

1 Proceedings

# 2 Models to Estimate the Bark Volume for *Larix* sp. in 3 Poland

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9 Received: date; Accepted: date; Published: date

10 **Abstract:** Bark constitutes one of the main residues and by-products of the timber harvest.  
11 Therefore, in order to conduct effective forest management, it is essential to estimate the possible  
12 amount of that product, which can be obtained during the harvest process. Our objective was to  
13 develop a model to estimate bark volume and bark volume fraction (i.e. its share in the total  
14 volume of a tree). For the study we choose larch (*Larix* sp.) that is a rare but valuable forest raw  
15 material in Poland. The research material was collected in northern (2 sites), central (1 site), and  
16 southern (2 sites) Poland. In total, we obtained data from nearly 600 trees growing on oligo-, meso-  
17 and eutrophic sites. We used tree's breast height diameter, height and total volume as independent  
18 variables. Both analysed bark parameters varied significantly with regard to location, site type and  
19 age class. Bark volume is strongly and significantly dependent on tree's breast height diameter,  
20 height and total volume. For bark volume fraction this correlation is significant but very weak. The  
21 best results of bark volume estimation are achieved for model with total tree volume as  
22 independent variable. Because of the strong effect of location on bark volume estimates, it is  
23 recommended to elaborate locally-based models for this parameter determination.

24 **Keywords:** bark volume; bark volume fraction; larch; modelling

## 25 1. Introduction

26 As timber sale is the primary source of income in the forestry, it is crucial to estimate timber  
27 volume and its value with the highest achievable accuracy. In Central Europe, it is a common  
28 practice to sell the timber with the bark, however the customer pays for the volume estimated under  
29 the bark. The conversion of over-bark measurements to under-bark records is made using various  
30 methods of bark thickness or bark volume determination [1,2]. Therefore, in addition to the  
31 improvement of the measurement methods and equipment, efforts are undertaken to develop better  
32 and better models that allow to determine bark features precisely [3].

33 Recent shift of the bark reception from a harvest by-product or residue towards the perspective  
34 and commercially important fuel or biomaterial and source of tannins caused the increase interest in  
35 bark volume estimates development [4-6]. Moreover, assessment of bark volume is also important  
36 for quantifying carbon stocks [7].

37 Our objectives included: (i) analysis of the variability of bark volume and bark volume fraction  
38 for *Larix* sp. in Poland as well as (ii) development of models to estimate the investigated parameters  
39 with regard to the basic dendrometric attributes.

## 40 2. Material and methods

41 Data that served for bark volume (bV) and bark volume fraction (%bV) modelling was collected  
42 in 5 locations in various parts of Poland: Dobrzany and Kolbudy – northern part, Rogów – central  
43 Poland, Pińczów and Prudnik – southern, upland and mountain part of the country. Altogether we

44 measured 599 trees on 62 study plots that differed in growth conditions (oligo-, meso- and eutrophic  
45 sites) and age (19-127 years).

46 For each tree we obtained its breast height diameter (d), height (h) and total volume (V,  
47 determined with section-wise method). Using bark gauge we determined the bark thickness that  
48 allowed to convert the over-bark volume to under-bark volume. The difference of these two  
49 constituted bark volume. Bark volume fraction was calculated as a ratio of bark volume and total  
50 over-bark volume of a tree.

51 For the distribution of bV and %bV differed significantly from the normal one (Shapiro-Wilk  
52 test,  $p < 0.001$ ), we used Kruskal-Wallis test to assess the impact of location (5 variants), site type (3  
53 variants of growth conditions) and age class (we distinguished 4 ones: <40, 40-60, 60-80, >80  
54 years-old) on the analysed bark attributes. Pearson correlation was applied to evaluate the  
55 relationship between bV or %bV and d, h or V as well as between one another.

