ORIGINAL ARTICLE

Optimization of thermally-compressed wood of *Trewia nudiflora* species using statistical Box–Behnken design and desirability function

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Abstract A multi-criteria decision strategy based on desirability function and Box–Behnken design as response surface methodology have been used to concomitantly optimize process parameters and two important mechanical properties in compressed wood production from *Trewia nudiflora* species. The steaming time (min), pressing time (min) and pressing temperature (°C) were selected as process parameters where as modulus of elasticity (MOE) and modulus of rupture (MOR) as responses. The optimized parameters for compressed wood production were found to be: steaming time 60 min, pressing time 10 min and pressing temperature 183.16 °C and under these conditions optimized responses were 1,229.2 N/mm² for MOE and 45.46 N/mm² for MOR.

Keywords Optimization · Mechanical properties · Heat treatment · Box–Behnken design · Desirability function · *Trewia nudiflora*

Introduction

Nowadays, thermally-compressed or densified wood is very well known and eco-friendly wood material resulted from solid wood or from laminated wood without impregnating it with any synthetic resins and chemicals. In Germany date back in 1930, compressed solid wood had been manufactured under the trade name "Lignostone" and compressed laminated wood under the trade name "Lignofol" (Seborg et al. 1962). The heterogeneous nature

Md. A. Islam (⊠) · Md. A. Razzak · B. Ghosh Forestry and Wood Technology Discipline, Khulna University, Khulna 9208, Bangladesh e-mail: iazharul@gmail.com of wood properties within the individual as well as between trees whether of the same or differing species make the wood more vulnerable to suitable use with uniformity. Compressions of wood reduce the void volume and increase the density of wood as well as increase the mechanical properties of wood and finally solve the nonuniformity problem Bowyer et al. (2003). Thus, the compressed wood materials can be used in many industrial applications across the world due to its high density and strength properties comparable to other materials available in the market. But there has an established springback problem when this compressed wood come to contact with water (Kutnar and Kamke 2012a; Navi and Girardet 2000; Hsu et al. 1988). There has been several substantial and convincing work proposed by several researchers in order to overcome this springback problem of thermally compressed wood (Fang et al. 2012; Laine et al. 2012; Dwainto et al. 1997; Inoue et al. 1993a, b).

The overwhelming growth of human population is causing an increasing pressure on forests with high quality timber for furniture, joinery and construction purposes. On the other hand, the forest of Bangladesh has nearly 500 hardwood species but only 40 species are producing high quality timber to fulfill the above mentioned purposes (Sheikh et al. 1993). Thus a vast majority of the timber species are not potentially utilized or unutilized due to their inferior properties. Trewia nudiflora (Family: Euphorbiaceae) is a moderate sized to large deciduous tree having low density, light, fine even textured and straight grained timber mainly used for fuel wood, picture frames, agricultural implements, toys, planking, and match manufacture (Alam et al. 1991). According to Norimoto (1993) and Wang et al. (2000) the first growing low density and low quality timber species can be used to produce compressed wood for their potential utilization. Thus, timber of T.

nudiflora species can be effectively converted into high quality compressed material to reduce the burden on the over exploited conventional species.

The quality of compressed wood depends on several working parameters such as: pressing time, pressing temperature, pressure, heating rate, types of species, pretreatment of the sample, uses of additives, treatment after thermal pressing etc. (Santos et al. 2012). Among these parameters; pressing time, pressing temperature and steaming time has been known to be important according to literature review (Anshari et al. 2011; Rowell et al. 2002; Inoue et al. 1993a, b; Hsu et al. 1988). It is also noted that the compression of wood has a positive impact on strength properties such as modulus of rupture (MOR) and modulus of elasticity (MOE) of wooden products. The MOR measures the overall strength of a wooden samples but MOE measures the wood's deflection rather than eventual strength. Therefore, dealing with many factors at a time with several responses is a great challenge for any quality products because most of the industrial and manufacturing quality characteristics are multidimensional. This challenge can be carefully handled firstly by building appropriate response surface methodology (RSM) for each response and later attempt to find a set of operating conditions that optimizes all responses at least within the range (Zadbood and Noghondarian 2011; Raissi and Eslami 2009). In order to resolve the multiresponse problems several researchers proposed versatile methodology (Cornell and Khuri 1987; Tai et al. 1992; Pignatiello 1993; Layne 1995; Byrne and Taguchi 1987).

