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ARTICLE

***Trilepisium madagascariense* fruit-wastes as cheap feedstock for bioethanol production**

Ademakinwa Adedeji Nelson^{1,3*}, Agunbiade Mayowa Oladele², Agboola Femi Kayode³

¹Department of Physical and Chemical Sciences, Elizade University, Ilara-Mokin, Nigeria

²Biocatalysis and Technical Biology Research Group, Institute of Biomedical and Microbial Biotechnology, Cape Peninsula University of Technology, South Africa

³Department of Biochemistry and Molecular Biology, Obafemi Awolowo University, Ile-Ife, Nigeria

ABSTRACT *Trilepisium madagascariense* fruits are carbohydrate-rich and this study directly fermented the fruit wastes into bioethanol without the need for nutrient supplementation. The total reducing sugar (TRS) present in the mesocarp and seed of *T. madagascariense* fruit wastes (*Tmfw*) was fermented to bioethanol using *Aureobasidium pullulans*. Bioethanol production by *A. pullulans* was also optimized using Box-Behnken response surface methodology (RSM). The TRS in the mesocarp and seed of *Tmfw* were 11.2 ± 0.8 and 17.1 ± 1.2 g/L, respectively and further hydrolysis with cellulase resulted in increased TRS indicating the presence of cellulose. Pre-optimization, the bioethanol yield (Y_{ps}) and volumetric productivity (Q_p) obtained from the fermentation of the seed by *A. pullulans* were 0.57 ± 0.03 g/g and 0.21 ± 0.02 g/L⁻¹h⁻¹, respectively. The optimum conditions for maximum bioethanol production were pH (5.95), time (24 h) and substrate concentration (5 g/L) resulting in Y_{ps} , Q_p of 0.66 ± 0.06 g/g and 0.27 ± 0.01 g/L⁻¹h⁻¹, respectively after model validation. *Tmfw* served as a suitable, cheap, non-toxic and readily available substrate especially in Nigeria to produce bioethanol while *A. pullulans* is a fungus that might be utilized for large-scale industrial bioethanol production.

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*Corresponding author
E-mail: adedeji.ademakinwa@elizadeuniversity.edu.ng

Introduction

Bioethanol is produced by microbial fermentation of plants containing sugar, starch and lignocellulosic materials (Joshi et al. 2012). In contrast to ethanol, bioethanol can be obtained from biomass-based waste materials or other renewable sources such as high sugar-containing plant materials (Dash 2017). It can be used as fuel, chemical feedstock and as a solvent in various industrial processes (Alma et al. 2015).

Bioethanol is referred to as a sustainable alternative energy source, which is both renewable and environmentally acceptable (Lebaka 2013). It is by far most widely used in transportation and it is oxygenated, thereby provides the potential to reduce particulate emissions in compression ignition engines (Razmovski et al. 2012). A variety of plant materials has been used for the first, second and third generation bioethanol production. The first-generation bioethanol production involves plant materials rich in sucrose (sugarcane, sugar beet, sweet sorghum, and fruits) and starch (corn, wheat, rice, potato, cassava, sweet potato and barley). The sugar-based ethanol is predominantly produced from sugarcane while starch-based ethanol is

mainly from corn but also from grains. The production of first-generation ethanol poses a low risk and does not require harsh pretreatment of the substrate. However, the utilization of edible agricultural crops solely for biofuel production conflicts with food and feed production (Sharma and Sharma 2018). The focus of this study is, therefore, to exploit the under-utilized *Trilepisium madagascariense* fruit wastes for the possible production of bioethanol since these seeds have been reported to be rich in carbohydrates (60%) (Adewuyi et al. 2010). Several reports exist that shows the plant has medicinal values as it is applied in the treatment of some ailments mostly because of its rich antioxidant/phytochemical properties (Nwamarah et al. 2015). These fruit wastes, therefore, could serve as a source of bioethanol production in certain countries (e.g., Nigeria) because it is cheap, cost-effective and readily available. The cost-effectiveness of bioethanol production is highlighted in this study since the process of production of this biofuel is free of enzymatic hydrolysis which is hitherto expensive. The economics of bioethanol production is greatly influenced by the cost of the feedstock, which accounts for more than half of the production costs. Hence, utilizing the cheap waste of the ripened fruits of *T. madagascariense* which is rich in

reducing sugars is exploitable for large-scale production of bioethanol locally in countries like Nigeria.

