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Article

Carbon Yield Model for a Makino Bamboo (*Phyllostachys makinoi* Hayata) Plantation by Various Thinning Intensities

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Abstract: This study aimed to develop a carbon yield model for a Makino bamboo plantation and addressed a monoculture Makino bamboo plantation in central Taiwan. We established a long-term trial to monitor the stand dynamic following thinning. A total of 16 plots were installed based on four thinning treatments, each sharing four plots. After thinning, data were collected from two investigations, resulting in 32 records. We calculated the number of culms (N), mean diameter at breast height (MDBH), and basal area (BA) for each record. An allometric function developed by previous research was used to predict aboveground biomass and to obtain aboveground carbon storage (AGCS). The model used the N, MDBH, and BA with various types (reciprocal and natural logarithm types) as independent variables to predict $\ln(\text{AGCS})$ by the stepwise method. According to the Rad^2 value, the best predictive models were: $\ln(\text{AGCS}) = -0.004 + 0.820 \ln(\text{MDBH}) + 1.008 \ln(\text{BA})$, and $\ln(\text{AGCS}) = 7.550 - 0.820 \ln(N) + 1.828 \ln(\text{BA})$. From these two models, we found the factors, either BA and MDBH or BA and N, can effectively predict AGCS. It indicated that stand density was critical in affecting AGCS, especially the variable BA.

Keywords: bamboo plantation; thinning; aboveground biomass; aboveground carbon storage; allometric model

1. Introduction

Bamboo resources possess many advantages, including rapid growth, high productivity, a short yield period, and bamboo plants with multiple utilization [1–18]. Therefore, bamboo resources are essential for people, especially in Asia. Because the weather and environmental factors are suitable for bamboo growth, over 150,000 ha of bamboo forests cover terrestrial areas, and most such forests are plantations, providing multiple ecological services in Taiwan [7,8,18–25]. Traditionally, culms and bamboo shoots are two main products that provide jobs and increase income for local people in villages [7,8,25–28,30,31].

However, selective cutting or thinning is necessary to improve productivity in managed bamboo plantations, regardless of harvesting culms or bamboo shoots [21,30,32–35]. Such an approach only removes old culms, and most culms are still maintained in the lands of bamboo plantations after harvesting, indicating the lands are always covered by most bamboo plants in bamboo forests. Because the process of harvesting bamboo forests is friendly to the environment, the products of bamboo culms are also regarded as non-timber forest products (NTFPs) [15,25]. In recent years, bamboo forests with high potential for carbon storage have been discovered worldwide for various species because their rapid growth results in fast accumulation of dry mass [9,21,23,25,28,29,33,36–40].

Makino bamboo (*Phyllostachys makinoi*) is a native species with a particular ecological meaning in Taiwan. Plantations of this species are widely distributed in northern and central Taiwan [8,21,24,41]. Meanwhile, this bamboo is a crucial bamboo species because both culms and bamboo shoots have high economic value. Usually, they are planted by monoculture for financial purposes [8,19–21,24,41,42]. Since Makino bamboo possesses high ecological and economic values, numerous

studies have addressed this bamboo species in various aspects. Those studies included cost analysis for managed plantations [43], assessing growth and biomass accumulation for the stand level [19], analyzing stand structure of plantations [21,42], the impact of thinning on the growth and biomass accumulation [20], quantifying culm height growth by growth functions [41], and assessing the ability of carbon storage for the stand level [9,21,25].

However, rare studies addressed the carbon yield model and explored the factors affecting carbon yield for the Makino bamboo plantations. Developing a carbon yield model helps assess the contribution of carbon storage for this bamboo species. Therefore, the objectives of this study were to (1) collect data from stands of various stand densities resulting from thinning treatments, (2) predict aboveground carbon storage (AGCS) based on aboveground biomass (AGB), (3) develop a carbon yield model, and (4) analyze the main factors to affect AGCS, for Makino bamboo.

2. Materials and Methods

2.1. Study Areas

The study site was located in central Taiwan, belonging to the Tai Keng region, Beitun District, Taichung City (between 120.74° and 120.76°E and 24.16° and 24°18'N), with the lower mountainous area (at an elevation of 226 m). The detailed location of the study site is shown in Figure 1.



Figure 1. Location of this study.