56 Following previous studies [5-6] we used breast height diameter, height and total tree volume  
57 as an independent variables in models to estimate bark volume and bark volume fraction. We chose  
58 the following equations for the model elaboration:

$$\hat{y} = a + b \cdot x, \quad (1)$$

$$\hat{y} = a \cdot x^b + c, \quad (2)$$

$$\hat{y} = a \cdot (\exp^x \cdot b) + c, \quad (3)$$

$$\hat{y} = a \cdot x / (b + x), \quad (4)$$

$$\hat{y} = a / (1 + b \cdot \exp^{-c \cdot x}), \quad (5)$$

$$\hat{y} = a \cdot \exp^{(b \cdot \exp^c \cdot x)}, \quad (6)$$

59 where:  $\hat{y}$  – estimated bark parameter,  $x$  – independent variable (d – breast height diameter, h –  
60 height, V – total tree volume), a, b, c – model parameters.

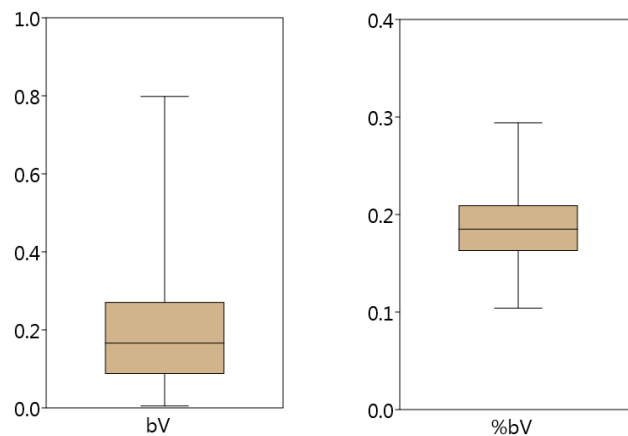
61 Based on bark volume distribution we split our data into calibration and validation sets in  
62 proportion  $2/3$  to  $1/3$ . Obtained sets did not differ significantly in terms of tree's age, breast height  
63 diameter, height, total tree volume, bark volume and bark volume fraction (Mann-Whitney test;  $p$   
64  $> 0.4$ ). Best model selection was based on AIC and  $R^2$  goodness-of-fit measures. We chose two best  
65 performing ones and verified them based on the data from validation set using  $R^2$  and residuals RMSE  
66 as an evaluation measures. Finally, we tested obtained residuals for the impact of location, site type  
67 and age class (Kruskal-Wallis test).

68 All statistical analyses were performed with PAST4.03 software [8].

### 69 3. Results and discussion

#### 70 3.1. Variability of bark volume and bark volume fraction

71 Bark volume of the analysed larches ranged from 0.0048 to 0.7984 m<sup>3</sup>, with mean value  
72 amounting to 0.1985 ± 0.006 m<sup>3</sup>. Its distribution is characterized by strong positive asymmetry as  
73 skewness equals to 1.29 (Figure 1). Coefficient of variation for that attribute was high end reached  
74 75.5%. In turn, bark volume fraction was not so diversified as its coefficient of variation amounted to  
75 17.7%. Observed values varied from 0.104 to 0.294, with mean amounting to 0.188 ± 0.001. They were  
76 rather symmetrically distributed, for skewness reached 0.39 (Figure 1). These values are a little bit  
77 lower than those reported for larch in Europe [6].



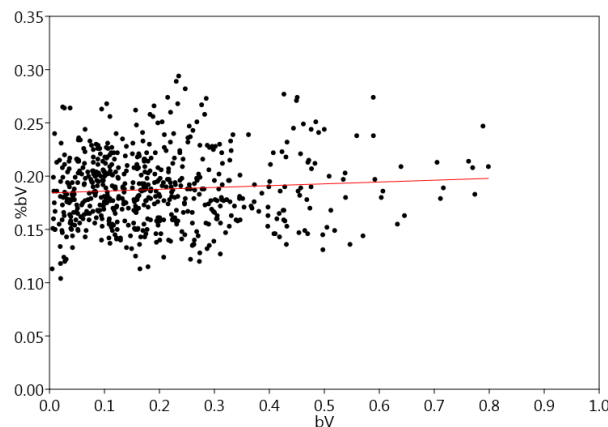
78

79 **Figure 1.** Bark volume (bV, m<sup>3</sup>) and bark volume fraction (%bV) for the *Larix* sp. in Poland. Box  
80 represents 25<sup>th</sup> percentile, mean and 75<sup>th</sup> percentile, whiskers – minimum and maximum values.