In this study, there was a determined effort to investigate if *Aureobasidium pullulans* can hitherto ferment reducing sugars present in the *T. madagascariense* into ethanol anaerobically. *A. pullulans* have been reported to contain enzymes that can hydrolyze sucrose to produce glucose and fructose e.g., invertases and fructosyltransferases (Ademakinwa et al. 2017). This will offer a new fungus-based method for large scale production of bioethanol in countries where *T. madagascariense* available.

Materials and Methods

Reagents

Trichoderma reesi cellulase, dinitrosalicylic acid, sodium-potassium-tartrate, sodium hydroxide, glucose, carboxymethyl cellulose, sodium acetate, ethanol, sodium dichromate dihydrate, sulphuric acid and glacial acetic acid were of analytical grade and purchased from Sigma-Aldrich (USA).

Microorganisms and culture conditions

The industrial strain of *S. cerevisiae* was obtained from the Department of Chemical Engineering, Obafemi Awolowo University, Ile-Ife, Nigeria. *A. pullulans* were previously isolated from soil containing decayed plant litters and its molecular identification was based on sequencing of the ITS1-ITS4 genomic region (Ademakinwa and Agboola 2016). Both fungi were maintained on malt extract agar (MEA) for 96 hours at 4 °C on agar slants. Preparation of the inoculum and growth medium were as described by Ademakinwa et al. (2017). All the yeast was incubated anaerobically in an anaerobic jar during ethanol production.

Preparation of *T. madagascariense* seeds and mesocarp

Ripened fruits wastes were collected from the base of the *T. madagascariense* trees located in the Botanical Garden, Obafemi Awolowo University, Ile-Ife. The fruit wastes were then washed in sterile distilled water. The mesocarps were carefully separated from the seeds using a sterile razor blade and homogenized separately in distilled water (1:2) w/v. The homogenate was then clarified by centrifugation at 4000 g for 20 min. The supernatant was stored at 4 °C prior to further use. The total reducing sugars present were quantified using the dinitrosalicylic acid method described by Miller (1959) with glucose as standard.

Enzymatic treatment of the homogenate

The supernatant obtained after centrifugation was investigated for possible increased release of more reduc-

ing sugar by hydrolysis of the cellulose present in the mesocarp and seeds. Cellulase assay was carried out according to methods described by Quadri et al. (2017) using 3,5 dinitrosalicylic acid (Miller 1959). Cellulase (0.1-10% w/v) was added to the clarified homogenate for the hydrolysis of cellulose present in the seeds and mesocarp. The reducing sugars released were quantified as described above.

Fermentation and conditions

The clarified supernatant served as the medium for fermentation and it was fermented by *A. pullulans* and *S. cerevisiae* in 100 ml Erlenmeyer flasks that contained 20 ml of the clarified supernatant in an anaerobic jar. The medium for fermentation was inoculated with 1% (v/v) fungal cultures as inoculum. The effects fermentation time on reducing sugar consumption and ethanol production were determined. After every 24 h, 2 ml was aseptically withdrawn, centrifuged at 4000 g for 10 min and the ethanol and reducing sugar present in the supernatant were quantified. The ethanol yield and productivity were calculated using the Equation 1.:

$$Y_{(p,s)}, (g/g) = \frac{\text{Ethanol Produced (g/L)}}{\text{Sugar consumed (g/L)}}$$

$$Q_p(g/L \times h) = \frac{\text{Ethanol Produced (g/L)}}{\text{Fermentation time (h)}}$$

Analytical processes

Ethanol was quantified using the dichromate method described by Hormitz (1980) as modified by Betiku et al. (2015). The reducing sugar was quantified using the dinitrosalicylic method described by Miller (1959).

Optimization of the bioethanol production processes

Response surface methodology (RSM) was used to optimize the bioethanol production process from the seed of *T. madagascariense* and to investigate the influence of different fermentation process variables on the bioethanol yield. The variables considered are pH, fermentation time (h) and substrate concentration (g/L). To evaluate the effect of initial pH, the medium for fermentation

Table 1. Process parameters for Box-Behnken response surface methodology.