Because this region is abundant in natural resources and has beautiful scenery, it was designated as a scenic area by the Taichung City Government in 1976, called the “Dakeng Scenic Area” [44]. According to the weather data of Taichung City from 1981 to 2010, the monthly temperature was from 17.0°C (in January) to 28.9°C (in July), the mean was 23.7°C, the relative humidity was 74.5%, and the annual rainfall was 1,762.8 mm [45]. This region is also rich in bamboo resources, and Makino bamboo is a critical bamboo species. Most of the Makino bamboo are plantations managed by farmers for economic benefits. In recent decades, bamboo shoot production had higher monetary values than the culm harvest; therefore, harvesting bamboo shoots is a significant management type for Makino bamboo plantations [24]. To develop a growth and yield system for this bamboo species, a site of a

monoculture plantation by private farmer management was selected as an experimental site (Figure 1). The farmer allowed us to practice thinning and monitoring in its plantation.

2.2. Materials

This study established a long-term trial to monitor the stand dynamics of a Makino bamboo plantation with various thinning treatments in 2019. Four treatments, namely Treatments I, II, III, and IV, by different thinning intensities, were designed and installed on this site. The thinning intensity used in this study is listed in Table 1.

Table 1. The various thinning treatments used in this study.

Treatment	Thinning intensity	Performance
I	Heavy thinning	Thinning 75% of the culm number and 25% of the culm number were reserved.
II	Moderate thinning	Thinning 50% of the culm number and 50% of the culm number were reserved.
III	Light thinning	Thinning 25% of the culm number and 75% of the culm number were reserved.
V	No thinning	Without thinning, 100% of the culm number was reserved.

Treatments I, II, III, and IV were thinning 75%, 50%, 25%, and 0% of the culm number (hereafter also called heavy, moderate, light, and no thinning), respectively. The principle of the thinning was thinning from the older culms and abnormal culms. In March 2019, after the thinning project was determined, 16 sample plots of 5 × 5 m were randomly installed. Thinning treatments were performed in May 2019. Each treatment shared four plots. At the same time, we measured DBH and determined the culm age for each culm within plots after thinning treatments. The culm age was determined based on the colour and status of the culms. Please refer to Yen et al. [21] and Liu et al. [24] for detailed approaches. In February 2021, we resurveyed the same items of culms for each plot and recorded new culms developing in this period. However, we did not harvest bamboo shoots for all plots to monitor the dynamic of bamboo stands after thinning.

2.3. Data Analysis

We calculated the AGCS for the plots of various thinning treatments with two investigations. The detailed processes were described as follows:

1. The allometric model predicted AGB for individual bamboo plants based on their DBH. Because the allometric function for predicting AGB has been built for Makino bamboo in this region by Yen et al. [21], the present study directly cited this model for estimating the AGB of each bamboo plant. The model is shown as Equation (1) [21].

$$AGB = 0.156 \times DBH^{2.118} \quad (1)$$

where AGB is aboveground biomass, and DBH is the diameter at breast height.

2. The AGB of plots was obtained from the summation of each individual within plots when the AGB of individuals was predicted. Consequently, the AGB of all plots was obtained, and we formatted the unit of AGB as Mg ha⁻¹.
3. The AGCS prediction was based on the AGB and percentage of carbon content (PCC), indicating that AGCS equals AGB × PCC. However, aboveground consists of foliage, branches, and culms for bamboo plants. The study required the proportion of each section's biomass to AGB and its PCC and obtained the PCC of AGB from the summation of each section's biomass × its PCC [30]. In a previous study, Yen et al. [21] determined the proportion biomass of foliage, branches and culms to be 8.4, 15.7 and 75.9%, respectively, and their PCC was 40.08, 46.06 and 47.65 %, respectively. Consequently, the aboveground PCC was calculated as: (8.4% × 40.08%) + (15.7% × 46.06%) + (75.9% × 47.65%) = 46.76%. In this study, we used AGB × 46.76% for predicting AGCS.

4. We calculated the AGCS (Mg ha^{-1}), culm number (culm ha^{-1}), mean DBH (cm), and BA ($\text{m}^2 \text{ha}^{-1}$) for each plot with two periods (2019 and 2021). The regression model employed the culm number, mean DBH, and BA as independent variables and $\ln(\text{AGCS})$ as a dependent variable. The model is shown as Equation (2)

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \varepsilon \quad (2)$$

where Y is aboveground carbon storage (AGCS) taken the natural log as $\ln(\text{AGCS})$; X_1 is number of culms per hectare (N), $1/N$, or $\ln(N)$; X_2 is mean diameter of breast area (MDBH), $1/\text{MDBH}$, or $\ln(\text{MDBH})$; and X_3 is basal area per hectare (BA), $1/\text{BA}$, or $\ln(\text{BA})$; and ε is the error.

