no obvious relationship among predictors or input variables, *WD* and *DBH*. In contrast, both scatter plots for *TAGB* vs *WD* and *TAGB* vs *DBH*, which were basically output vs input variables, demonstrated a direct relationship. The strength of this correlation was further examined using Pearson's coefficient test and the results were depicted in Table 4. Considering the value of Pearson's correlation of coefficient (*r*) in the table, where  $r > 0.6$ , it seems that *DBH* and *WD* were reliable input variables in predicting *TAGB*. This finding is

Table 3. Result of wood density measurements

Equations	Coefficient		Multicollinearity Test	
	Symbol	Value	Predictor	<i>VIF</i>
$\text{Log}(TAGB) = c + \alpha \text{Log}(DBH) + \beta \text{Log}(WD)$	c	-0.267	Log( <i>DBH</i> )	1.6
	α	2.23	Log( <i>WD</i> )	1.6
	β	0.649	-	-
$TAGB = c + \alpha DBH + \beta WD^2$	c	-557	<i>DBH</i>	1.4
	α	42.4	<i>WD</i> <sup>2</sup>	1.4
	β	540	-	-
$TAGB = c + \alpha WD + \beta DBH^2$	c	-387	<i>WD</i>	1.4
	α	710	<i>DBH</i> <sup>2</sup>	1.4
	β	0.891	-	-

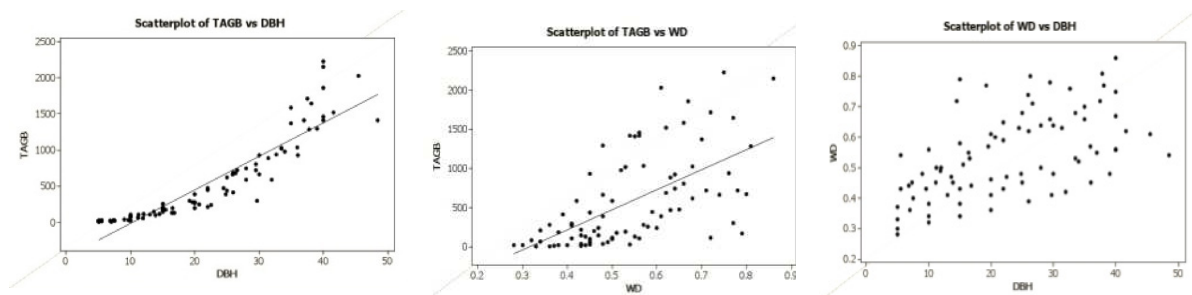


Figure 3. Scatter plots for output-input ( $TAGB$  vs  $DBH$ ;  $TAGB$  vs  $WD$ ) and input-input ( $WD$  vs  $DBH$ ) correlations

Table 4. Pearson's correlation of coefficient for output-input variables

Equations	Output (y)	Input (x)	$r^*$	P value
$\text{Log}(TAGB) = c + \alpha \text{Log}(DBH) + \beta \text{Log}(WD)$	$\text{Log}(TAGB)$	$\text{Log}(DBH)$	0.97	<0.001
	$\text{Log}(TAGB)$	$\text{Log}(WD)$	0.68	<0.001
$TAGB = c + \alpha DBH + \beta WD^2$	$TAGB$	$DBH$	0.91	<0.001
	$TAGB$	$WD^2$	0.60	<0.001
$TAGB = c + \alpha WD + \beta DBH^2$	$TAGB$	$WD$	0.61	<0.001
	$TAGB$	$DBH^2$	0.94	<0.001

\*Pearson's correlation of coefficient

in agreement with studies by Chaturvedi and Raghubanshi (2015) as well as Hunter (2015) that describe the important use of  $DBH$  and  $WD$  as input variables. Moreover, based on the value of  $r$  in the table, it appears that  $TAGB$  has more direct relationship with  $DBH$  ( $r > 0.9$ ) than  $WD$  ( $0.6 < r < 0.7$ ).

Furthermore, in Table 3 normal distribution of residual test was conducted for each established local equation. The result of this test is shown in Figure 4. It illustrates that the residual points for each equation fall near to a straight line in the normal probability plot. As explained by Fahrmeir et al. (2013), this indicates that errors during observation have been normally distributed in every  $x$ -value and expresses the validity of the normality of the residual assumption.

