

# ANALYSIS OF FACTORS AFFECTING PRODUCTION RICE IN LANGKAT REGENCY WITH METHODS BACKWARD IN MULTIPLE LINEAR REGRESSION YEAR 2018

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## Article Info

### Article history:

Received October 1, 2019

Revised November 2, 2019

Accepted December 2, 2019

### Keywords:

lowland rice production, method backward

## ABSTRACT

Rice is a food crop that is needed by the people of Indonesia, because the rice plant as a producer of rice that can be processed as a staple food for the population in Indonesia. Factors considered to have an effect on lowland rice production in Langkat Regency are: yearend crop residues, land area, rainy day rainfall, seeds, and fertilizers. In determining the factors that greatly influence the production of lowland rice, it is necessary to use multiple linear regression equations such as using the method backwards. That is, where all X variables are regressed with Y variables.

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## 1. INTRODUCTION

Rice is a food crop as a staple food for Indonesia, the need for rice will always increase with the increase in the population of Indonesia, with the increase in Indonesia's population, the supply of raw materials must also be increased, especially rice as a staple in Indonesia. (May, 2015)

In order to produce a high enough production, it is necessary to do research on the factors that affect rice production such as rainfall, rainy days, seeds, land area, fertilizer, final crop residue.

In multiple linear regression discusses the relationship pattern of several variables in the model, how the direct influence of the independent variable (independent) on the dependent variable. In this study, it is analyzed how big the influence of land area, rainfall (mm), seedlings, rainy days, fertilizer use (kg), and year-end crop residues.

## 2. RESEARCH METHODE

### 2.1 Regression analysis

Regression analysis is a statistical analysis tool that utilizes the relationship between two or more variables. The goal is to make a reliable estimate (prediction) to make called the dependent variable or dependent variable or (response), if the value of

another variable whose relationship is known (commonly called the independent variable or independent variable or predictor).

### 2.2 Multiple linear regression

Multiple linear regression is a linear regression consisting of one variable bound (dependent) and more than one independent variable  $X_1, X_2, \dots, X_k$  (independent). Meaning of linear is linear in its parameters and variables, which means that each parameter is only to the power of one and is not multiplied or divided by other parameters. The form of the multiple linear regression equation is as follows: (Gujarati, 2004).

$$Y_i = \theta_0 + \theta_1 X_{i1} + \theta_2 X_{i2} + \dots + \theta_k X_{ik} + \epsilon_i; i = 1, 2, \dots, n$$

### 2.3 Linear Regression Model in Matrix

There are several models of linear regression equations in the matrix. (Kuniawan, 2016)

(2.1)

$$Y_1 = \theta_0 + \theta_1 X_{11} + \theta_2 X_{21} + \dots + \theta_p X_{p1} + \epsilon_1 \quad Y_2 = \theta_0 + \theta_2 X_{12} + \theta_2 X_{22} + \dots + \theta_p X_{p2} + \epsilon_2$$

$$Y_3 = \theta_0 + \theta_3 X_{13} + \theta_3 X_{23} + \dots + \theta_p X_{p3} + \epsilon_3$$

Linear regression model with dependent variables  $Y$  and  $p$  independent variable  $X_1, X_2, \dots, X_p$ , can generally be written as follows:

$$Y_n = \theta_0 + \theta_1 X_{1n} + \theta_2 X_{2n} + \dots + \theta_p X_{pn} + \epsilon_n$$

### 2.4 Method Backward This method is the opposite of the advanced

This method is the opposite of the advanced selection method. First of all, all independent variables are included in the model. Suppose there are 4 independent variables ( $X_1, X_2, X_3, X_4$ ), then the initial model can be

$$\text{written: } Y = \theta_0 + \theta_1 X_1 + \theta_2 X_2 + \theta_3 X_3 + \theta_4 X_4$$

### 2.5 Parameter Estimation of Multiple Linear Regression Model

If you have a random sample of size  $n$ , that is  $(Y_i, X_{1i}, X_{2i}, \dots, X_{pi})$  where  $i = 1, 2, \dots, n$  from a population, then the regression model can be written: (Ria, 2013)

$$Y_i = \theta_0 + \theta_1 X_{1i} + \theta_2 X_{2i} + \dots + \theta_p X_{pi} + U_i$$

$$U_i = Y_i - (\theta_0 + \theta_1 X_{1i} + \theta_2 X_{2i} + \dots + \theta_p X_{pi})$$