After modelling, the R_{adj}^2 value was used as an indicator to evaluate the models' performances, where a higher value indicated a model with better predictive ability. The model with the highest R_{adj}^2 value was chosen to analyze bias. The root mean square error (RMSE) was used to evaluate the fitness of the models for observed and predicted AGCS. The detailed formula for RMSE is given in Equation (3)

$$RMSE = \sqrt{\frac{\sum_{i=1}^n (Y_i - \hat{Y}_i)^2}{n}} \quad (3)$$

where Y_i and \hat{Y}_i are the i observation and the predicted aboveground carbon storage by the model, respectively, and n is the total number of observations.

3. Results

3.1. Stand Characteristics at Two Investigations

The N, MDBH, and BA were calculated based on individuals. According to Equation (1) and the PCC, the AGB and AGCS were obtained. Table 2 shows the above stand characteristics with various thinning treatments at two investigations.

Table 2. The culms number (N), mean diameter at breast height(MDBH), basal area (BA), aboveground biomass (AGB), and aboveground carbon storage (AGCS) with various thinning treatments in 2019 and 2021.

Investigation time	Treatment	N (culms ha^{-1})	MDBH (cm)	BA ($\text{m}^2 \text{ha}^{-1}$)	AGB (Mg ha^{-1})	AGCS (Mg ha^{-1})
2019	I	11,500 ± 1,438	2.71 ± 0.15	3.11 ± 0.34	15.32±1.93	7.16±0.90
	II	22,900 ± 1,149	2.50 ± 0.09	5.72 ± 0.48	26.26±0.08	12.28±1.44
	III	31,500 ± 2,600	2.23 ± 0.18	7.02 ± 0.95	29.93±0.28	13.99±2.47
	IV	49,200 ± 19,713	1.95 ± 0.16	9.42 ± 3.04	35.30±10.14	16.51±4.74
2021	I	29,000 ± 8,116	2.39 ± 0.14	6.88 ± 1.72	29.51±7.32	13.80±3.42
	II	41,300 ± 4,530	2.40 ± 0.08	9.88 ± 0.92	43.17±3.98	20.19±1.86
	III	53,500 ± 13,808	2.25 ± 0.19	12.08 ± 3.51	52.54±17.17	24.57±8.03
	IV	80,900 ± 35,743	2.00 ± 0.14	15.96 ± 6.27	60.68±24.43	28.37±11.42

Table 2 provides fundamental information for stand characteristics with various thinning treatments and their development. In 2019, after thinning, N, BA, AGB, and AGCS decreased, and MDBH increased, with thinning intensity increasing. The stand characteristic also displayed the same pattern in 2021. During stand development, we found that N, BA, AGB, and AGCS rose from 2019 to 2021, regardless of thinning intensity. Reasonably, their increase resulted from stand development. However, our study's purpose was to develop a carbon yield model and did not further analyze the relationship between thinning treatment and stand development in the present study.

3.2. Carbon Yield Models Based on Stand Characteristics

The carbon yield model used N, MDBH, and BA (including their reciprocal and natural logarithm types) as independent variables to predict $\ln(\text{AGCS})$ for Makino bamboo. The multiple regression models employed stepwise approaches to solve parameters. Table 3 shows 27 regression models.

Table 3. The regression coefficients and R_{adj}^2 of regression models predicting aboveground carbon storage by stand characteristics.