The desirability function approach is one of the most frequently used multi-response optimization techniques in practice developed by Harrington (1965) and was later modified by Derringer and Suich (1980). Desirability function is intuitive, simple and available in all statistical software. Therefore, in this study, Derringer's desirability function as multi-criteria decision making approach was used to evaluate the compressed wood properties. Desirability function simultaneously determines the optimum settings of input variables that lead to desired compromise solutions. The theoretical concepts and application of multivariate optimization using multi-objective desirability function have been discussed intensively in a number of illuminating articles (Islam et al. 2009, 2010, 2012; Costa et al. 2011; Jeong and Kim 2009).

Over the last century, considerable research has been conducted on different aspects of compressed wood (Santos et al. 2012; Kutnar and Kamke 2012b; Mohebby and Bami 2011; Navi and Heger 2004) but there has no work in optimizing the manufacturing parameters and responses simultaneously using RSM and desirability function for the production of compressed wood. Thus, the objective of this study was to optimizing and manufacturing of high quality compressed wood from *T. nudiflora* species by using response surface methodology and desirability function.

Materials and methods

Sample preparation for compressed wood production

Trewia nudiflora tree was collected as raw material for the production of compressed wood with varying density ranges from 0.40 to 0.44 g/cm³. The height of the tree was about 8 m and diameter was 45 cm. For the purpose of preparing samples the middle portion of the tree was selected which was clean and straight. The dimension of samples was $22 \times 22 \times 4$ cm. In this study, fifteen samples were prepared for fifteen experimental trials that were designed using Box-Behnken design. The finished samples were then kept in a plastic bag at room condition before any further treatment. The first operation of the experiment was steaming which was carried out using an autoclave (Jeiotech, Korea, model-AC-11) at constant temperature 121 °C and pressure 0.18 MPa, respectively. After steaming the samples were conditioned for 30 min at room condition and then oven dried for 24 h at 103 °C \pm 2. In the next step, pressing of the steamed samples was carried out using laboratory hot-press (model-XLB, Manufacture no. 120049, Qingdaoyadong, China) at a constant pressure of 6 MPa. After compression samples were conditioned for 7 days at room condition and cut into desired dimension for the determination of MOR and MOE by Universal Testing machine (Model-UTN100, SR no. 11/98-2441, Fuel Instruments and Engineers Pvt. Ltd., India) according to ASTM D 4761-11 (ASTM 2011).

Response surface methodology and multi-response optimization

Response surface methodology (RSM) is a collection of mathematical and statistical techniques useful for the modeling and analysis of problems in which a response of interest is influenced by several variables and the objective is to optimize this response and to find a suitable approximation for the true functional relationship between the variables and the responses (Montgomery and Douglas 2005). RSM can be used to improve quality through reducing the variability with an improved process and product performance with fewer numbers experimental runs to evaluate the effect of different factors and their interactions (Zadbood and Noghondarian 2011). Building a response surface methodology needs 3 steps of sequential experimentations:

Table 1 Three factor Box-Behnken design matrix for compressed wood manufacture

Factors		Range a	Range and label				
		Low (-	-1)	Middle (0)	High (+)		
Steaming time (min) Pressing time (min) Pressing temperature (°C)		60		120 15 180			
		10					
		170					
Run	Steaming time	Pressing time	Pressing temperature	MOE (N/mm ²)	MOR (N/mm ²)		
1	-1	-1	0	1,310.59	46.58		
2	1	-1	0	476.93	29.06		
3	0	-1	1	740.03	36.19		
4	0	1	-1	754.66	46.97		
5	-1	1	0	755.13	35.89		
6	1	0	-1	445.11	31.04		
7	0	1	1	720.44	30.40		
8	-1	0	1	684.97	36.04		
9	0	-1	-1	640.03	34.18		
10 (C)	0	0	0	784.99	39.28		
11 (C)	0	0	0	804.77	39.63		
12 (C)	0	0	0	848.85	41.38		
13	-1	0	-1	496.87	35.02		
14	1	0	1	691.11	30.31		
15	1	1	0	986.93	38.06		