Variable	Coded factor levels		
	-1	0	+1
pH	4.95	5.95	6.95
Substrate concentration (g/L)	5	7.5	10
Time (h)	24	36	48

Table 2. Experimental runs for the Box-Behnken design for ethanol optimization.

Runs	pH	Substrate (g/L)	Time (h)	Bioethanol yield (Yps) (g/g)	
				Experimental value	Predicted value
1	4.95 (-1)	2.5 (-1)	36 (0)	0.53	0.53
2	6.95 (+1)	2.5 (-1)	36 (0)	0.57	0.57
3	4.95 (-1)	7.5 (+1)	36 (0)	0.53	0.53
4	6.95 (+1)	7.5 (+1)	36 (0)	0.57	0.57
5	4.95 (-1)	5.0 (0)	24 (-1)	0.56	0.56
6	6.95 (+1)	5.0 (0)	24 (-1)	0.54	0.54
7	4.95 (-1)	5.0 (0)	48 (+1)	0.48	0.48
8	6.95 (+1)	5.0 (0)	48 (+1)	0.59	0.59
9	5.95 (0)	2.5 (-1)	24 (-1)	0.53	0.53
10	5.95 (0)	7.5 (+1)	24 (-1)	0.53	0.53
11	5.95 (0)	2.5 (-1)	48 (+1)	0.52	0.52
12	5.95 (0)	7.5 (+1)	48 (+1)	0.51	0.51
13	5.95 (0)	5.0 (0)	36 (0)	0.63	0.63
14	5.95 (0)	5.0 (0)	36 (0)	0.64	0.63
15	5.95 (0)	5.0 (0)	36 (0)	0.63	0.63
16	5.95 (0)	5.0 (0)	36 (0)	0.64	0.63
17	5.95 (0)	5.0 (0)	36 (0)	0.63	0.63

was subjected to pH 4.95, 5.95, and 6.95 using 1N NaOH and HCl, respectively. The Box-Behnken method was selected for the optimization of ethanol concentration. All variables were set at a central coded value of zero. The minimum and maximum ranges used in this optimization were selected based on, the basis of previous one-factor at-a-time independent study. The variables, factors, and levels are referenced in Table 1. Seventeen individual runs were conducted for the three independent variables (Table 2) for the quantification of ethanol. Ethanol yield was analyzed by using a second-order polynomial equation and data-fitting by multiple regression techniques using Design-Expert (version 6.0, Stat-Ease, Minneapolis, USA). The model equation for analysis is given in Equation 2.:

$$\alpha = \mu_0 + \mu_1 Z_1 + \mu_2 Z_2 + \mu_3 Z_3 + \mu_{12} Z_1 Z_2 + \mu_{13} Z_1 Z_3 + \mu_{23} Z_2 Z_3 + \mu_{11} Z_1^2 + \mu_{22} Z_2^2 + \mu_{33} Z_3^2$$

Where the predicted response (Bioethanol yield g/g) is denoted by α , μ_0 is the model constant, Z_1 , Z_2 and Z_3 are independent variables, μ_1 , μ_2 and μ_3 are linear coefficients, μ_{12} , μ_{13} and μ_{23} are cross product coefficients and μ_{11} , μ_{22} and μ_{33} are the quadratic coefficients representing the constant process effect in total. The linear (α_i), quadratic effect (α_j) and the interaction effect between α_i and α_j for the production. The experimental/predicted values and the response surface plots were compared to determine the optimum conditions for bioethanol production.

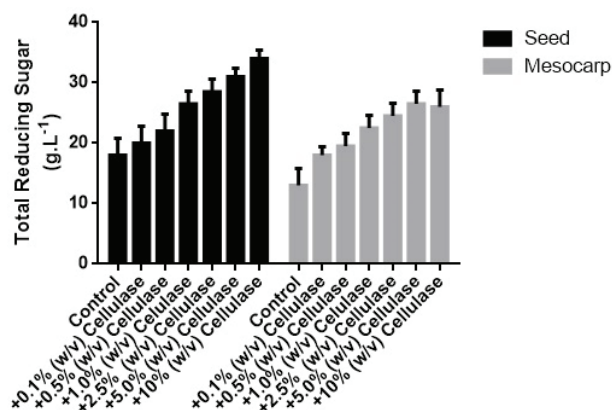


Figure 1. Total reducing sugars present in the seed and mesocarp of *T. madagascariense* fruit waste (Tmfw) with and without cellulase (0.1-10% w/v) pretreatment of the mesocarp and seed of (Tmfw).

