

## Research Article

# Village Funds Policy and its Impact on Improvement and Autonomy Status of Villages in Indonesia

Rosita Novi Andari<sup>1\*</sup> Rizky Fitria<sup>2</sup>

<sup>1</sup>Research Center of Community Welfare, Village and Connection, National Institute of Research and Innovation Agency

<sup>2</sup>National Institute of Public Administration

### ORCID

Rosita Novi Andari: <https://orcid.org/0000-0002-1244-6320>

Rizky Fitria: <https://orcid.org/0000-0002-4484-6564>

### Abstract.

"End poverty in all its forms everywhere" is the ultimate goal and crucial element of the world's SDGs 2030 transformative agendas, as well as in Indonesia. The COVID-19 pandemic has hindered Indonesia from achieving its SDGs targets and reducing poverty, including rural poverty. The village funds policy sourced from the state budget is one of the policies that aims to alleviate rural poverty; thus, there will be a reduction of disadvantageous villages and an enhancement of self-sustained villages in Indonesia. This study aims to investigate the impact of village funds on the improvement and autonomy status of villages in Indonesia. It utilizes a quantitative method, namely linear regression of panel data using the Village Building Index (VBI) as the dependent variable and village funds as the independent variable over 2018-2021 in each regency/city in Indonesia. The result suggests that the village fund policy, through the 10% allocation of the state budget, has a positive and significant effect on changes in the improvement and autonomy status of the villages in Indonesia. The finding using the model (lin-log) tells us that a 1% increase in village funds will raise the VBI by 0.0013. This indicates that the village funds policy has contributed to the achievement of rural poverty alleviation goals, or it could be a pro-poor policy in order to achieve the purpose of the 1<sup>st</sup> SDGs goal.

**Keywords:** SDGs, poverty, village funds policy, improvement and autonomy status of villages

## 1. Introduction

Sustainable development is a transformation process to ensure the fulfillment of the needs of the present generation without reducing the development opportunities of future generations through integrated actions in the economic, social and environmental fields (1). The successful implementation of the Sustainable Development Goals (SDGs) agenda is the only way forward to address global sustainability challenges in order to ensure human well-being, economic prosperity, and environmental protection(2).

Corresponding Author: Rosita  
 Novi Andari; email:  
[rositanovi@gmail.com](mailto:rositanovi@gmail.com),  
[fitria.rizky@gmail.com](mailto:fitria.rizky@gmail.com)

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Ending poverty in all forms (end poverty in all its forms everywhere) is the first goal and an important element of the transformative agenda of the 2030 SDGs (3) and the foundation for achieving other SDGs goals (4). Poverty Alleviation (SDGs1) has the most synergistic relationships with other SDGs objectives (2,5,6). The impoverishment has negative effects, namely being a threat to the implementation of almost all of the SDGs and keeping the poverty gap between industrialized and developing countries remains (3). Covid-19 also has implications in the achievement of the SDGs with widening the poverty (4) and hindering developing countries to overcome poverty because they have fragile economic systems and no financial support (3).

There are still few studies that tell about the topic of achieving the 1<sup>st</sup> SDGs since the topic has not become the main focus yet (7). Several previous studies have suggested future research on intergenerational poverty; urban poverty and poverty alleviation design with heterogeneous strategies (8); poverty reduction that can pose major obstacles to the implementation of the SDGs due to the presence of unexpected pandemics such as COVID-19 and a series of important items to encourage the implementation of one of the key SDGs (3).

In order to achieve the 1<sup>st</sup> SDGs goal, impoverishment in rural areas becomes a concern around the world, especially in developing countries. Urban and rural are organisms, thus urban poverty should give an equal attention to rural poverty, particularly in developing countries (9). Previous research mentioned that in the future it is necessary to conduct studies on the evaluation of the effectiveness of poverty reduction around the world and in various countries; studies that illustrate the gap between poverty alleviation and sustainable development goals, studies that analyze constraints that affect the achievement of poverty alleviation goals, and studies of models of poverty reduction strategies globally that are suitable for different regions, including in rural areas because eradicating poverty is the basis for revitalizing rural areas around the world (9).

Indonesia is one of the developing countries that still experiences obstacles in achieving the 1st SDGs goal, especially poverty in rural areas. Multidimensional poverty in Indonesia is clearly higher in rural areas than it in urban ones, although the gap in both is narrowing(10). The Covid-19 pandemic crisis has also prevented Indonesia from achieving the SDGs target and reducing the poverty rate (11) including rural poverty.

The condition of poverty in Indonesia shows that there is a gap in rural poverty which tends to be higher than urban areas. Table 2 tells that the percentage of population number of poverty, Poverty Gap Index and Poverty Severity Index in urban areas have decreased from 2018 to 2019. In 2020 (the Covid Pandemic period) it got the higher

number and then diminished again in 2021. However, the percentage of population number of poverty, Poverty Gap Index and Poverty Severity Index in rural areas tend to be higher when compared to urban ones.

TABLE 1: Rural and Urban Poverty in Indonesia.

Poverty Indicator	Year			
	2018	2019	2020	2021
1. % population number in poverty				
Rural	13.1	12.6	13,2	12,53
Urban	6.89	6.56	7,88	7,6
2. Poverty Gap Index				
Rural	2.32	2.11	2,39	2,25
Urban	1.08	1.02	1,26	1,23
3. Poverty Severity Index				
Rural	0.62	0.53	0,68	0,59
Urban	0.25	0.23	0,31	0,29

Source: (12,13)

The policy of village income sourced from the allocation of the State Budget (village funds) is one of the macro policies set by the government to overcome poverty in rural areas. The policy regulated by Law No.6 of 2014 concerning the Villages is one of the important changes in village policy during the Reformation period because it is different from the regulation of village income sources in the previous Law, namely Law No.22 of 1999 and Law No.32 of 2004, therefore starting in 2014 the central government directly guarantees the legality of financing government administration, development and community empowerment and village community through regulations on the amount and method of allocating, distributing and using village income sourced from the state budget, as much as 10% for the villages(14). The village funds is a form of state recognition to the village(15).

One of the objectives of the village funds policy is to alleviate poverty(15). In its implementation, several studies state that the village funds policy has an influence on the progress of villages in various regions in Indonesia. Research by Sigit & Kosasih(16) stated that village funds and village funds allocations negatively affected poverty at the district/city level in Indonesia in 2015-2017. The one conducted by Jumiati & Adam (17) stated that partially, the village funds program affects economic growth, expansion of employment and business opportunities and simultaneously there is an influence of the village funds program on economic growth, increase of employment and business opportunities in 10 villages in Purwakarta Regency. Arfiansyah (18) stated that the village

funds negatively affects poverty in Central Java Province, where the increasing village funds decreases the poverty rate. Study by Faoziyah & Salim (19) stated that 25.35% of areas that experienced proliferation (expansion) received a significant increase in village funds, but was not enough to reduce the poverty rate. Another study carried on by Gusti et al.,(20) stated that there was no significant relationship between village funds and the decline in poor heads of families or an increase in the number of village funds was not accompanied by a decrease in the poverty rate in Pesisir Selatan Regency. From some of those studies, it can be seen that the village funds policy affects poverty, economic growth, job expansion and business opportunities.