(C) Center points run

- (i) Designing a series of experiment for data collection through selection of appropriate factors and their levels.
- (ii) Sequential mathematical model fitting.
- (iii) Optimization in order to find the optimal sets of experimental parameters to produce best output of the response by analyzing the response model.

Similarly, the multi-response problem consists of three stages: data collection, model building, and optimization. Here data collection and model building procedures are followed based on RSM and multi-response optimization was performed by desirability function.

In this study Box–Behnken design was used as response surface methodology to determine the relationship between the variables and response functions MOE and MOR. Box– Behnken design is a fractional factorial design that allows establishing statistical relationship between experimental variables and response (test results), from which the optimal experimental conditions could be predicted for achieving the optimal response. Besides, Box–Behnken design needs few numbers of experiments resulting low experimental costs, more efficient and easier to arrange as well as interpret in comparison to others. Box–Behnken design has specific positioning of the design points. So this design is useful in avoiding treatment combinations that are extreme (Islam et al. 2009). Three important parameters steaming time, hot press time, hot press temperature and their three levels for each have been selected in this study according to the literature review (Anshari et al. 2012; Inoue et al. 1993a, b; Inoue and Norimoto 1991). Table 1 show three factors (steaming time, hot press time, hot press temperature) and three levels of each factor coded -1, 0, and +1 for low, middle and high values, respectively. STATISTICA[®] statistical software was used and a second order polynomial model used to fit the response to the independent variables is shown below.

$$Y = \beta_0 + \sum \beta_i X_i + \sum \beta_{ii} X_i^2 + \sum \beta_{ij} X_i X_j, \tag{1}$$

where *Y* is the response (MOE or MOR), β_0 is the intercept and β_i , β_{ii} , β_{ij} are the coefficients of parameters for linear, squared and interaction effects, respectively.

In the next step, Derringer desirability function was used as multi-criteria decision making approach in order to compromise between MOE and MOR resulting from selected factors based on RSM mathematical models that might convert a multi-response problem into a singleresponse. The Derringer's desirability function, D, is the geometric mean of the individual desirability.

The desirability function first converts the response (y_i) into an individual desirability function (d_i) that varies from

Table 2 ANOVA for MOE	Source	Sum of squares	df	Mean square	F value	Prob > F	Remarks
	(1) Steaming time (L)	52,403.81	1	52,403.80	49.03	0.019	Significant
	Steaming time (Q)	3,870.62	1	3,870.64	3.62 0.28	0.197 0.645	
	(2) Pressing time (L)	307.30	1	307.34			
	Pressing time (Q)	38,337.72	1	38,337.70	35.87	0.026	Significant
	(3) Pressing temperature (L)	31,235.03	1	31,235.00	29.22	0.032	Significant
	Pressing temperature (Q)	149,146.71	1	149,146.70	139.54	0.007	Significant
	1L by 2L	283,801.30	1	283,801.32	265.53	0.003	Significant
	1L by 3L	838.11	1	838.11	0.78	0.469	
	2L by 3L	4,503.80	1	4,503.84	4.21	0.176	
$R^2 = 0.9052$; adjusted	Lack of fit	58,100.44	3	16.81	18.12	0.052	Not significant
$R^2 = 0.7348$	Pure error	2,137.60	2	1,068.80			
<i>L</i> linear effect; <i>Q</i> quadratic effect	Total SS	636,019.84	14				