Regression model $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3$							
Stand characteristics ¹			Regression coefficients				R_{adj}^2
X_1	X_2	X_3	β_0	β_1	β_2	β_3	
N	MDBH	BA	2.954	-3.341×10^{-5}	-0.477	0.253	0.9468
		1/BA	1.608	1.013×10^{-5}	0.586	-4.257	0.9830
		ln(BA)	-0.173	– ²	0.365	1.011	0.9956
1/ MDBH	BA	BA	0.575	-3.884×10^{-5}	2.903	0.278	0.9476
		1/BA	4.191	1.108×10^{-5}	-2.992	-3.948	0.9854
		ln(BA)	1.473	–	-1.790	1.003	0.9955
ln (MDBH)	BA	BA	2.834	-3.596×10^{-5}	-1.184	0.265	0.9471
		1/BA	1.806	1.063×10^{-5}	1.340	-4.101	0.9845
		ln(BA)	-0.004	–	0.820	1.008	0.9958
1/N	MDBH	BA	1.397	-1.226×10^4	0.523	0.065	0.9879
		1/BA	4.631	5.578×10^4	-0.378	-20.361	0.9367
		ln(BA)	-0.173	–	0.365	1.011	0.9956
1/ MDBH	BA	BA	3.589	-1.106×10^4	-2.401	0.068	0.9859
		1/BA	3.147	4.975×10^4	1.396	-18.917	0.9342
		ln(BA)	1.473	–	-1.790	1.003	0.9955
ln (MDBH)	BA	BA	1.629	-1.163×10^4	1.132	0.066	0.9872
		1/BA	4.363	5.239×10^4	-0.727	-19.551	0.9354
		ln(BA)	-0.004	–	0.820	1.008	0.9958
ln (N)	MDBH	BA	-9.716	1.014	0.815	–	0.9946
		1/BA	-9.716	1.014	0.815	–	0.9946
		ln(BA)	7.550	-0.820	–	1.828	0.9958
1/ MDBH	BA	BA	-5.292	0.922	-3.788	0.009	0.9955
		1/BA	-6.734	1.074	-4.109	0.533	0.9951
		ln(BA)	7.550	-0.820	–	1.828	0.9958
ln (MDBH)	BA	BA	-9.287	1.008	1.828	–	0.9958
		1/BA	-9.287	1.008	1.828	–	0.9958
		ln(BA)	7.550	-0.820	–	1.828	0.9958

¹⁾ N is culm number per ha; and MDBH is mean diameter at breast height; BA is basal area per ha. ²⁾ “–” indicates that this variable does not enter the model.

From Table 3, high R_{adj}^2 values were shown for all models, indicating they had high predictive ability for carbon yield. We also found that only two variables entered some regression models but still had high R_{adj}^2 . Among 27 models, the two models simultaneously had the highest value of R_{adj}^2 (0.9958), and they only had two independent variables, that is, $\ln(ABCS) = -0.004 + 0.820 \ln(DBH) + 1.008 \ln(BA)$, and $\ln(ABCS) = 7.550 - 0.820 \ln(N) + 1.828 \ln(BA)$. Therefore, we chose these two models to predict the carbon yield of Makino bamboo. Notably, these two models commonly showed $\ln(BA)$, and this variable was first entered into the models by the stepwise approach, regardless of which model. It indicated that $\ln(BA)$ was critical in predicting $\ln(ABCS)$. Only used this variable already had high predictive ability. The simple regression for the model was $\ln(ABCS) = 0.933 + 0.882 \ln(BA)$, $R_{adj}^2=0.9693$. The R_{adj}^2 of this model was also higher than some models in Table 3. We further assessed the predictive ability for this model and the two models with the highest R_{adj}^2 (hereafter called Models

I, III, and III for $\ln(ABCS) = 0.933 + 0.882 \ln(BA)$, $\ln(ABCS) = -0.004 + 0.820 \ln(MDBH) + 1.008 \ln(BA)$, and $\ln(ABCS) = 7.550 - 0.820 \ln(N) + 1.828 \ln(BA)$, respectively.

3.3. The Scatter Diagram and RMSE for Each Model

To show the relationship between observations and predicted AGCS, a scatter diagram was used to display the predictive effects of each model, as shown in Figure 2.