Although the village funds policy is a pro-poor policy, it is still unable to alleviate rural poverty evenly and still causes various problems. Research by Hastono & Shah(21) states that the village funds is a fund used for village welfare, even so, the current management of the village funds is considered too convoluted and full of uncertainty so that it can give rise to conflicts of interest and corrupt practices. Ernawati et al. (22) mentions that the village funds encourages inclusive growth as a pro-poor and pro-job policy but not pro-equality. As a pro-poor policy, the funds are allocated in accordance with the economic development needs of rural communities and encouraging employment opportunities in rural areas. In addition, the expansion of leading sectors and their supporting industries, such as agriculture and community empowerment, drives economic activity and creates new job opportunities. Rural infrastructure development also plays a role in creating new jobs through cash-for-work. However, village development policies are uneven, indicating that the programs financed from the allocation of these funds have not reduced the income gap of the community.

Based on the description of the problem and previous research, it can be seen that poverty alleviation in rural areas of Indonesia through the implementation of the village funds policy is one of the government's strategic steps to achieve the 1st SDGs goal. The Village Building Index (VBI) is an indicator to measure the improvement and autonomy status of villages to attain one of the village development goals, namely tackling poverty (Village Ministerial Regulation Number 2 of 2016 concerning the Building Village Index). Several previous studies have examined the effect of village funds policies on VBI in several regions of Indonesia. Research by Yulitasari & Tyas (23) stated that the change in the magnitude of village funds did not have a significant effect on changes in the status of villages in Central Java Province. Arina et al (24) stated that the village funds has a significant effect on VBI in Southeast Minahasa Regency. Research conducted by Dewi (25) stated that the village funds had a positive effect on VBI in Klaten Regency. One by Kharisma et al. (26) mentioned that the increase in the village funds budget and the

geographical difficulty index had a negative impact on the VBI in Riau Province. Study by Iftitah & Wibowo (27) stated that the use of Village Funds for capital participation of Village-Owned Enterprises and Village Original Income has a positive influence on VBI in Gowa Regency. Alhaqi (27) proposed that the use of village funds has a strong impact on the development of village independence in Hanging District. Study by Adekayanti & Achyani (29) stated that the village funds has a positive effect on VBI and has a negative influence on the poorness.

Even though previous studies have examined the effect of the village funds policy on VBI, the research is still partial in certain periods and regions in Indonesia. There has been no research that scrutinizes the influence of the village funds policy on VBI in Indonesia thoroughly. Therefore, this study aims to investigate how the village funds policy affects VBI in every Regency and City in Indonesia from 2018-2021. This research contributes on providing an empirical picture of the improvement and autonomy status of the villages in Indonesia in an effort to achieve of the 1<sup>st</sup> SDGs goal, such as tackling poverty through the village funds policy and conceptually for the development of further research on rural poverty in Indonesia.

## 2. THEORETICAL STUDY

Poverty is a diverse, dynamic and multidimensional phenomenon, it is not a static and singular phenomenon (29–31). As a dynamic phenomenon, poverty manifests in the changes in well-being and socioeconomic status that individuals exhibit over time (30). Meanwhile, as a multidimensional phenomenon, poverty includes economic indicators consisting of income per capita; income poverty line; and income inequality and non-economic indicators consisting of education; health and nutrition; and environment (32). In addition, poverty also involves many aspects such as the geographical, socioeconomic, system and cultural environment (9) that occurs both at the economic and social levels and can be caused by behavioral, structural and political factors (33).

There are several factors that affect poverty in rural areas. Rural poverty factors at the macro level identify the causes of poverty in a country or region changing over time while poverty factors at the micro level identify the causes of some households in villages being poorer than others (34). Important factors that cause poverty at the village level include the type of terrain, the area of agricultural land per capita, the ratio of access to safe drinking water, the ratio of the labor force, and the ratio of rural residents enrolled in pension insurance in each Village. Meanwhile, the key factors that cause poverty at the district level contain per capita income, rough participation rates

in the first three years, the ratio of poor villages to passenger buses, vegetation cover, and terrain relief. Differences in the impact of the frequency of natural disasters and the area of cultivated land per capita between districts are influenced by factors at the district level. Meanwhile, individual and group effects have a significant impact on the incidence of poverty (35). The main factors influencing rural poverty in this case are the number of minors, the number of migrant workers, the number of peasant farmers and the proportion of wage equivalent income have significant effectiveness against rural poverty, while the status of the head of household, health status and accessibility traffic have little influence (36).

Furthermore, there is Multidimensional Poverty Index (MPI) that aim to measure acute poverty in more than 100 developing countries. This index consists of 3 dimensions and 10 indicators, for instance (1) the dimension of health (nutrition and child mortality); (2) educational dimensions (school year and school attendance); and (3) standard of living (fuel for cooking, sanitation, drinking water, electrical energy, homes, and asset ownership). In the global MPI, people are considered multidimensionally poor if they are deficient in one-third or more than 10 indicators with each indicator having the same weight in each dimension (37). The Multidimensional Poverty Line (MPL) can be used to estimate the incidence of poverty and assess the success or failure of the implementation of poverty strategies, policies and programs in Indonesia. The measurement consists of 3 variables, namely the adequacy dimension capability (10 indicators); empowerment variables dimensions of physical limitations, public services, gender equality and legal equality (4 indicators) and opportunities dimensions of access to loan, employment opportunities, access to business and training, access to roads, access to electricity and energy, market access, education, health, water and sanitation (21 indicators) (38).

VBI is one of the measures to see how far villages in Indonesia are able to overcome poverty. According to the Regulation of the Minister of Villages Number 2 of 2016, VBI is the basis for determining the classification of village status, that are: (a) Self-sufficient Village is a village that has the ability to carry out village development to improve the quality of life and life as much as possible for the welfare of rural communities with social, economic, and ecological resilience in a sustainable manner ( $VBI > 0.8155$ ); (b) Advanced Village is a village that has the potential for social, economic and ecological resources, as well as the ability to manage them to improve the welfare of rural communities, the quality of human life and overcome poverty ( $VBI \leq 0.8155$  and  $> 0.7072$ ); (c) Developing Village is a potential village to become a Developed Village, which has the potential for social, economic and ecological resources but has not managed them optimally

for improving the welfare of rural communities, the quality of human life and tackling poverty ( $VBI \leq 0.7072$  and  $> 0.5989$ ); (d) Disadvantaged Villages are villages that have the potential for social, economic and ecological resources but have not managed them in an effort to improve the welfare of rural communities, the quality of human life and overcome poverty in its various forms ( $VBI \leq 0.5989$  and  $> 0.4907$ ); and (5) Very Disadvantaged Village is a village that experiences problems due to natural disasters, economic shocks, and social conflicts so that it is not able to manage the potential of social, economic and ecological resources and experiences poverty in its various forms ( $VBI \leq 0.4907$ ). The determination of in the improvement and autonomy status of the villages is an instrument between the Central Government, Regional Governments and Village Governments in carrying out the development and empowerment of village communities and specifically for the needs of mapping the village typologies and preparing priorities for the use of village funds.