0 to 1. The desirability 1 is for maximum and desirability 0 is for non-desirable situations or minimum. Derringer and Suich (1980) proposed the equation for three response types, which can be expressed as:

(1) Nominal-The-Best: the estimated response is expected to get a particular target value (T).

$$d_{i} = \left[\frac{y-L}{T-L}\right]^{wi} \text{ or } \left[\frac{y-H}{T-H}\right]^{wi} L \le y \le T,$$

when $y < L$ or $y > H$ then $d_{i} = 0$
and $y = T$ then $d_{i} = 1.$ (2)

(2) Larger-The-Best: the estimated response is expected to be larger than a lowest boundary.

$$d_i = \left[\frac{y-L}{H-L}\right]^{w_i}, \quad L \le y \le H, \quad \text{when } y < H$$

then $d_i = 0$ and $y > H$ then $d_i = 1.$ (3)

(3) Smaller-The-Best: the estimated response is expected to be smaller than a highest boundary.

$$d_i = \left[\frac{y-H}{L-H}\right]^{w_i}, \quad L \le y \le H, \quad \text{when } y > H \qquad (4)$$

then $d_i = 0$ and $y > L$ then $d_i = 1$.

In Eqs. (2), (3) and (4), L and H are the lowest and the highest values, respectively obtained from the mathematical models in RSM and w_i is the weight. Above the both cases, d_i will vary non-linearly while approaching the desired value but with a weight of 1, d_i varies linearly.

Subsequently the transformations of all individual desirability points for the predicted values are converted into overall desirability function, D, by computing their geometric mean by following equation.

$$D = [d_1 \times d_2 \times \dots \times d_n]^{1/n} \tag{5}$$

The above equation can be extended when the weight or importance has been considered:

$$D = \begin{bmatrix} d_1^{r_1} \times d_2^{r_2} \times \dots \times d_n^{r_n} \end{bmatrix}^{1/n}, 0 \le r_i \le 1 \ (i = 1, 2, \dots, n), \quad \sum_{i=1}^n r_i = 1,$$
(6)

where r_i represents the importance that reflects the difference in the importance of different responses. Importance (r_i) varies from the least important, a value of 1, to the most important, a value of 5 and the outcome of the overall desirability D depends on the r_i values.

Results and discussion

Analysis of Box–Behnken design and adequacy of the model

In this study in order to evaluate the linear, squared and interaction effects of the selected factors on mechanical properties of compressed wood, a second order polynomial model were developed using Box-Behnken design. The significant terms in the model were evaluated by Analysis of variance (ANOVA) that presented in Tables 2 and 3 for MOE and MOR, respectively. The model adequacy was checked by the coefficient of determination, R^2 , a global statistic parameter to access the fit of a model. A coefficient of determination (R^2) value of 0.90 for MOE and 0.897 for MOR is highly agreement with the experimental results, indicating 90 and 89.7 % of the variability can be revealed by the model and are left with 10 and 10.3 % residual variability for MOE and MOR. For further validation of the model, adjusted R^2 was used for confirming the model adequacy. The adjusted R^2 was calculated to be 0.734 and

Table 3 ANOVA-MOR for response surface quadratic	Source	Sum of squares	df	Mean square	F value	Prob > F	Remarks
model	(1) Steaming time (L)	78.56	1	78.56	61.94	0.015	Significant
	Steaming time (Q)	39.41	1	39.41	31.07	0.030	Significant
	(2) Pressing time (L)	3.52	1	3.52	2.77	0.237	
	Pressing time (Q)	1.18	1	1.18	0.93	0.436	
	(3) Pressing temperature (L)	25.47	1	25.47	20.08	0.046	Significant
	Pressing temperature (Q)	51.39	1	51.39	40.52	0.023	Significant
	1L by 2L	96.95	1	96.95	76.43	0.012	Significant
	1L by 3L	0.75	1	0.75	0.59	0.519	
	2L by 3L	86.15	1	86.15	67.92	0.014	Significant
$R^2 = 0.8972$; adjusted	Lack of fit	41.00	3	13.66	10.77	0.086	Not significant
$R^2 = 0.7121$	Pure error	2.53	2	1.26			
<i>L</i> linear effect; <i>Q</i> quadratic effect	Total SS	423.60	14				



Fig. 1 Observed versus predicted plot for a MOE and b MOR

0.712 for MOE and MOR, which indicates a good model for using in the field conditions.