81 Both bark volume and bark volume fraction varied significantly with regard to the analysed  
82 location, site type and age class ( $p < 0.001$ ). The highest bV was observed for Rogów and Kolbudy  
83 ( $0.2891 \pm 0.019 \text{ m}^3$  and  $0.2662 \pm 0.016 \text{ m}^3$ , respectively), while the lowest for Dobrzany ( $0.1528 \pm 0.011$   
84  $\text{m}^3$ ). The more fertile site, the more bark larches have – bV for eutrophic sites equaled to  $0.2238 \pm 0.008$   
85  $\text{m}^3$ , while for oligotrophic –  $0.1364 \pm 0.012 \text{ m}^3$ . Also older have more bark than younger ones ( $0.3119$   
86  $\pm 0.015 \text{ m}^3$  (V age class) *vs.*  $0.083 \pm 0.006 \text{ m}^3$  (II age class)). For %bV the highest values were noted in  
87 Rogów ( $0.236 \pm 0.026$ ), while the lowest in Kolbudy ( $0.168 \pm 0.02$ ). The least fertile site characterize  
88 with higher values than the other sites ( $0.200 \pm 0.003$  *vs.*  $0.182 \pm 0.003$  and  $0.188 \pm 0.002$ ). In turn the  
89 oldest trees had lower %bV than the other age classes ( $0.176 \pm 0.003$ ), while the highest values were  
90 noted for IV age class ( $0.192 \pm 0.003$ ). Observed relationships confirm previous findings about  
91 dependence of bark parameters on various factors [5,6,9].

92 *3.2. Relationships between bark parameters and dendrometric attributes*

93 Bark volume was significantly and strongly correlated with d ( $r = 0.939$ ,  $p < 0.001$ ), h ( $r = 0.781$ ,  $p$   
94  $< 0.001$ ) and V ( $r = 0.956$ ,  $p < 0.001$ ). In turn, for bark volume fraction we observed weak but significant  
95 negative relationship with d ( $r = -0.106$ ,  $p < 0.001$ ), h ( $r = -0.121$ ,  $p = 0.003$ ) and V ( $r = -0.164$ ,  $p < 0.001$ ).  
96 The analysed features were insignificantly correlated one to other ( $r = 0.077$ ,  $p = 0.060$ ) (Figure 2).  
97 Similar relationships are reported for many other species in Latvia [5] or Mexico [6].



98

99 **Figure 2.** Relationship between bark volume (bV, m<sup>3</sup>) and bark volume fraction (%bV) for the *Larix*  
100 sp. in Poland.

101 3.3. Models for bark parameters estimation

102 For both investigated bark parameters, the lowest AIC values in case of each independent  
 103 variable (d, h and V) were found for linear (eq. #1) and Michaelis-Menten (eq. #4) models. Both these  
 104 equations showed similar performance for bV as well as for %bV. For bark volume, the lowest R<sup>2</sup>  
 105 and RMSE values characterized models using height as the independent value, while the highest  
 106 ones were observed for equations based on total tree volume (Table 1). Such relationship only  
 107 partially confirms previous findings as height turns to perform weaker as a bark volume descriptor  
 108 than reported by other authors [5-6].