Statistical analysis

The statistical software Design Expert 6.0.7. (Stat-Ease, Minneapolis, USA) was used for the design of experiments, regression, and graphical analysis of the data obtained and for statistical analysis of the model to evaluate the analysis of variance (ANOVA) and it was used also for the optimization of the bioethanol fermentation process.

Validation of the model under the optimized conditions

The optimum conditions obtained after the statistical optimization was used in bioethanol production to correlate the values obtained. The bioethanol yield and volumetric productivity were estimated as described in Equation 1.

Result and Discussion

Reducing sugars present in the *T. madagascariense*

T. madagascariense is known to be rich in carbohydrate (60%) Adewuyi et al. (2015). The mesocarp and seeds contained about 17.1 g/L and 11.2 g/L of the total reducing sugar without pretreatment. The addition of cellulase to the clarified homogenate increased the reducing sugar to 33.2 and 39 g/L in the mesocarp and seed, respectively (Fig. 1). This could be connected to the cellulose content of the seed and mesocarp that upon enzymatic digestion results in the release of soluble reducing sugars. There was not any significant ($p < 0.05$) increase in the total reducing sugar concentration in both seed and mesocarp of *T. madagascariense* as the enzyme concentration increased from 5.0 to 10% (w/v). This might be indicative of the complete digestion of the cellulose present in the mesocarp and seeds.

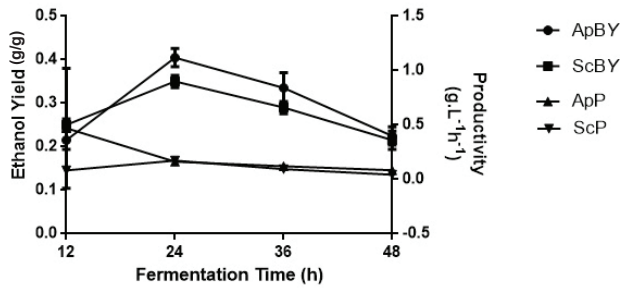


Figure 2. Bioethanol yield and volumetric productivity estimation by fermentation of the mesocarp of *T. madagascariense* fruit waste (Tmfw) using both *A. pullulans* and *S. cerevisiae*. ApBY and ScBY represent the bioethanol yield when *A. pullulans* and *S. cerevisiae* were used for the fermentation process while ApP and ScP represents the volumetric productivity when *A. pullulans* and *S. cerevisiae* were used for the fermentation of the mesocarp of (Tmfw).

Fermentation of the clarified homogenate to bioethanol

Depending on the yeast used for fermentation, the maximum ethanol yield (g/g) and % ethanol yield varied accordingly. *S. cerevisiae* and *A. pullulans* fermented the clarified supernatant of the seed and mesocarp to optimally produce ethanol after 24 h (Fig. 2). Comparatively, *A. pullulans* had higher bioethanol yield and productivity than *S. cerevisiae* and this suggests that *A. pullulans* might be exploited industrially for bioethanol production as a useful alternative to *S. cerevisiae*. The decline in bioethanol production observed after 24 h might be due to the decrease in the total reducing sugars present for the microorganisms to act upon. The findings in this present

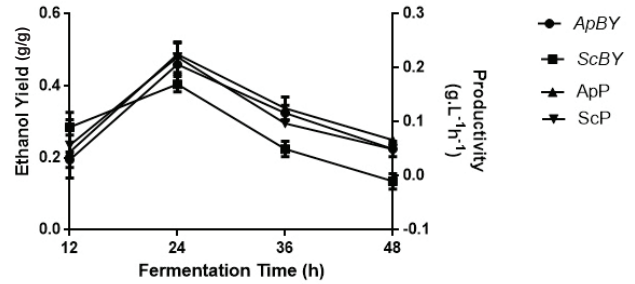


Figure 3. Bioethanol yield and volumetric productivity estimation by fermentation of the seeds of *T. madagascariense* fruit waste (Tmfw) using both *A. pullulans* and *S. cerevisiae* when *A. pullulans* and *S. cerevisiae* were used for the fermentation process while ApP and ScP represents the volumetric productivity when *A. pullulans* and *S. cerevisiae* were used for the fermentation of the seeds of (Tmfw).