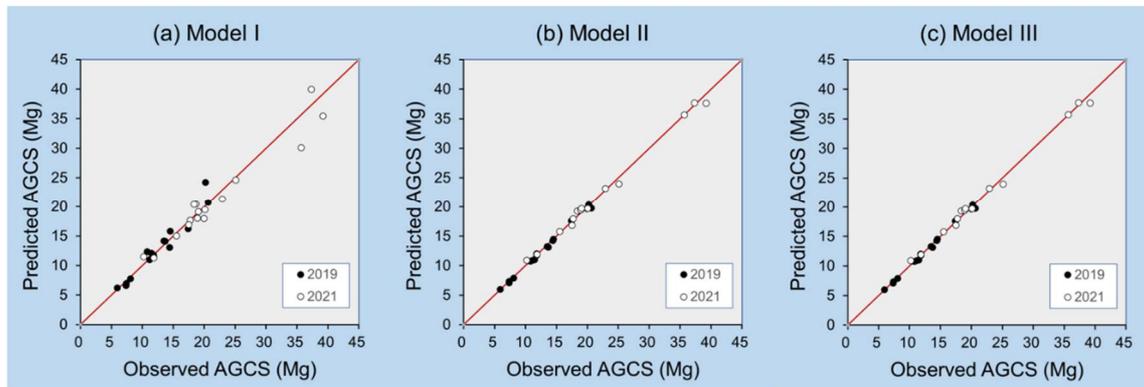


Figure 2. Relationships between observations and predicted aboveground carbon storage (AGCS) with a perceivable diagonal line (1:1 for X and Y) for each model by the scatter diagram for three models, where Model I is $\ln(ABCS) = 0.933 + 0.882 \ln(BA)$, Model II is $\ln(ABCS) = -0.004 + 0.820 \ln(MDBH) + 1.008 \ln(BA)$, Model III is $\ln(ABCS) = 7.550 - 0.820 \ln(N) + 1.828 \ln(BA)$, N is number of culms per hectare, MDBH is mean diameter of breast area, BA is basal area per hectare.

A perceivable diagonal line implies that the observations are equal to the predicted values, which also helps understand the bias of the models. According to the distribution of dots and a perceivable diagonal line, Models II and III had better prediction effects than Model I because the distribution of dots was closer to the line in Models II and III than Model I (Figure 2). Models II and III displayed a similar distribution pattern in Figure 2. Moreover, we used RMSE to show the bias between observed AGCS and predicted AGCS. The results are shown in Table 3.

Table 3. The root mean square error (RMSE) between observed aboveground carbon storage (AGCS) and predicted AGCS for three models.

Model ¹⁾	Number of samples	Observed AGCS (Mg ha ⁻¹)	Predicted AGCS (Mg ha ⁻¹)	RMSE (Mg ha ⁻¹)
I	32	17.11 ± 8.16	17.03 ± 7.77	1.717
II	32	17.11 ± 8.16	17.09 ± 8.06	0.525
III	32	17.11 ± 8.16	17.12 ± 8.07	0.523

¹⁾Model I is $\ln(ABCS) = 0.933 + 0.882 \ln(BA)$, Model II is $\ln(ABCS) = -0.004 + 0.820 \ln(MDBH) + 1.008 \ln(BA)$, Model III is $\ln(ABCS) = 7.550 - 0.820 \ln(N) + 1.828 \ln(BA)$, N is number of culms per hectare, MDBH is mean diameter of breast area, BA is basal area per hectare.

From Table 3, Models II and III had smaller RMSE values, indicating better prediction effects. This phenomenon also reflected the results of Figure 2, where predicted values matched a perceivable diagonal line more closely.

4. Discussion

Bamboo plantations consist of individuals of various age classes, which display an uneven-aged structure [8,18–21,33,34]. This structure results from their development pattern and management approach [8,18,34]. The development of bamboo plants is based on asexual reproduction by rhizome, and new individuals sprout out year by year. Therefore, harvesting or thinning old culms (usually over 4-5 years old) is necessary because it helps increase growth space for new culms and maintain vitality for the entire stands [21,30,32–34]. Because thinning is strongly related to stand development, it is crucial for stocking bamboo plantations. It indicates that thinning old culms in bamboo

plantations management is necessary, regardless of whether their management purpose is on harvesting culms or bamboo shoots.

On the other hand, fertilization is another approach to improve productivity for bamboo plantations. Intensive management (IM) and extensive management (EM) are two critical strategies widely used in bamboo plantations [30,34,46]. IM involves thinning and fertilization, while EM only includes thinning. Numerous studies have confirmed higher productivity in bamboo plantations performing IM [8,30,46,47]. However, the productivity may only partially reflect the stocking of bamboo plantations. If farmers harvest a significant number of bamboo shoots and only keep a few for further developing culms, bamboo plantations with IE might show lower stocking. It indicates that productivity does not fully reflect stocking because it is affected by the factor of harvesting bamboo shoots. As a result, biomass or carbon stocking does not directly correlate with bamboo plantations performing IE or ME but depends on farmers' decision to harvest culms or bamboo shoots [30,46]. Usually, bamboo plantations for culm harvest have higher stocking while for shoot harvest have lower stocking, regardless of IM or EM.

Despite most bamboo plants possessing wooden structures, they differ from timber trees in stems. Bamboo plants have hollow structures in culms, while boles of timber trees are solid [33,48]. Therefore, weight or biomass is better than volume when measuring bamboo plants to evaluate stocking or productivity [33,48]. Numerous studies have proposed using the allometric model for predicting bamboo plants because this approach has many advantages, such as the model being easy to use, the parameters with special meanings to explain biomass accumulation and a high predictive ability between DBH (or DBH and culm height) and biomass [18,49,50]. As a result, the allometric model plays a critical role in biomass prediction in bamboo studies [9,18,21,23,28,29,33,34,36–38,47,50–53].