VBI covers three dimensions, which are (1) the social dimension which includes the sub-dimensions of education, health, social capital and settlements; (2) economic dimension includes economic diversity of community production, availability and access to loan and banking, transportation (infrastructure and modes of transportation), access to trade centers (markets) and services; and (3) ecological dimensions related to environmental quality with water, soil and air quality components and awareness of disaster risks. In other words, VBI is a composite index produced from the average ecological resilience index (IKL), economic resilience index (IKE) and social resilience index (IKS) of each village.

### 3. METHods

This study utilizes VBI as dependent variable and village funds as the independent one. Independent variables were taken in the period 2018-2021, while the VBI variables used were from the period 2019-2022; this is due to the update of the VBI value in 2018 (t) published in 2019 (t+1). Both are secondary data in each Regency and City throughout Indonesia obtained from the Ministry of Finance (40–44) and (39–43) (45–49)(44–48). Based on the Decree of the Minister of Home Affairs No. 050-145 of 2022 concerning the Provision and Updating of Codes, Data on Government Administration Areas, and Islands in 2021, Indonesia has 416 Regencies and 98 Cities. However, not all of these areas receive the village funds; moreover, there is a data void in both the village funds and the VBI variable in the Regency/City for the certain years. Hence the data used in this study involved 404 regencies/cities in the 2018-202 and 2019-2022 time spans.

This study uses a quantitative method in the form of regression of panel data. Panel data is cross-section data that is viewed in a certain time frame. There are some of the advantages of using panel data analysis, for example it can explicitly take into account the heterogeneity present in each unit; providing more informative data, more variability, reducing collinearity between variables, and giving higher degrees of freedom and efficiency; providing a better picture of the dynamics of change; detecting and measuring unobserved influences on time-series and cross-section data; explaining complex behavioral models and minimizing biases that can arise from aggregated data.

Based on previous research, the hypothesis proposed in this study is that the village funds policy has a positive and significant effect on VBI. Testing of the hypothesis is carried out using a panel data regression model in the form of the following equation:

$$VBI_{it} = \alpha_0 + \alpha_1 \ln DD_{it} + e_{it}$$

**Equation 1** Panel Data Regression Model.

VBI is the i-th District/City Building Village Index at the t-th time, DD is the i-th District/City village funds at the t-th time,  $\alpha_0$  is constant (*intercept*),  $\alpha_1$  is the variable coefficient of the village funds, and  $e_{it}$  is an error term. In the model above, the village funds variable is transformed into a natural logarithm (ln) for reasons of interpretation. The change in ln data is an approximation of the relative change (in percent) so that the interpretation of the impact of the change of explanatory variable on dependent variables becomes easier to understand.

In the regression analysis of the data panel, there are three test models that can be carried out, including (1) the Common Effect Model (CEM) which is often called Pooled Regression; (2) Fixed Effect Model (FEM); and (3) Random Effect Model (REM). CEM is a panel data regression model that combines time series and cross section data assuming that there is no individual specific influence, so estimation can be done with the Ordinary Least Square (OLS) model. Meanwhile FEM is a model that pays attention to the diversity of independent variables according to individuals (49). Furthermore, REM is a model of panel data estimation where error terms can be related between times and individuals.

Before determining the best model among them to estimate the effect of the village funds on VBI, it is necessary to carry out several stages of testing such as Figure ??.

**Figure ??** Stages of Data Panel Regression Testing.

The Chow test or F-statistical test is useful for determining whether the better model to use is CEM or FEM. This test is a fixed effect significance test to decide whether the model assumes of a fixed slope and intercept between individuals and between times



(common effect), or whether it is necessary to add a dummy variable to determine the difference in intercept (fixed effect). Furthermore, the Hausman Test is a test that is carried out to select the use of FEM or REM. This test follows the chi square statistical distribution with a degree of freedom of  $k$  where  $k$  is the number of independent variables.

After obtaining the best model in estimation, it needs to carry out the classical assumption test to see if the model had fulfill the assumptions thus it gives the efficient result. This test will also determine the best weighting on the pre-selected panel data regression model. Additionally, t-statistical examination was performed to determine whether the explanatory variable affect the dependent one significantly. This test is a way to prove that the regression parameters in a model are statistically significant or not. The whole process of collecting and analyzing the data is using the help of the EViews 12.

#### 4. RESULT AND DISCUSSION

The alleviation of rural poverty in Indonesia through the implementation of the village funds policy is one of the government’s strategic steps to achieve the 1st SDGs goal. In accordance with the data used in this study, there were 404 regencies/cities that fully receive the village funds during the 2018-2022 period. The average village funds per regency/city in 2018 was 0.14 TrillionRp or 140 BillionRp, and then increased to 0.16 TrillionRp in 2019. In 2020, on average, regency/city in Indonesia received Rp. 0.17 trillionRp of village funds, and returned to 0.16 trillionRp in 2021 and 2022 as shown in Table 2.

TABLE 2: Village Funds in 2018-2020.

Year	Village Funds (in trillions of rupiah)	Average Village Funds per District/City (in trillion rupiah)
2018	55,60	0,14
2019	64,61	0.16
2020	67,21	0.17
2021	66,42	0.16
2022	63,84	0.16

Source: Authors’ calculation

Furthermore, the ability of villages in Indonesia to overcome poverty can be seen according to data on in the improvement and autonomy status of the villages in regencies/cities according to VBI. Table 3 depicts that there is a change in the progress and autonomy status of villages in the Regency/City every year. For example, in 2018

there are no Regencies/Cities that have the status of Self-sufficient Villages, meanwhile in 2022 there have been those that have obtained this status, including Denpasar City, Batu City, Badung Regency, Banjar City, Full River City, Banyuwangi Regency, Bantul Regency, Tabanan Regency, Klungkung Regency, Mempawah Regency, Gianyar Regency, Jembrana Regency, Banda Aceh City, Sleman Regency, Sambas Regency, Karangasem Regency, West Kotawaringin Regency and Ambon City. This means that those regions already have the ability to carry out village development to improve the quality of life and life as much as possible for the welfare of rural communities with social, economic, and ecological resilience in a sustainable manner.

TABLE 3: Improvement and Autonomy Status of Village in Regencies/Cities According to VBI.