The final phase of the evaluation for regression assumptions conducted in this study was testing the constant variance of residuals. This evaluation was crucial to make sure that error terms or 'residuals' were constant, and

had a mean close to zero. In this study, this test was conducted based on residuals versus fitted values. The result of this test is depicted in Figure 5. Based on this graphical illustration, it can clearly be seen that only the log-based model ( $\text{Log}TAGB = -0.267 + 2.23 \text{Log}DBH + 0.649 \text{Log}WD$ ) produces valid result to fulfill the assumption of constant variance of residuals. Points on the plot for this log-based model appear to be randomly scattered all over the zero line. Thus, it is highly reasonable to assume that the residuals may have a nearly zero mean and it is virtually constant (Gardner & Urban, 2003; Fahrmeir et al., 2013).

In contrast, there were noticeable U-shaped patterns for two other non log-based equations, namely  $TAGB = -557 + 42.4DBH + 540WD^2$ , and  $TAGB = -387 + 710WD + 0.981DBH^2$ . Points of residuals for these two equations were scattered on the positive sides with large

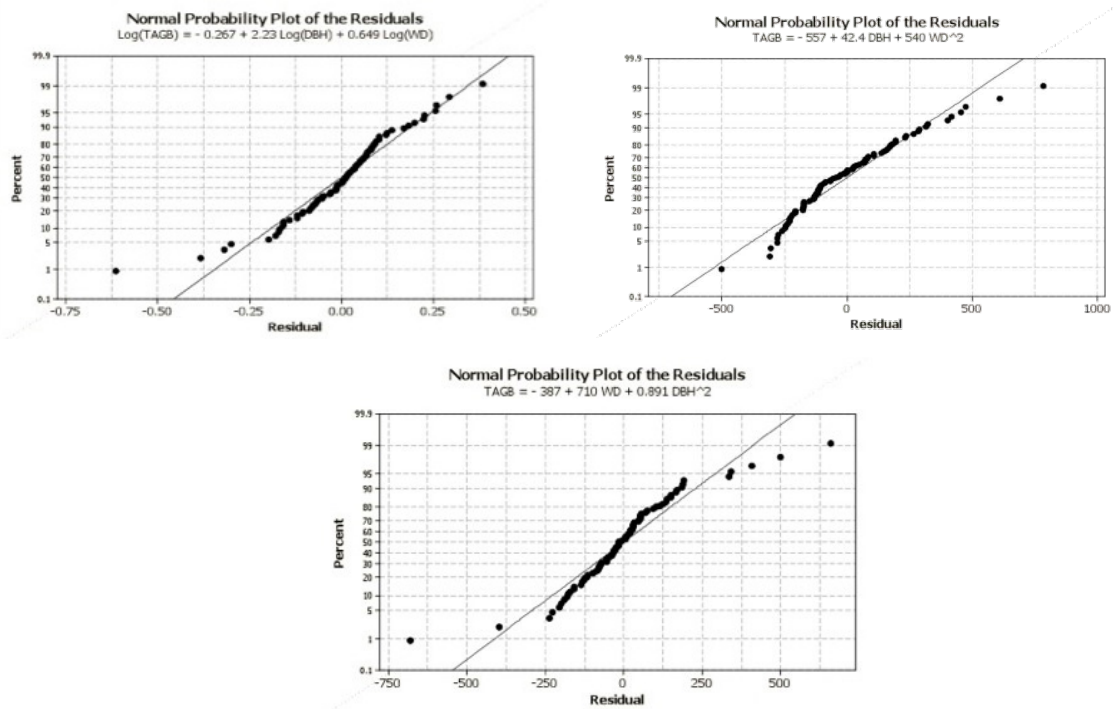


Figure 4. Normal probability plots of the residuals for each equation

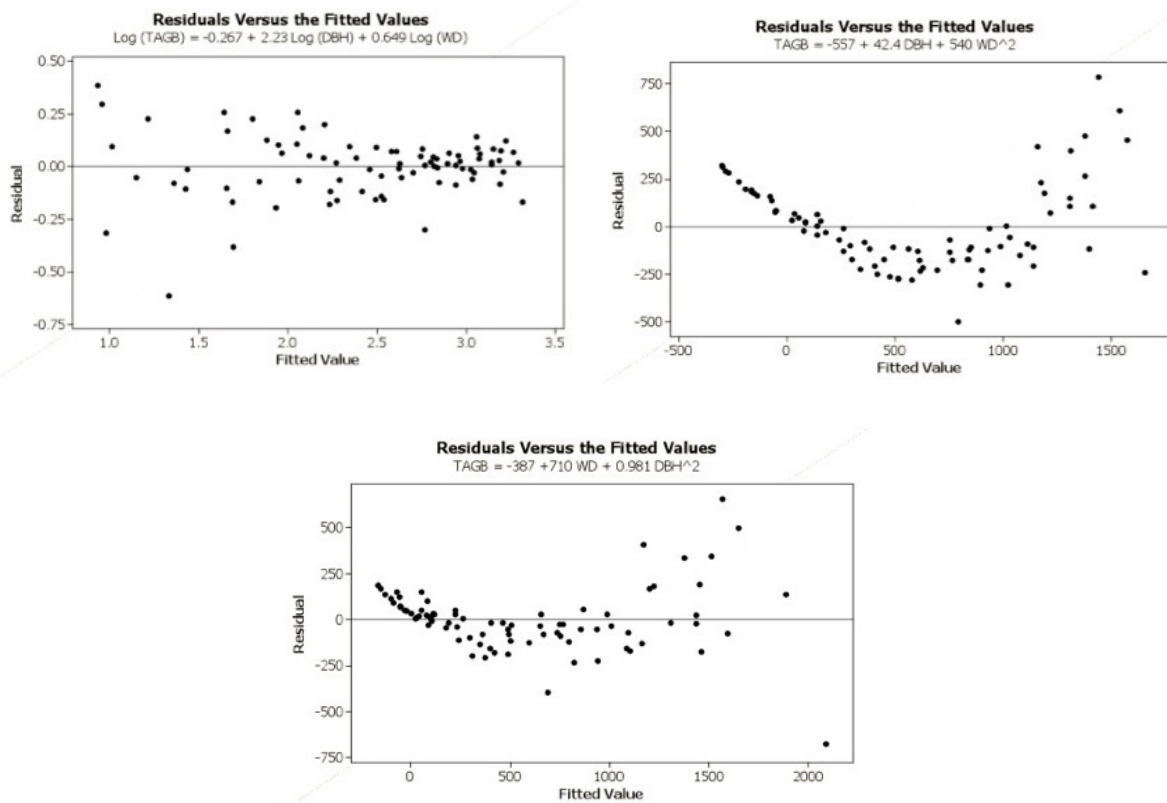


Figure 5. Results of constant variance of residuals test on three developed equations



Table 5. Evaluation of established local equation

Equations	Coefficient		Standard error of the coefficient	T-stat	R <sup>2</sup> (%)	R <sup>2</sup> -adj (%)	RMSE	P value
	Symbol	Value						
$\text{Log}(TAGB) = c + \alpha \text{Log}(DBH) + \beta \text{Log}(WD)$	c	-0.267	0.14	-1.89	95.1	95	0.149	<0.001
	$\alpha$	2.23	0.078	28.35				
	$\beta$	0.649	0.186	3.49				
$TAGB = c + \alpha WD + \beta DBH^2$	c	-387	81.53	-4.75	91	90.8	178.64	<0.001
	$\alpha$	710	165.1	4.3				
	$\beta$	0.891	0.04	21.94				
$TAGB = c + \alpha DBH + \beta WD^2$	c	-557	62.44	-8.91	85	84.6	230.31	<0.001
	$\alpha$	42.4	2.609	16.25				
	$\beta$	540	191.8	2.82				

or small fitted values, while holding negative values in the middle. This pattern implies that residuals were less likely to be consistently scattered around the zero line from left to right. Hence, the consistency of the variance of the residuals for both of the non log-based models might become questionable.