study were akin to that observed by Betiku and Taiwo (2015) where the authors reported that it took 24 h for *S. cerevisiae* to ferment bread fruit hydrolysate for production of bioethanol.

RSM optimization for bioethanol production and ANOVA analysis

In the present study, seventeen independent experiments to evaluate bioethanol yield using the mesocarp of *T. madagascariense* fruit waste by considering the interactive effects of pH, incubation time and substrate concentration. The observed and predicted bioethanol yield values are shown in Table 3. The optimum condition for bioethanol production was pH of 5.95, substrate

Table 3. ANOVA for response surface quadratic model.

Source	Sum of squares (x 10 ³)	DF	F-value	Prob>F	
pH	0.32	1	335.18	< 0.0001	Significant
Substrate	0.03	1	230.49	< 0.0001	Significant
Time	0.39	1	2.33	0.1706	Not significant
pH x pH	0.39	1	28.41	0.0011	Significant
Substrate x substrate	0.011	1	280.06	< 0.0001	Significant
Time x time	0.015	1	822.73	< 0.0001	Significant
pH x substrate	0.001	1	1116.84	< 0.0001	Significant
pH x time	4.029	1	0.088	0.7755	Not significant
Substrate x time	0.11	1	291.26	< 0.0001	Significant
Model	0.0470	9	212.86	<0.0001	Significant
Residual	0.1220	7			
Lack of fit	2.48x10 ⁻⁵	3	1.92x10 ⁻⁴	1.0	Not significant
Pure error	0.172	4			
R ²	0.9964				
Adj R ²	0.9917				
Pred R ²	0.9943				
Adeq precision	41.370				
C.V.	0.8700				