Year	Number of District/ City	Number of District/City Status According to VBI				
		Self-sufficient	Advance	Developing	Disadvantage	Very Disadvantage
2019	404	2	28	258	107	9
2020	404	6	54	270	68	6
2021	404	10	94	244	51	5
2022	404	18	151	196	36	3

Source: Authors' calculation

In the meantime, regencies/cities that have the status of disadvantaged villages have decreased from 107 in 2019 to 36 in 2022. This means that they have the potential for social, economic and ecological resources but has not or lacks management in an effort to improve the welfare of rural communities, the quality of human life and overcome poverty in its various forms. Besides, regencies/cities that have very disadvantaged village status have decreased from 9 in 2019 to 3 in 2022, namely West Nias Regency, Arfak Mountains Regency, and Tambrauw Regency. It tells that the district is experiencing problems with natural disasters, economic shocks, and social conflicts so that it is not able to manage the potential of social, economic and ecological resources and experiences poverty in its various forms.

To examine the influence of the village funds policy on the in the improvement and autonomy status of the villages in Indonesia in an effort to achieve the 1<sup>st</sup> SDGs goal of tackling poverty, a panel data regression analysis was carried out by testing the effect of the amount of village funds allocation on VBI in 404 Regencies and Cities in Indonesia in the 2018-2021 period. The descriptive statistical analysis of the variables used in this study presented at Table 4.

The table shows that all the values of variabels used int the paper are positive, and the increase in village funds is argued to produce an increase in VBI.

TABLE 4: Descriptive Statistical Analysis.

	VBI	Village Funds
Mean	0,660724	1.57e+08
Std. Error	0,070445	99862150
Observations	1616	1616

Source: Authors' calculation

#### 4.1. Estimated Panel Data Regression Model

The estimation of the panel data regression model with the help of the Eviews 12 program shows the estimation results for CEM in Table 5; FEM in Table 6; and REM in Table 7.

TABLE 5: Estimation of Common Effect Model (CEM).

Variables	Coef.	Standard Error
village funds	-0,008***	0,003
Intercept	0,803***	0,051

\*, \*\*, \*\*\* denote significance at 10% ,5% and 1%

The results of the CEM estimate are:

$$IDM_{it} = 0,803 - 0,008 \ln DD_{it}$$

TABLE 6: Fixed Effect Model (FEM).

Variables	Coef.	Standard Error
village funds	0,112***	0,005
Intercept	-1.423***	0,100

\*, \*\*, \*\*\* denote significance at 10% ,5% and 1%

The results of the FEM estimation are:

$$IDM_{it} = - 1,423 + 0,112 \ln DD_{it}$$

TABLE 7: Random Effect Model (REM).

Variables	Coef.	Standard Error
village funds	0.047***	0,004
Intercept	-0,225***	0,075

\*, \*\*, \*\*\* denote significance at 10% ,5% and 1%

The results of the REM estimate are:

$$IDM_{it} = - 0,225 + 0,047 \ln DD_{it}$$

## 4.2. Selection of Panel Data Regression Model

Accuracy testing between CEM and FEM models via Chow Test, where:

Initial hypothesis ( $H_0$ ) = CEM

Alternative hypothesis ( $H_1$ ) = FEM

Chow Test calculation against previous CEM and FEM models in Eviews12 shows that the Prob value was 0.00 or less than  $\alpha = 5\%$ ; therefore,  $H_0$  was rejected. Hence, it is concluded that the suitable model to be used is FEM. These results showed that there were non-observed observation-unit specific effects and affected the model.

Furthermore, Hausman test was carried out to determine the accuracy between the FEM and REM models, as follows

Initial hypothesis ( $H_0$ ) = REM

Alternative hypothesis ( $H_1$ ) = FEM

The test was performed with Eviews 12 and found a Prob>chi square value of 0.00 which was smaller than  $\alpha = 5\%$ , or it is concluded that  $H_0$  was rejected. Thus, it is save to say that the more qualified model to be conducted is FEM.

After obtaining the right model through some of the previous tests, the classic assumption test are carried out to ensure the efficiency of the variables in the model. The first is the heteroskedasticity test aimed at seeing whether the residuals of the model used have a constant variance or not.  $H_0$  in this test is homoskedasticity and  $H_1$  is heteroskedasticity. Based on calculations carried out with the Wald test method on the previous FEM, a Prob>chi square value of 0.00 was obtained thus  $H_0$  was rejected, meaning that the residual of the FEM model is heteroskedastic. Then, in panel data regression models with heteroskedasticity problem, there is a further testing that needs to be done in order to see the presence of cross-sectional dependence. The test was carried out with a CD Distribution (cross-sectional dependence) test to test whether the residual are correlated across entitites (50). The  $H_0$  is that residuals are not correlated; on the contrary,  $H_1$  tells that the residuals are correlated. The result shows that the FEM has cross-sectional dependence since we reject the null hypothesis ( $Pr=0,00$ ).

As a result, FEM in this study has heteroscedasticity dan cross-sectional dependence, hence the proper model to be used is FEM with Seemingly Unrelated Regression (SUR). However, since the number of cross-section data ( $n$ ) is larger than period of the data ( $t$ ), the more suitable model to be used is cross section Seemingly Unrelated Regression Panel Corrected Standard Errors (SUR PCSE) in cross-section weights estimator(51).

### 4.3. Panel Data Regression Final Model Estimation

FEM estimation results with cross section Seemingly Unrelated Regression Panel Corrected Standard Errors (SUR PCSE) in cross-section weights estimator is as follows:

$$IDM_{it} = -1,764 + 0,130 \ln DD_{it}$$

TABLE 8: Final Estimation of FEM.

Variables	Coef.	Standard Error
village funds	0,130***	0,034
Intercept	-1,764***	0,635

\*, \*\*, \*\*\* denote significance at 10% ,5% and 1%

The value of intercent  $\alpha_i$  is different for each district/ city, presented in the appendix to this study.

The result shows that the village funds has a significant and positive effect on VBI. In addition, the estimates on the cross section model Seemingly Unrelated Regression Panel Corrected Standard Errors (SUR PCSE) in cross-section weights estimator explains that the increase in the village funds will increase the VBI in every Regency/City in Indonesia. In other words, the village funds has a positive and significant effect on VBI. Then, the model used (lin-log) is a model to calculate the absolute change of VBI to the percentage change in the village funds, so that a 1% increase in the village funds will increase the VBI by 0.0013.

### 4.4. Discussion

The results of the regression analysis shows that the village funds policy through the allocation of budgets sourced from 10% of the state budget to villages has a positive and significant effect on changes in the improvement and autonomy status of the villages in regencies/cities in Indonesia. The results of this study support previous research which also stated that the village funds had a positive effect on VBI(24,28,52).

Meanwhile, the small percentage of the effect can be caused of the data utilized in this study have not covered the distribution of the fund and the priority of using the funds for poverty reduction; for instance, there is a priority of the policy direction in 2020 for using the village in providing affirmative allocations for disadvantages and very disadvantages village(15). Consequently, in-depth analysis of how the village fund is used for poverty reduction from each district/city in Indonesia is urgently needed to support the findings of this study (26). Additionally, the small percentage of the effect

of village funds on the VBI also shows that there are other factors exist and are notable to be take into account. Future research needs to elaborate these factors.