Having evaluated the assumptions of regression for each established equation, it seems that the most appropriate model to estimate total aboveground biomass in Papua tropical forest is  $\text{Log}(TAGB) = -0.267 + 2.23 \text{Log}(DBH) + 0.649 \text{Log}(WD)$  since it produces complete valid results for all given assumption tests. Besides, as shown in Table 5, this log-based equation has the highest coefficient of determination ( $R^2$ ) with about 95%, meaning that it enables to explain up to 95% of the variability of data response around its mean. This model also has the lowest value of standard error of the coefficient and *RMSE*. Additionally, this selected log-based model was also completed with its corresponding correction factor for back transforming the model. Correction factor (*CF*) for the selected log-based model in this study is 1.013. To the extent of greater accuracy, final estimation result obtained from this model should be multiplied with the correction factor.

## B. Pantropical vs Local Equation

Following the selection of locally developed equation, the selected equation  $\text{Log}(TAGB) = -0.267 + 2.23 \text{Log}(DBH) + 0.649 \text{Log}(WD)$  was then compared to two pantropical equations by Chave et al., (2005) namely  $TAGB = WD \exp[c + \alpha \text{Ln}(DBH) + \beta (\text{Ln}(DBH))^2 + d (\text{Ln}(DBH))^3]$  for moist type forest and  $TAGB = WD \exp[c + \alpha \text{Ln}(DBH) + \beta (\text{Ln}(DBH))^2 + d (\text{Ln}(DBH))^3]$  for wet type forest, and an improved model by Chave et al., (2014), which is  $TAGB = 0.0673 \times DBH^2 \times WD \times H^{0.976}$ . Compared to the actual biomass data that was calculated directly at the research site based on destructive sampling and field weighing of mixed genus trees with the *DBH* range as covered in the model of this study, it seems that the local model has the lowest value of average deviation and relative bias, with only 6.47% and 5.37 points respectively. On the other hand, both Chave et al. (2005) and Chave et al. (2014) equations have more than 15% average deviation and 10 points of relative bias. Furthermore, in order to make it clear, Figure 6 shows a graphical illustration on the comparison of the results of the estimations from each equation with the actual biomass data.