The adequacy of the model was further evaluated through ANOVA lack of fit (LOF) test. The lack of fit (LOF) is the variation of the data around the fitted model. The significant lack of fit explained that the model doesn't describe the trend in the data well enough and suggests new higher order model in order to explain the data but in this study non-significant LOF relative to the pure error, indicating good response to the model.

After model fitting, in order to validate the assumptions drawn in the ANOVA, the normality of

residuals was determined by numerically with Shapiro-Wilk normality test due to its more appropriateness for small sample sizes (<50 samples). If the *P* value of the Shapiro-Wilk test is greater the 0.05, the data is normal. If it is below 0.05, the data significantly deviate from a normal distribution. In this study, Shapiro-Wilk normality test gave non-significant value of *W* statistics (W = 0.985, p = 0.993 for MOE and W = 0.975, p = 0.926 for MOR), indicating model may predict very well. The model was again validated by the observed versus predicted plot. The points of all predicted and actual responses (Fig. 1a, b) fell in 45° inclined straight line indicating also good response to the model.

The final predicted mathematical models in terms of significant actual factors for compressed wood production governed by different factors are given below:

MOE = -63,205.71 - 14.78 * steaming time - 0.008* steaming time² - 102.3 * pressing time + 3.9* pressing time² + 724 * pressing temperature

- $-1.972 * \text{pressing temperature}^2 + 0.888$
- * steaming time * pressing time + 0.012
- * steaming time * pressing temperature -0.671
- * pressing time * pressing temperature

MOR = -1,377.042 + 0.050 * steaming time - 0.001* steaming time² + 14.192 * pressing time + 0.023 * pressing time² + 14.732 * pressing temperature - 0.037 * pressing temperature² + 0.016 * steaming time

- * pressing time -0.001 * steaming time
- * pressing temperature -0.093 * pressing time
- * pressing temperature 0.095 * pressing tim
- * pressing temperature





Fig. 3 3D graph showing the positive effect of pressing time and steaming time on MOE

Effects of different factors on MOE

The importance of different actual factors and their interactions are showed by standardized pareto chart (Fig. 2) for selected response MOE. The vertical line which overpass through the standardized factors determine the statistical significance at 95 % confidence intervals. The sign + and – reflects the positive and negative effect of the corresponding factors, respectively. Positive coefficients indicate the MOE is favored and negative coefficients indicate unfavorable by the factors and their interactions.

In this study, it has been observed from ANOVA for MOE (Table 2) and standardized pareto chart (response of MOE) that the linear effect of steaming time and pressing

temperature, the quadratic effect of pressing time and pressing temperature has significant impact on the MOE. It has been also showed that the linear interaction effect between steaming time and pressing time (i.e. 1L by 2L) has the most significant impact on MOE.

From the Fig. 3, it has been attributed that the synergistic effect of lowest steaming time and lowest pressing time produced compressed wood with high MOE. It has also been observed that the MOE of compressed wood attained the highest value 1,400 N/mm² for low pressing time ranges 8-10 min and low steaming time ranges 40-70 min. As the steaming time and pressing time increases, the MOE of compressed wood decreases gradually. Jimenez et al. (2011) showed that MOE decreases with the increases of treatment time. Similar observation was found by Esteves and Pereira (2009). They observed that MOE increases with softer treatment and decreases with severe treatment. Mitchell (1988) studied that MOE decreases regularly with the time of treatment. MOE and bending strength decreases with the increases treatment time and temperature (Esteves et al. 2007). Kubojima et al. (2000) reported that the MOE and the bending strength increases in the beginning of the treatment and decreases afterwards more for the treatments.