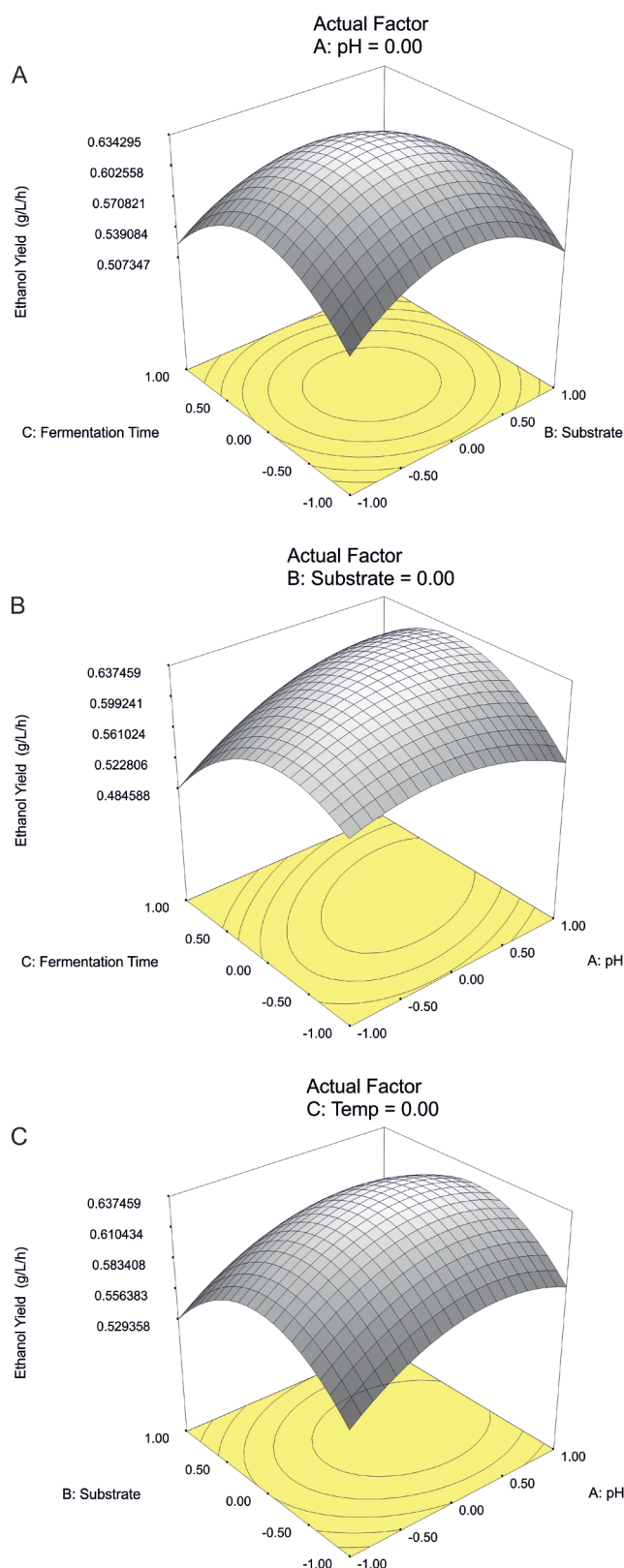


Figure 4. (A) Substrate vs pH with fermentation time held constant. (B) Fermentation time vs pH with substrate concentration held constant. (C) Fermentation time vs substrate with the pH held constant.

concentration of 5.0 g/L and incubation time of 24 h. It is reported that optimum ethanol production occurs at a pH range of 4-6 and ethanol production is influenced by pH of the broth as it affects bacterial contamination, yeast growth, fermentation rate, and byproduct formation. From the ANOVA of the quadratic model, it was noted that model terms such as pH, time, pH x pH, substrate concentration x substrate concentration, time x time and pH x time were significant ($p < 0.05$). The model F-value of 212.9 implied that the model was significant. The F-value is often used as a measure of how the factors aptly describe the variation in the data set. The F-values obtained for the data set in this study indicates that the model is significant when also considering the P-value (< 0.0001). The quadratic model was used in the theoretical prediction of the bioethanol yield reliably due to the R^2 value (Table 3). It is reported that the R^2 values must fall between 0.75 and 0.80 for a good model fit. The values obtained in this study allows for the predictability of the bioethanol yield from the quadratic equation in coded factors as shown below (Equation 3):

$$\begin{aligned} \text{Ethanol Yield} = & +0.64 + 0.020 \times A - 2.008\text{E-}003 \times B - \\ & 7.009\text{E-}003 \times C - 0.034 \times A^2 - 0.056 \times B^2 \\ & - 0.064 \times C^2 + 5.512\text{E-}004 \times A \times B + 0.032 \\ & \times A \times C - 5.276\text{E-}003 \times B \times C \end{aligned}$$

where A, B, and C represent pH, substrate concentration and time of fermentation, respectively.

The "Pred R^2 " of 0.9943 is in reasonable agreement with the "Adj R^2 of 0.9917." Adeq Precision" measures the signal to noise ratio. A ratio greater than 4 is desirable. The ratio of 41.371 obtained in this process indicates an adequate signal. The summary of the ANOVA analysis is shown in Table 4.

Interaction between factors

From the response surface plots (Fig. 3-5), it was deduced that there were moderate interactions between the variables considered in this study. To obtain these response surface curves, the interactions between two variables were investigated by obtaining 3D response surface plots while the third variable was kept constant. The optimum pH obtained was mildly acidic and increasing the pH resulted in a decrease in the bioethanol yield. The pH plays a crucial role in fermentation as it is directly tied to most biological processes (Manohar and Divakar 2005). Also, decreasing or increasing the fermentation period above 24 h resulted in a decline the bioethanol production.

Conclusion

A readily available, cheap and non-toxic feedstock of *T. madagascariense* fruit wastes investigated in this study provides a novel source for bioethanol production. The reducing sugar present with and without enzymatic saccharification indicates that these agricultural wastes could be converted inexpensively to bioethanol. Validation of the model using the optimum conditions predicted after statistical optimization indicated that the volumetric productivity and bioethanol yield obtained was 0.24 g/L/h and 0.63 g/g, respectively.

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