### 2.6 Determination Correlation Coefficient (Determination Index)

The correlation coefficient of determination (Determination index) is expressed by  $R^2$ . This coefficient states the large proportion/contribution of  $X_1, X_2, X_3, \dots, X_k$  together against variation or fluctuation of  $Y$ . Price  $R^2$  obtained by the following formula (Ria, 2013)

$$R^2 = JKR/JKT \text{ with } 0 \leq R^2 \leq 1$$

## 2.7 Research procedure

The research procedures applied in this study in order to achieve the research objectives are as follows:

- Swallowing references that support research
- take data to the Langkat district office
- estimate multiple linear regression coefficients using the OLS . method
- Choosing the best model with the backward method metode
- Interesting conclusion.

## 3. RESULT AND ANALYSIS

### 3.1 Multiple Linear Regression Model with Matrix Approach

The value of the regression coefficient is calculated using a matrix.

From the above results obtained the value (( 0 , 1 , 2 , 3 , 4 , 5 , 6), then it can form

The general formula for multiple linear regression is as follows:

$$Y=2710109.994+5970,712536 X 1 -50556.0824 X 2+783,8277963 X 3 -26346.8268 X 4 +351.1459899 X 5+244.198574 X 6$$

### 3.2 Multiple Regression Equation between Y and X 1, X 2 , X 3 X 4 , X 5 X 6

Multiple Regression Coefficient Between Y and X 1, X 2 , X 3 X 4 ,X 5 ,X 6 looking for similarities multiple linear regression with SPSS.

Coefficients <sup>a</sup>					
Model	Unstandardized Coefficients		Standardized Coefficients	t	Sig.
	B	Std. Error	Beta		
(Constant)	2710109.944	7724606.719		.351	.731
Leftovers end of year	5870.713	3399.386	.256	1.727	.106
Land area	-50556.082	10779.144	-2.853	-4.690	.000
Rainfall	783.828	1202.047	.025	.652	.525
Rainy day	-26346.827	44660.232	-.028	-.590	.565
Seeds	351.146	725.827	.404	.484	.636
Fertilizer	244.199	97.474	3.164	2.505	.025

a. *Dependent Variables: Rice paddy production*

From result output above can be obtained  $b_0 = 2710109,944$ ,  $b_1 = 5870.713$ ,  $b_2 = -50556,082$ ,  $b_3 = 783,828$ ,  $b_4 = -26346.827$ ,  $b_5 = 351.146$ ,  $b_6 = 244.199$

From the results above, it can be in the form of a multiple linear regression formula as follows:  $Y = 2710109.994+5870.70 X 1 -50556.082 X 2+783,828 X 3 -26346.827 X 4+351.146 X 5+244.199 X 6$

### 3.3 Multiple Regression Significance Test

ANOVA results between Y and X 1, X 2 , X 3 X 4 , X 5 ,X 6

**ANOVA<sup>a</sup>**

Model	Sum of Squares	df	Mean Square	F	Sig.
Regression	1182998288770 0130.000	6	1971663814616688.20 0	135.303	.000 <sup>b</sup>
Residual	2040103332522 54.660	14	14572166660875.332		
Total	1203399322095 2384.000	20			

a. Dependent Variables: Rice paddy production b. Predictors: (Constant), Fertilizer, Rainy day, Rainfall, Year

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Based on the table output The SPSS above is known for the value of sig. is 0.000. Because value sig.  $0.000 < 0.005$ , then the hypothesis is accepted. known value F count = 135,303 while the level value of real 0.05 obtained the value of F table =  $F(6,14) = 2.85$  because F count  $>$  F table then it can be concluded that coefficient means. To find out the coefficients that come out of the model, we can see from partial F values below.

**ANOVA<sup>a</sup>**

Model	Sum of Squares	df	Mean Square	F	Sig.
Regression	109774311170132 64.000	1	10977431117013264 .000	197.406	.000 <sup>b</sup>
Residual	105656210393912 0.900	19	55608531786269.52 0		
Total	120339932209523 84.000	20			

a. *Dependent Variable:* Hasil produksipadi sawah

a. *Predictors: (Constant), Sisa tanam akhir tahun*

From result output above can be seen the value of F Partial from X 6 Fertilizer = 277,868 with sig. 0.000.