Nevertheless, the finding of this study may indicate that the village funds policy has contributed to the achievement of the goal of alleviating rural poverty and can be said as a pro-poor policy, in line with results found by Imanuddin et al., 2019(15) and Ernawati et al. (22). Equally important, the finding of this research can provides a different view compared to the results of previous studies which stated that the village funds policy had a negatif effect on the poverty as proposed by Sigit & Kosasih (16) and Faoziyah & Salim (18)(19).

Furthermore, the findings may also denote that there is a need to elaborate and develop VBI as one of the measure tools to examine the conditions of poverty alleviation in rural areas in Indonesia as a multidimensional manner; on this occasion, by paying attention to the factors, dimensions, variables and indicators of MPI and MPL(37)(39)(38) as the measurement base. Ultimately, VBI is not only qualify the proggres and autonomy status of the villages as an instrument among the Governments in carrying out development and empowerment of village communities and specifically for the needs of mapping Village typologies and preparing priorities for the use of village funds, but it also comprehensively can be a database of multidimensional poverty in rural Indonesia.

## 5. Conclusion

Using the data of village funds and VBI from 404 districts and cities in Indonesia, this study suggests that the village funds policy (through the allocation as much as 10% of budget sourced from the State Budget) has a positive and significant effect on the proggres and autonomy status of the villages. The result obtained from the FEM with SUR PCSE in cross-section weights estimator explains that the increase in the village funds will increase the VBI in regency/city in Indonesia. Additionally, the calculation with the model (lin-log) shows that a 1% increase in the village funds will raise the VBI by 0.0013. This indicates that the village funds policy has contributed to the achievement of the goal of alleviating rural poverty or it can be said as a pro-poor policy in the context of support efforts to achieve the 1<sup>st</sup> SDGs goal.

Nevertheless, this study has limitations in terms of methods; type of data; and aspects analyzed. Therefore, further research needs to examine the influence of village funds policies on in the improvement and autonomy status of the villages both quantitatively and qualitatively, for example by including other aspects such as its effect

on multidimensional poverty in rural areas, using more comprehensive data up to the micro level thus it could find a new viewpoint to the extent of impact of village funds policy in supporting the achievement of the 1<sup>st</sup> SDGs goal.

## 6. AUTHORS' CONTRIBUTION

Rosita Novi Andari and Rizky Fitria contributed equally as the main contributors to this article. Both authors have researched, read and approved to this article.

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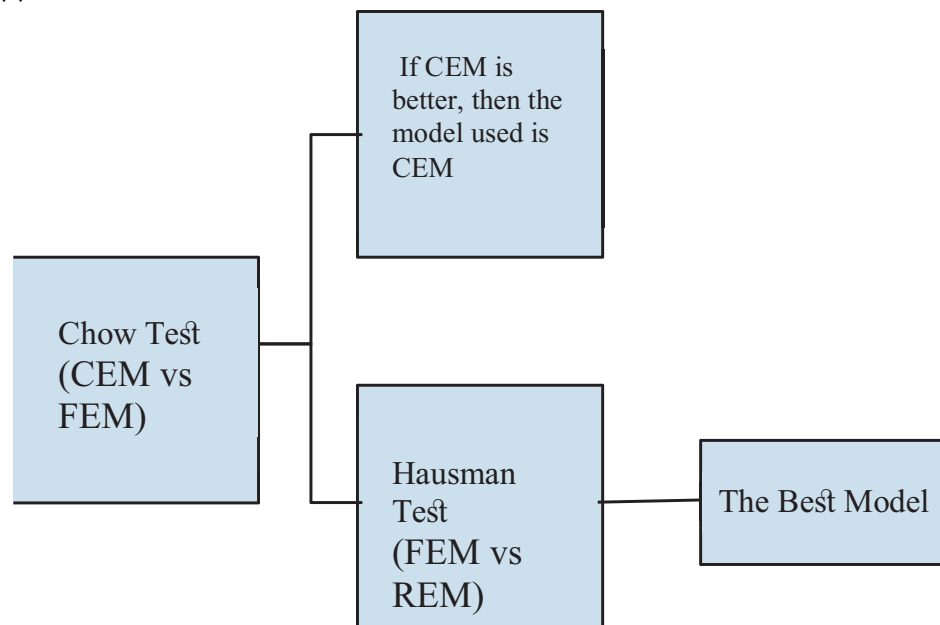
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## APPENDICES

### Appendix 1. CEM Eviews result



## Appendix 2. FEM Eviews result

Dependent Variable: IDM  
 Method: Panel Least Squares  
 Date: 10/29/22 Time: 20:42  
 Sample: 2018 2021  
 Periods included: 4  
 Cross-sections included: 404  
 Total panel (balanced) observations: 1616

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.803395	0.051163	15.70269	0.0000
LNDD	-0.007638	0.002738	-2.790195	0.0053
R-squared	0.004800	Mean dependent var		0.660724
Adjusted R-squared	0.004184	S.D. dependent var		0.070445
S.E. of regression	0.070297	Akaike info criterion		-2.470937
Sum squared resid	7.975866	Schwarz criterion		-2.464269
Log likelihood	1998.517	Hannan-Quinn criter.		-2.468462
F-statistic	7.785187	Durbin-Watson stat		0.139534
Prob(F-statistic)	0.005330			

### Appendix 3. REM Eviews result

Dependent Variable: IDM  
 Method: Panel Least Squares  
 Date: 10/29/22 Time: 20:45  
 Sample: 2018 2021  
 Periods included: 4  
 Cross-sections included: 404  
 Total panel (balanced) observations: 1616

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-1.423143	0.100050	-14.22425	0.0000
LNDD	0.111564	0.005356	20.82860	0.0000

#### Effects Specification

##### Cross-section fixed (dummy variables)

R-squared	0.896947	Mean dependent var	0.660724
Adjusted R-squared	0.862568	S.D. dependent var	0.070445
S.E. of regression	0.026115	Akaike info criterion	-4.239874
Sum squared resid	0.825902	Schwarz criterion	-2.889612
Log likelihood	3830.818	Hannan-Quinn criter.	-3.738727
F-statistic	26.08967	Durbin-Watson stat	1.385248
Prob(F-statistic)	0.000000		

## Appendix 4. Chow-test result

Dependent Variable: IDM  
 Method: Panel EGLS (Cross-section random effects)  
 Date: 10/29/22 Time: 20:50  
 Sample: 2018 2021  
 Periods included: 4  
 Cross-sections included: 404  
 Total panel (balanced) observations: 1616  
 Swamy and Arora estimator of component variances

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-0.224794	0.069477	-3.235521	0.0012
LNDD	0.047408	0.003716	12.75936	0.0000