Effects of different factors on MOR

The actual factors which have significant effect on MOR of compressed wood are shown in ANOVA for MOR (Table 3) and also represented by standardized pareto chart which is developed by the software depicted in Fig. 4.

chart for MOR



50 45 40 35 MOR 30 25 20 45 15 40 2 35 $-\infty$ 30 100 25 × 3 0 00 R.

Standardized Effect Estimate (Absolute Value)



Fig. 5 3D graph showing the positive effect of steaming time and pressing time on MOR

It has been observed from ANOVA for MOR and standardized pareto chart (response of MOR) that the linear effect of steaming time and pressing temperature, the quadratic effect of pressing temperature and steaming time has significant impact on MOR. It has been also observed that the positive linear interaction effect between steaming time and pressing time (i.e. 1L by 2L), the negative linear interaction effect between pressing time and pressing temperature (i.e. 2L by 3L) has the most significant impact on MOR.

It has been observed from the Fig. 5 that the MOR of compressed wood attained the highest value 45 N/mm² for low pressing time ranges 8-11 min and low steaming time ranges 40-90 min. As the pressing time and steaming time start to increase from 11-22 and 90-200 min, respectively,

Fig. 6 3D graph showing the negative effect of pressing temperature and pressing time on MOR

the MOR start to decrease sharply. Jimenez et al. (2011) described in their investigation that MOR decreases with the increase of treatment time of thermally modified wood. This strength (MOR) loss in wood is actually related to the progressive degradation of hemi-cellulose components due to steaming and this reduction increases with increasing steaming duration (Yilgor et al. 2001; Boonstra and Tjeerdsma 2006).

From the standardized pareto chart (Fig. 4) a negative interaction effect exists between pressing temperature and pressing time which depicted in (Fig. 6). Kim et al. (1998) described that there is a close relationship between the loss of MOR and the process condition (time and temperature).



Fig. 7 Profiles for predicted values and desirability for simultaneous optimization of compressed wood

In another study Inoue et al. (1993a, b) reported that MOR depends more on treatment temperature rather than time.

The above figure illustrates that high MOR values obtained when the pressing temperature ranges between 182 to 190 °C for short pressing time ranges 8–9 min and at low pressing temperature for long pressing period ranges 20–22 min. The Fig. 6 also demonstrates that the MOR start to decrease with the increase of temperature above 190 °C and the combination of high pressing temperature and time has detrimental effect on MOR.

The above findings are very much similar to the previous findings. Many investigations and researches explained that a decrement of MOR occurred with the increment of treatment temperature around and above 200 °C (Poncsak et al. 2006; Kamdem et al. 2002). Dubey (2010) had reported that MOR decreases slightly at 180 °C for long pressing time and decreases sharply at 210 °C for all treatment time. This decrement of MOR is due to the depolymerization reactions of wood polymers (Kotilainen 2000). Curling et al. (2002) described that bending property losses are highly associated with the kind of carbohydrate being degraded whereas MOR decreases with the decrease of cellulose. Cellulose degraded to a lesser extent compared to the hemicelluloses of thermally treated wood but the degradation is very significant at 210 °C, therefore greater loss of MOR (Dubey 2010).

Optimization by using desirability function

The optimization process was done by selecting software profile and desirability option. The desirability function first took the maximum and minimum values of responses from the statistical model based on Box Behnken design. The general function optimization procedure is used (the simplex method of function optimization) to find the optimal settings of the independent variables (within the specified experimental range) for overall response desirability (STATISTICA 2004). From Table 1, the maximum MOE (1310.6) was assigned as desirability 1.0, minimum (445.11) as desirability 0.0 and middle (877.85) as desirability 0.5 (Fig. 7). In a similar ways, for MOR maximum (46.971) assigned as desirability 1, minimum (29.06) as desirability 0.0 and middle (38.016) as desirability 0.5. Then, the predicted MOE and MOR at each levels, holding all other factors constant at their current setting are calculated individually (Table 4), and this individual