109 **Table 1.** Goodness-of-fit measures for the best models for estimation of the bark volume (bV) or bark  
 110 volume fraction (%bV) based on tree’s breast height diameter (d), height (h) or total volume (V).

equation		bV			%bV		
		d	h	V	d	h	V
#1	R <sup>2</sup>	0.880	0.631	0.912	0.008	0.007	0.022
	RMSE	0.0515	0.0905	0.0442	0.0328	0.0328	0.0326
#4	R <sup>2</sup>	0.878	0.669	0.915	0.000	0.001	0.001
	RMSE	0.0521	0.0870	0.0435	0.0329	0.0329	0.0329

111 The goodness-of fit measures obtained for validation dataset proved the good performance of  
 112 the best models chosen based on AIC for bark volume prediction in case of d and V as independent  
 113 variables (Table 2). For height, R<sup>2</sup> and RMSE values were lower than ones calculated with calibration  
 114 dataset. Residues of the validated models were not normally-distributed and their means differed  
 115 significantly from 0 indicating systematic bias (Table 2). As models developed for %bV showed poor  
 116 relationship of this feature with d, h and V no validation was performed in that case.

117 **Table 2.** Goodness-of-fit measures for the validation of the chosen models for estimation of the bark  
 118 volume (bV) based on tree’s breast height diameter (d), height (h) or total volume (V) and  
 119 characteristics of the residues distribution (p(norm) – assessment of the distribution normality with  
 120 Shapiro-Wilk test, M – mean value, p(M=0) – Wilcoxon test p-value).

equation		d	h	V
#1	R <sup>2</sup>	0.886	0.567	0.916
	RMSE	0.0511	0.0996	0.0440
	p(norm)	<0.001	<0.001	<0.001
	M	-0.001	0.008	0.004
	p(M=0)	0.341	<0.001	0.042
#4	R <sup>2</sup>	0.900	0.605	0.915
	RMSE	0.0478	0.0951	0.0441
	p(norm)	<0.001	<0.001	<0.001
	M	0.006	0.001	0.002
	p(M=0)	0.001	0.041	0.070

121 **Table 3.** Effect (p-value in Kruskal-Wallis test) of location, site type and age class on residuals of the  
 122 chosen models for estimation of the bark volume based on tree’s breast height diameter (d),

123 height (h) or total volume (V).

equation		Location	Site type	Age class
#1	d	<0.001	0.613	0.236
	h	0.007	0.918	0.004
	V	<0.001	0.024	0.137
#4	d	<0.001	0.363	<0.001
	h	0.005	0.649	0.768

V	<0.001	0.046	0.232
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124 We found significant effect of location on the residuals of the chosen best models for estimation  
 125 of bark volume (Table 3), which indicates the necessity of elaboration of locally-based formulae. Site  
 126 type influenced significantly the residuals of the models based on V as the independent variable,  
 127 while age class affected the results of models based on tree's height.

#### 128 4. Conclusions

129 Both analysed bark parameters varied significantly with regard to location, site type and age  
 130 class. Bark volume is strongly and significantly dependent on tree's breast height diameter, height  
 131 and total volume. For bark volume fraction this correlation is significant but very weak. The best  
 132 results of bark volume estimation are achieved for model with total tree volume as independent  
 133 variable. For the weak relationship with dendrometric parameters modelling o bark volume fraction  
 134 seems to be pointless and a constant ratio should be applied. Because of the strong effect of location  
 135 it is recommended to elaborate locally-based models for bark volume estimation.

136 **Author Contributions:** Conceptualization, Sz.B. and T.Cz.; methodology, Sz.B.; K.B and M.Z.; validation, A.B.  
 137 and M.Z.; formal analysis, Sz.B.; investigation, Sz.B., T.Cz, R.W. R.T.; data curation, P.B.; writing—original draft  
 138 preparation, Sz.B.; writing—review and editing, X.X.; visualization, X.X.; supervision, X.X.; project  
 139 administration, A.B.; funding acquisition, M.Z. All authors have read and agreed to the published version of  
 140 the manuscript.

141 **Funding:** This research was partially supported by Polish State Forests , National Forest Holding within grant  
 142 number OR.5001.3.1.2017.

143 **Conflicts of Interest:** The authors declare no conflict of interest. The funders had no role in the design of the  
 144 study; in the collection, analyses, or interpretation of data; in the writing of the manuscript, or in the decision to  
 145 publish the results.

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 168 doi:10.1093/forestry/cpx047.

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