Effects Specification		S.D.	Rho
Cross-section random		0.063703	0.8561
Idiosyncratic random		0.026115	0.1439

Weighted Statistics			
R-squared	0.079325	Mean dependent var	0.132674
Adjusted R-squared	0.078755	S.D. dependent var	0.029439
S.E. of regression	0.028256	Sum squared resid	1.288658
F-statistic	139.0619	Durbin-Watson stat	0.745008
Prob(F-statistic)	0.000000		

Unweighted Statistics			
R-squared	-0.244515	Mean dependent var	0.660724
Sum squared resid	9.973966	Durbin-Watson stat	0.096257

## Appendix 5. Hausman-test result

Redundant Fixed Effects Tests  
 Equation: Untitled  
 Test cross-section fixed effects

Effects Test	Statistic	d.f.	Prob.
Cross-section F	26.014437	(403,1211)	0.0000
Cross-section Chi-square	3664.602229	403	0.0000

Cross-section fixed effects test equation:  
 Dependent Variable: IDM  
 Method: Panel Least Squares  
 Date: 10/29/22 Time: 20:46  
 Sample: 2018 2021  
 Periods included: 4  
 Cross-sections included: 404  
 Total panel (balanced) observations: 1616

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.803395	0.051163	15.70269	0.0000
LNDD	-0.007638	0.002738	-2.790195	0.0053
R-squared	0.004800	Mean dependent var		0.660724
Adjusted R-squared	0.004184	S.D. dependent var		0.070445
S.E. of regression	0.070297	Akaike info criterion		-2.470937
Sum squared resid	7.975866	Schwarz criterion		-2.464269
Log likelihood	1998.517	Hannan-Quinn criter.		-2.468462
F-statistic	7.785187	Durbin-Watson stat		0.139534
Prob(F-statistic)	0.005330			



## Appendix 6. FEM SUR PCSE Eviews result

Correlated Random Effects - Hausman Test

Equation: Untitled

Test cross-section random effects

Test Summary	Chi-Sq. Statistic	Chi-Sq. d.f.	Prob.
Cross-section random	276.527806	1	0.0000

Cross-section random effects test comparisons:

Variable	Fixed	Random	Var(Diff.)	Prob.
LNDD	0.111564	0.047408	0.000015	0.0000

Cross-section random effects test equation:

Dependent Variable: IDM

Method: Panel Least Squares

Date: 10/29/22 Time: 20:51

Sample: 2018 2021

Periods included: 4

Cross-sections included: 404

Total panel (balanced) observations: 1616

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-1.423143	0.100050	-14.22425	0.0000
LNDD	0.111564	0.005356	20.82860	0.0000

### Effects Specification

Cross-section fixed (dummy variables)

R-squared	0.896947	Mean dependent var	0.660724
Adjusted R-squared	0.862568	S.D. dependent var	0.070445
S.E. of regression	0.026115	Akaike info criterion	-4.239874
Sum squared resid	0.825902	Schwarz criterion	-2.889612
Log likelihood	3830.818	Hannan-Quinn criter.	-3.738727
F-statistic	26.08967	Durbin-Watson stat	1.385248
Prob(F-statistic)	0.000000		

## Appendix 7. Cross-section Fixed Effects

Dependent Variable: IDM

Method: Panel EGLS (Cross-section weights)

Date: 10/29/22 Time: 20:53

Sample: 2018 2021

Periods included: 4

Cross-sections included: 404

Total panel (balanced) observations: 1616

Linear estimation after one-step weighting matrix

Cross-section SUR (PCSE) standard errors & covariance (d.f. corrected)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-1.764052	0.635074	-2.777711	0.0056
LNDD	0.129815	0.033999	3.818163	0.0001

### Effects Specification

Cross-section fixed (dummy variables)

### Weighted Statistics

R-squared	0.977475	Mean dependent var	1.140303
Adjusted R-squared	0.969960	S.D. dependent var	0.843970
S.E. of regression	0.025779	Sum squared resid	0.804804
F-statistic	130.0765	Durbin-Watson stat	1.598553
Prob(F-statistic)	0.000000		

### Unweighted Statistics

R-squared	0.895959	Mean dependent var	0.660724
Sum squared resid	0.833820	Durbin-Watson stat	1.520296

	DISTRICTS	Effect
1	Kab. Aceh Barat	-0.125591
2	Kab. Aceh Besar	-0.161435
3	Kab. Aceh Selatan	-0.096166
4	Kab. Aceh Singkil	-0.025249
5	Kab. Aceh Tengah	-0.100410
6	Kab. Aceh Tenggara	-0.127865
7	Kab. Aceh Timur	-0.159269
8	Kab. Aceh Utara	-0.292265
9	Kab. Bireuen	-0.235813
10	Kab. Pidie	-0.239848
11	Kab. Simeulue	-0.022008
12	Kota Banda Aceh	0.182717
13	Kota Sabang	0.326484
14	Kota Langsa	0.121372
15	Kota Lhokseumawe	0.142728
16	Kab. Gayo Lues	-0.029315
17	Kab. Aceh Barat Daya	-0.005515
18	Kab. Aceh Jaya	-0.008650
19	Kab. Nagan Raya	-0.082249
20	Kab. Aceh Tamiang	-0.041125
21	Kab. Bener Meriah	-0.063298
22	Kab. Pidie Jaya	-0.073185
23	Kota Subulussalam	0.043059
24	Kab. Asahan	-0.028270
25	Kab. Dairi	-0.056134
26	Kab. Deli Serdang	-0.108730
27	Kab. Karo	-0.067475
28	Kab. Labuhanbatu	0.076134
29	Kab. Langkat	-0.063351
30	Kab. Mandailing Natal	-0.184916
31	Kab. Nias	-0.176493
32	Kab. Simalungun	-0.117016
33	Kab. Tapanuli Selatan	-0.073232
34	Kab. Tapanuli Tengah	-0.088724
35	Kab. Tapanuli Utara	-0.118341
36	Kab. Toba Samosir	-0.079797
37	Kota Padang Sidempuan	0.161416
38	Kab. Pakpak Bharat	0.114502
39	Kab. Nias Selatan	-0.293027
40	Kab. Humbang Hasundutan	-0.043946
41	Kab. Serdang Bedagai	-0.048230
42	Kab. Samosir	-0.010025
43	Kab. Batu Bara	0.015341
44	Kab. Padang Lawas	-0.151045
45	Kab. Padang Lawas Utara	-0.207281
46	Kab. Labuhanbatu Selatan	0.178477
47	Kab. Labuhanbatu Utara	0.028238
48	Kab. Nias Utara	-0.172101
49	Kab. Nias Barat	-0.145124
50	Kota Gunungsitoli	-0.022399