The target of the allometric model usually addresses a certain bamboo species or a combination of various bamboo species. The development of an allometric model should consider DBH distribution and age class for sampling because bamboo plantations display uneven-aged structure [18,21]. Suppose a high allometric relationship exists between DBH and biomass in samples. In that case, the model can effectively scale out for the whole stands and extend to the same species, obtaining the biomass of the entire stands from the summation of individuals. Consequently, researchers could easily predict carbon yield because carbon storage is approximately half of the biomass [9,11,15,16,18,21,23,29,33,34,36–38,50,52]. Numerous researchers evaluated carbon yield based on the above processes, and the present study also followed it to predict AGCS. Since Makino bamboo is a critical bamboo in Taiwan, the allometric model and PCC have been developed and determined in a previous study by Yen et al. [21]. Therefore, this study adopted this allometric to predict AGB, cited PCC to determine AGCS for Makino bamboo, and obtained a range from 7.16 to 28.37 Mg ha⁻¹. Moreover, Liu and Yen [25] reviewed the published papers and obtained 12 records of AGCS with 22.22 ± 24.66 Mg ha⁻¹ for Makino bamboo plantations. We found a significant standard deviation in the AGCS. As mentioned above, the variations in the stocking resulting from farmers' management purposes might lead to a substantial variation in Makino bamboo plantations, which covered the range of the AGCS predicted in our study.

Even with the current stocking of bamboo plantations determined by farmers' attitudes, researchers can obtain biomass or carbon yield through the allometric model. At the stand levels, N, BA, and MDBH are critical factors that affect AGCS [21,25]. Liu and Yen [25] used these three factors to develop a carbon yield model for bamboo plantations of various species and found that the factors effectively predicted AGCS. The present study referred to the same factors proposed by Liu and Yen [25] to develop carbon yield models for a single species, obtaining a satisfactory result for Makino bamboo. In Table 3, we found a high R_{adj}^2 over 0.93 in the 27 models, based on the three factors with various types. The results confirmed that N, BA, and MDBH were crucial factors in predicting AGCS. Even using only two variables (ln(DBH) and ln(BA) or ln(N) and ln(BA)) can obtain the highest R_{adj}^2 . From these two best predictive models, stand density played an essential role in predicting ABCS. Usually, N and BA are crucial variables representing stand density for forests, regardless of timber tree forests and bamboo plantations [21,25,54]. Our results showed that combining N and BA or using a single BA and MDBH effectively predicted AGCS. It indicated that BA played a critical role in AGCS prediction because this variable was simultaneously shown in the two best predictive models, and even only using this variable had a significant predictive ability (Figure 3).

The study used various thinning intensities to create different stand characteristics for a Makino bamboo plantation, especially in a wide range of stand densities and AGCS (Table 1). We used the data to develop the AGCS model and found a good fit for the models, indicating that these models can effectively be used to predict AGCS in this study area. However, if researchers would like to extend the model to other regions, adding more data from such areas to develop is necessary.

5. Conclusions

This study aimed to develop an AGCS model for Makino bamboo. We collected data from a Makino bamboo plantation with various thinning treatments and used the factors of N, BA, and MDBH to predict AGCS. The model used multiple regression to develop based on the stepwise method. The following conclusions were obtained:

1. We used thinning treatment to create different stand variables: N from 11,500 to 80,900 culms ha⁻¹, MDBH from 1.95 to 2.70 cm, and BA from 3.11 to 15.96 m² ha⁻¹.
2. According to the AGB and PCC, the ranges of AGB and AGCS for various treatments ranged from 15.32 to 60.68 and 7.16 to 28.37 Mg ha⁻¹, respectively.
3. The best two predictive models were $\ln(AGCS) = -0.004 + 0.820 \ln(MDBH) + 1.008 \ln(BA)$ and $\ln(AGCS) = 7.550 - 0.820 \ln(N) + 1.828 \ln(BA)$. It indicated that either BA and MDBH or BA and N can effectively predict AGCS.
4. The best predictive models showed that the factor of stand density was critical in affecting AGCS, especially the variable BA. Even using this variable alone had significant predictive ability.
5. This study's limitation was the model developed for a regional area. If researchers want to extend the model to other regions, adding more data from such areas is necessary for development.

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