Table 6. Evaluation against pantropical equations harnessing actual measurements data

Reference	Equations	Coefficient		Average deviation	Relative bias (+/-)
		Symbol	Value		
Chosen equation of this study	$\text{Log}(TAGB) = c + \alpha \text{Log}(DBH) + \beta \text{Log}(WD)$	c	-0.267	6.47%	5.37
		$\alpha$	2.23		
		$\beta$	0.649		
Chave et al. (2005); pantropical allometric for moist type forest	$TAGB = WD \exp[c + \alpha \text{Ln}(DBH) + \beta (\text{Ln}(DBH))^2 + d (\text{Ln}(DBH))^3]$	c	-1.499	16.22%	-13.46
		$\alpha$	2.148		
		$\beta$	0.207		
		d	-0.0281		
Chave et al. (2005); pantropical allometric for wet type forest	$TAGB = WD \exp[c + \alpha \text{Ln}(DBH) + \beta (\text{Ln}(DBH))^2 + d (\text{Ln}(DBH))^3]$	c	-1.239	34.63%	-28.74
		$\alpha$	1.98		
		$\beta$	0.207		
		d	-0.0281		
Chave et al. (2014); improved pantropical allometric	$TAGB = c \times (DBH^2 \times WD \times H)^x$	c	0.0673	15.27%	-12.67
		$\alpha$	0.976		

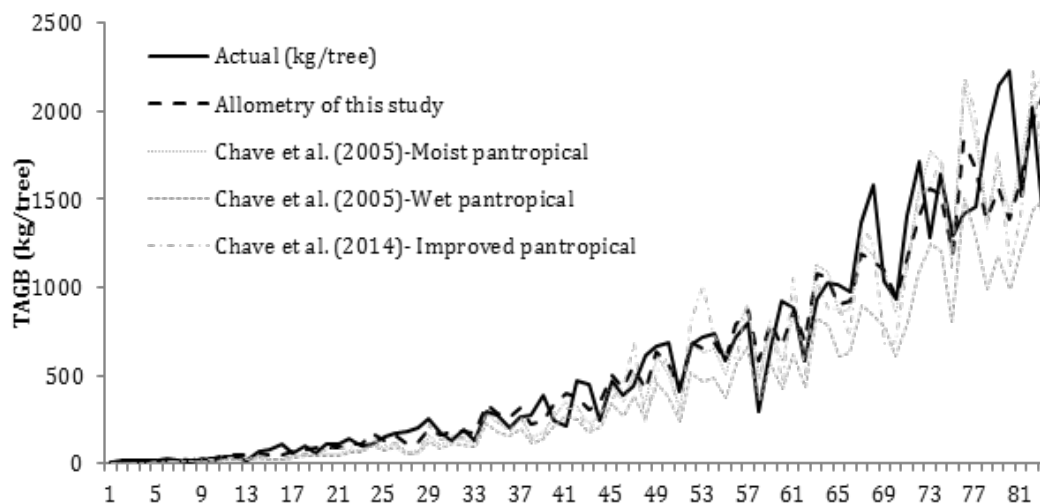


Figure 6. Comparison between actual biomass data and estimate values for each equation

**IV. CONCLUSION**

This study highlighted that the most appropriate local model to estimate total aboveground biomass in Papua tropical forest is  $\text{Log}(TAGB) = -0.267 + 2.23 \text{Log}(DBH) + 0.649 \text{Log}(WD)$  ( $CF=1.013$ ;  $VIF=1.6$ ;  $R^2=95\%$ ;  $R^2\text{-adj}= 95.1\%$ ;  $RMSE=0.149$ ;  $P<0.001$ ). In addition, this model is also a better option

compared to Chave et al. (2005) and Chave et al. (2014)'s improved pantropical equations with only 6.47% average deviation and 5.37 points of relative bias in estimating  $TAGB$  in Papua Island. This finding implies that the locally developed equation should be considered as a better option to produce more accurate site-specific total aboveground biomass estimation.

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