51	Kab. Limapuluh Kota	0.141837	101	Kab. Kaur	-0.034436
52	Kab. Agam	0.157482	102	Kab. Seluma	-0.031952
53	Kab. Kepulauan Mentawai	0.096310	103	Kab. Muko Muko	0.047654
54	Kab. Padang Pariaman	0.079730	104	Kab. Lebong	0.029977
55	Kab. Pasaman	0.165266	105	Kab. Kepahiang	0.037022
56	Kab. Pesisir Selatan	-0.001447	106	Kab. Lampung Barat	0.073877
57	Kab. Sijunjung	0.156264	107	Kab. Lampung Selatan	-0.060298
58	Kab. Solok	0.074133	108	Kab. Lampung Tengah	-0.069846
59	Kab. Tanah Datar	0.198930	109	Kab. Lampung Utara	-0.100613
60	Kota Sawahlunto	0.249339	110	Kab. Lampung Timur	-0.090233
61	Kota Pariaman	0.190639	111	Kab. Tanggamus	-0.105620
62	Kab. Pasaman Barat	0.230987	112	Kab. Tulang Bawang	-0.005094
63	Kab. Dharmasraya	0.160073	113	Kab. Way Kanan	-0.052134
64	Kab. Solok Selatan	0.206417	114	Kab. Pesawaran	0.002858
65	Kab. Bengkalis	0.066078	115	Kab. Pringsewu	0.050548
66	Kab. Indragiri Hilir	-0.077227	116	Kab. Mesuji	0.039482
67	Kab. Indragiri Hulu	-0.014191	117	Kab. Tulang Bawang Barat	0.062193
68	Kab. Kampar	-0.067863	118	Kab. Pesisir Barat	-0.015296
69	Kab. Kuantan Singingi	-0.043945	119	Kab. Bandung	-0.008025
70	Kab. Pelalawan	0.088543	120	Kab. Bekasi	-0.032447
71	Kab. Rokan Hilir	-0.013557	121	Kab. Bogor	-0.112853
72	Kab. Rokan Hulu	0.002847	122	Kab. Ciamis	-0.010772
73	Kab. Siak	0.071280	123	Kab. Cianjur	-0.114196
74	Kab. Kepulauan Meranti	0.005816	124	Kab. Cirebon	-0.104294
75	Kab. Batang Hari	0.102864	125	Kab. Garut	-0.141527
76	Kab. Bungo	0.035549	126	Kab. Indramayu	-0.074414
77	Kab. Kerinci	-0.066720	127	Kab. Karawang	-0.099262
78	Kab. Merangin	-0.018481	128	Kab. Kuningan	-0.054298
79	Kab. Muaro Jambi	0.036634	129	Kab. Majalengka	-0.069715
80	Kab. Sarolangun	-0.004800	130	Kab. Purwakarta	0.003403
81	Kab. Tanjung Jabung Barat	0.025099	131	Kab. Subang	-0.012366
82	Kab. Tanjung Jabung Timur	0.048312	132	Kab. Sukabumi	-0.087769
83	Kab. Tebo	0.053909	133	Kab. Sumedang	-0.010582
84	Kota Sungai Penuh	0.286872	134	Kab. Tasikmalaya	-0.116723
85	Kab. Lahat	-0.147838	135	Kota Banjar	0.391755
86	Kab. Musi Banyuasin	-0.065196	136	Kab. Bandung Barat	0.023053
87	Kab. Musi Rawas	-0.029049	137	Kab. Pangandaran	0.113267
88	Kab. Muara Enim	-0.065064	138	Kab. Banjarnegara	-0.078608
89	Kab. Ogan Komering Ilir	-0.105388	139	Kab. Banyumas	-0.088570
90	Kab. Ogan Komering Ulu	-0.040659	140	Kab. Batang	-0.035895
91	Kota Prabumulih	0.212730	141	Kab. Blora	-0.094994
92	Kab. Banyuasin	-0.115763	142	Kab. Boyolali	-0.013508
93	Kab. Ogan Ilir	-0.079265	143	Kab. Brebes	-0.150396
94	Kab. Ogan Komering Ulu Selatan	-0.118161	144	Kab. Cilacap	-0.060740
95	Kab. Empat Lawang	-0.033079	145	Kab. Demak	-0.086470
96	Kab. Penukal Abab Lematang Ilir	0.031141	146	Kab. Grobogan	-0.093334
97	Kab. Musi Rawas Utara	0.013670	147	Kab. Jepara	-0.039346
98	Kab. Bengkulu Selatan	0.020600	148	Kab. Karanganyar	0.007756
99	Kab. Bengkulu Utara	-0.009463	149	Kab. Kebumen	-0.121839
100	Kab. Rejang Lebong	-0.011491	150	Kab. Kendal	-0.024670

151	Kab. Klaten	-0.112526	201	Kab. Bengkulu	0.027657
152	Kab. Kudus	0.068586	202	Kab. Landak	-0.086115
153	Kab. Magelang	-0.105074	203	Kab. Kapuas Hulu	-0.088526
154	Kab. Pati	-0.106294	204	Kab. Ketapang	-0.079144
155	Kab. Pekalongan	-0.086262	205	Kab. Mempawah	0.228862
156	Kab. Pemalang	-0.062362	206	Kab. Sambas	0.025933
157	Kab. Purbalangga	-0.050448	207	Kab. Sanggau	0.009612
158	Kab. Purworejo	-0.106205	208	Kab. Sintang	-0.141946
159	Kab. Rembang	-0.063825	209	Kab. Sekadau	0.089676
160	Kab. Semarang	-0.013584	210	Kab. Melawi	-0.012624
161	Kab. Sragen	0.008920	211	Kab. Kayong Utara	0.151383
162	Kab. Sukoharjo	0.047755	212	Kab. Kubu Raya	0.072428
163	Kab. Tegal	-0.125259	213	Kab. Barito Selatan	0.043432
164	Kab. Temanggung	-0.046629	214	Kab. Barito Utara	0.035538
165	Kab. Wonogiri	-0.013825	215	Kab. Kapuas	-0.097629
166	Kab. Wonosobo	-0.021732	216	Kab. Kotawaringin Barat	0.157119
167	Kab. Bantul	0.203798	217	Kab. Kotawaringin Timur	-0.043190
168	Kab. Gunung Kidul	0.095037	218	Kab. Katingan	-0.071974
169	Kab. Kulon Progo	0.127932	219	Kab. Seruyan	0.001271
170	Kab. Sleman	0.180057	220	Kab. Sukamara	0.202873
171	Kab. Bangkalan	-0.114111	221	Kab. Lamandau	0.081436
172	Kab. Banyuwangi	0.095906	222	Kab. Gunung Mas	-0.020501
173	Kab. Blitar	0.035170	223	Kab. Pulang Pisau	0.053417
174	Kab. Bojonegoro	-0.059477	224	Kab. Murung Raya	-0.090658
175	Kab. Bondowoso	-0.031445	225	Kab. Barito Timur	0.056162
176	Kab. Gresik	-0.011770	226	Kab. Banjar	-0.080515
177	Kab. Jember	-0.031007	227	Kab. Barito Kuala	-0.024783
178	Kab. Jombang	-0.036444	228	Kab. Hulu Sungai Selatan	0.041488
179	Kab. Kediri	-0.086583	229	Kab. Hulu Sungai Tengah	0.031486
180	Kab. Lamongan	-0.104059	230	Kab. Hulu Sungai Utara	-0.047249
181	Kab. Lumajang	-0.007359	231	Kab. Kotabaru	-0.034491
182	Kab. Madiun	0.047956	232	Kab. Tabalong	0.054807
183	Kab. Magetan	0.037998	233	Kab. Tanah Laut	0.045815
184	Kab. Malang	-0.043734	234	Kab. Tapin	0.048430
185	Kab. Mojokerto	-0.004437	235	Kab. Balangan	0.020850
186	Kab. Nganjuk	-0.040186	236	Kab. Tanah Bumbu	0.036094
187	Kab. Ngawi	-0.008450	237	Kab. Berau	0.061644
188	Kab. Pacitan	0.041631	238	Kab. Kutai Kartanegara	-0.007700
189	Kab. Pamekasan	-0.023129	239	Kab. Kutai Barat	-0.025958
190	Kab. Pasuruan	-0.103535	240	Kab. Kutai Timur	-0.012034
191	Kab. Ponorogo	-0.057316	241	Kab. Paser	0.035883
192	Kab. Probolinggo	-0.098133	242	Kab. Penajam Paser Utara	0.236261
193	Kab. Sampang	-0.062180	243	Kab. Mahakam Ulu	0.035527
194	Kab. Sidoarjo	-0.045600	244	Kab. Bolaang Mongondow	-0.019653
195	Kab. Situbondo	0.030879	245	Kab. Minahasa	0.038513
196	Kab. Sumenep	-0.131853	246	Kab. Kepulauan Sangihe	-0.003457
197	Kab. Trenggalek	0.054209	247	Kab. Kepulauan Talaud	0.038851
198	Kab. Tuban	-0.027485	248	Kab. Minahasa Selatan	-0.010691
199	Kab. Tulungagung	-0.025149	249	Kab. Minahasa Utara	0.077999
200	Kota Batu	0.404113	250	Kab. Kep. Siau Tagulandang Biaro	0.105638