 Table 4
 Factor levels and predicted responses of desirability function for compressed wood

Factors	Factor level	Predicted MOE	Predicted MOR	Desirability value	-95 % CI	+95 % CI	-95 % CI	+95 % CI
Steaming time (min)	60.00	1,229.20	45.46	0.91	1,104.17	1,354.24	41.16	49.77
Steaming time (min)	90.00	1,082.12	43.82	0.78	987.57	1,176.68	40.56	47.07
Steaming time (min)	120.00	918.86	40.54	0.59	829.13	1,008.58	37.44	43.63
Steaming time (min)	150.00	739.40	35.62	0.35	644.84	833.96	32.36	38.88
Steaming time (min)	180.00	543.75	29.07	0.01	418.72	668.79	24.77	33.38
Pressing time (min)	10.00	1,229.21	45.46	0.91	1,104.17	1,354.24	41.16	49.77
Pressing time (min)	12.50	1,017.40	42.18	0.70	922.84	1,111.96	38.92	45.43
Pressing time (min)	15.00	856.54	39.17	0.52	766.82	946.26	36.08	42.26
Pressing time (min)	17.50	746.63	36.45	0.38	652.07	841.19	33.19	39.71
Pressing time (min)	20.00	687.67	34.01	0.28	562.64	812.71	29.70	38.32
Pressing temp (°C)	170.00	940.95	37.77	0.53	774.76	1,107.14	32.05	43.50
Pressing temp (°C)	175.00	1,132.47	42.22	0.76	1,001.97	1,262.96	37.72	46.71
Pressing temp (°C)	180.00	1,223.50	44.79	0.89	1,101.68	1,345.31	40.60	48.99
Pressing temp (°C)	185.00	1,214.03	45.51	0.90	1,083.54	1,344.53	41.01	50.00
Pressing temp (°C)	190.00	1,104.08	44.36	0.81	937.89	1,270.27	38.63	50.08

CI confidence interval

desirability scores for the predicted values for each MOE and MOR are then combined by computing their geometric mean according to Eq. (5). In this way, by using "larger the best" desirability approach provided the best optimized conditions for steaming time 60 min, pressing time 10 min and pressing temperature 183.16 °C that also simultaneously optimised 1,229.2 N/mm² for MOE and 45.46 N/ mm² for MOR of compressed wood product. The individual desirability scores of each parameter are illustrated in Fig. 7 (bottom).

After optimisation of each factors level and with their corresponding MOE and MOR, another single compressed wood was manufactured in order to check the repeatability of predicted desirable values. The data obtained from predicted desirability optimization (steaming time 60 min, pressing time 10 min and pressing temperature 183.16 °C that produces compressed wood of 1,229.2 N/mm² for MOE and 45.46 N/mm² for MOR) are closely related with the data obtained from real life application of above factors (steaming time 60 min, pressing time 10 min and pressing temperature 183.16 °C) in the production of compressed wood having MOE of 1,180.2 and 43.25 N/mm² for MOR. It is also mentionable that the MOE and MOR values of T. nudiflora wood without any treatments (control) were 419.34 and 12.60 N/mm², respectively while on the contrary after the treatment the MOE ranged from 445.11 to 1,310.6 N/mm² and MOR 29.06 to 46.58 N/mm². The average density also increased from 0.49 to 0.70 g/cm³. Therefore, these results explained the suitability of uses selected wood as well as modification procedures.

Conclusion

The Box–Behnken design offer a better insight into the effects of selected three parameters steaming time, pressing time and pressing temperature on two mechanical properties MOE and MOR of compressed wood. The intent parameters as well as responses MOE and MOR were optimized simultaneously by use of Derringer's desirability function. Finally, by using optimized parameters level a set of experiment was performed in order to see the model predictability. The experimental results obtained under optimized parameters were very close to the theoretical results, indicating suitability of the design within the range of parameters investigated. Likewise, *T. nudiflora* is a very potential tree species for quality compressed wood production.

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