**ANOVA<sup>a</sup>**

Model	Sum of Squares	df	Mean Square	F	Sig.
Regression	1089628221893 9466.000	1	10896282218939466 .000	181.970	.000 <sup>b</sup>

Residual	1137711002012 918.200	19	59879526421732.54 0		
Total	1203399322095 2384.000	20			

. Dependent Variables: Rice paddy production Total 20

b. Predictors: (Constant), Land are

From result output above can be seen the value of F Partial from X 2 Land area = 181.97 with sig. 0.000 .

ANOVA\*

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	1585803292553 370.000	1	1585803292553370. 000	2.884	.106 <sup>b</sup>
Residual	1044818992839 9014.000	19	549904733073632.3 00		
Total	1203399322095 2384.000	20			

a. Dependent Variables: Rice paddy production

b. Predictors: (Constant), Rainfall

From result output t above can be seen the value of F Partial from X Rainfall = 2.884 with3

ANOVAa

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	8391593687184 11.900	1	839159368718411.9 00	1.424	.247 <sup>b</sup>
Residual	1119483385223 3972.000	19	589201781696524.9 00		
Total	1203399322095 2384.000	20			

a. Dependent Variables: Rice paddy production Total 20

b. Predictors: (Constant), Rainy day

From result output above can be seen the value of F Partial from X 4 Rainy days = 1,424 with sig. 0.247.

ANOVA\*

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	1139454899470 7588.000	1	11394548994707588 .000	338.570	.000 <sup>b</sup>
Residual	6394442262447 96.100	19	33654959276041.90 2		
Total	1203399322095 2384.000	20			

a. Dependent Variables: Rice paddy production

b. Predictors: (Constant), Seeds

From result output above can be seen the value of F Partial from X 5 Seeds = 338,570 with sig. 0.000.

**ANOVA<sup>a</sup>**

Model	Sum of Squares	df	Mean Square	F	Sig.
1 Regression	1126379899909 5396.000	1	11263798999095396 .000	277.868	.000 <sup>b</sup>
1 Residual	7701942218569 87.800	19	40536537992473.04 0		
Total	1203399322095 2384.000	20			

a. Dependent Variable: Lowland rice production

b. Predictors: (Constant), Fertilizer

From result output above can be seen the value of F Partial from X 6 Fertilizer = 277,868 with sig. 0.000.

From the distribution list F obtained the value of F Partial smallest =  $1.424 < F(6,14) = 2.85$  on variable

X 4 i.e. a rainy day, because the value of F Partial is in the variable X 4, then variable X 4 out of model. d. Determination Correlation Coefficient (Determination Index) The correlation coefficient of determination (the index of determination is expressed by R<sup>2</sup> to find out the value of R then see the output below:

3.4 Determination Correlation Coefficient (Determination Index)

**Model Summary**

Model	R	R Square	Adjusted R Square	Std. Error of the Estimate
1	.991 <sup>a</sup>	.983	.976	3817350.73852

From result output above it can be seen that the value of R = 0.991 then the value of R<sup>2</sup> =  $0.991 \times 0.991 = 0.983$  is called R squares. To find out the percentage of the estimator, the value of R<sup>2</sup> x 100%.  $0.983 \times 100\% = 98.3\%$

This means that the value of the Y variation that can be explained by the method backward is 98.3% while the remaining 1.7% is influenced by other variables that are outside the model equation.

## CONCLUSION

Of the six factors studied as factors that affect the production of lowland rice, then There are 4 predictive factors that greatly influence the year-end crop residue, land area, seeds, and fertilizers.

The estimating equation formed from the method backward is

$$Y = 788623,070 - 4979,786 X_1 - 51051,961 X_2 + 182,936 X_3 + 264.100 X_4$$

$X_1$  = End of year cropping

$X_2$  = Land area

$X_3$  = Seeds

$X_4$  = Fertilizer

based on the discussion of estimators, the results obtained from the percentage of variation described by the backward is 98.3% while the remaining 1.7% is influenced by other variables that are outside the model equation.

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