251	Kab. Bolaang Mongondow Utara	0.044264	301	Kab. Buton Tengah	0.055217
252	Kab. Minahasa Tenggara	0.075036	302	Kab. Buton Selatan	0.065130
253	Kab. Bolaang Mongondow Timur	0.099442	303	Kab. Badung	0.339194
254	Kab. Bolaang Mongondow Selatan	0.117017	304	Kab. Bangli	0.223412
255	Kab. Banggai	-0.057068	305	Kab. Buleleng	0.080180
256	Kab. Banggai Kepulauan	-0.018259	306	Kab. Gianyar	0.246090
257	Kab. Buol	0.039292	307	Kab. Jembrana	0.258331
258	Kab. Toli-Toli	0.023244	308	Kab. Karangasem	0.187655
259	Kab. Donggala	-0.006576	309	Kab. Klungkung	0.251034
260	Kab. Morowali	-0.005571	310	Kab. Tabanan	0.189960
261	Kab. Poso	-0.013820	311	Kota Denpasar	0.395183
262	Kab. Parigi Moutong	-0.063626	312	Kab. Bima	-0.067979
263	Kab. Tojo Una Una	-0.039404	313	Kab. Dompu	0.103306
264	Kab. Sigi	-0.055784	314	Kab. Lombok Barat	0.021470
265	Kab. Banggai Laut	0.065611	315	Kab. Lombok Tengah	0.018512
266	Kab. Morowali Utara	-0.027467	316	Kab. Lombok Timur	-0.043828
267	Kab. Bantaeng	0.223316	317	Kab. Sumbawa	0.016753
268	Kab. Barru	0.221458	318	Kab. Sumbawa Barat	0.164160
269	Kab. Bone	-0.110740	319	Kab. Lombok Utara	0.143967
270	Kab. Bulukumba	0.068614	320	Kab. Alor	-0.129317
271	Kab. Enrekang	-0.007205	321	Kab. Belu	0.007158
272	Kab. Gowa	0.036625	322	Kab. Ende	-0.143332
273	Kab. Jeneponto	0.039659	323	Kab. Flores Timur	-0.043843
274	Kab. Luwu	-0.095855	324	Kab. Kupang	-0.125386
275	Kab. Luwu Utara	-0.033064	325	Kab. Lembata	-0.060244
276	Kab. Maros	0.076543	326	Kab. Manggarai	-0.087125
277	Kab. Pangkajene dan Kepulauan	0.080832	327	Kab. Ngada	-0.002666
278	Kab. Luwu Timur	0.027410	328	Kab. Sikka	-0.083635
279	Kab. Pinrang	0.079796	329	Kab. Sumba Barat	-0.018512
280	Kab. Sinjai	0.121317	330	Kab. Sumba Timur	-0.064901
281	Kab. Kepulauan Selayar	0.023463	331	Kab. Timor Tengah Selatan	-0.215543
282	Kab. Sidenreng Rappang	0.147137	332	Kab. Timor Tengah Utara	-0.110000
283	Kab. Soppeng	0.156420	333	Kab. Rote Ndao	-0.016936
284	Kab. Takalar	0.066307	334	Kab. Manggarai Barat	-0.112100
285	Kab. Tana Toraja	-0.058236	335	Kab. Nagekeo	0.007729
286	Kab. Wajo	0.020689	336	Kab. Sumba Barat Daya	-0.170278
287	Kab. Toraja Utara	-0.088771	337	Kab. Sumba Tengah	-0.000952
288	Kab. Buton	0.061809	338	Kab. Manggarai Timur	-0.109219
289	Kab. Konawe	-0.121335	339	Kab. Sabu Raijua	-0.033680
290	Kab. Kolaka	0.052491	340	Kab. Malaka	-0.083036
291	Kab. Muna	-0.053685	341	Kab. Maluku Tenggara Barat	0.082182
292	Kab. Konawe Selatan	-0.126493	342	Kab. Maluku Tengah	-0.044617
293	Kab. Bombana	0.002659	343	Kab. Maluku Tenggara	0.016304
294	Kab. Wakatobi	0.070861	344	Kab. Buru	0.028494
295	Kab. Kolaka Utara	-0.044769	345	Kota Ambon	0.263744
296	Kab. Konawe Utara	-0.040939	346	Kab. Seram Bagian Barat	-0.002778
297	Kab. Buton Utara	0.036412	347	Kab. Seram Bagian Timur	-0.102642
298	Kab. Konawe Kepulauan	0.004141	348	Kab. Kepulauan Aru	-0.089433
299	Kab. Kolaka Timur	-0.009123	349	Kota Tual	0.223226
300	Kab. Muna Barat	0.030299	350	Kab. Maluku Barat Daya	-0.066574