

# The Effect of GDP per Capita, Population, and Income Inequality on CO<sub>2</sub> Emissions in Indonesia

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*Abstract:* - This study aims to see the effect of GDP per capita, income inequality, and population on CO<sub>2</sub> emissions in Indonesia from 1990-2021. This research uses a descriptive quantitative method. The data used is secondary data, in the form of annual data for 32 years. The analytical method used is the error correction model (ECM) to see the short and long-term effects between the independent variable and the dependent variable. The results of this study indicate that GDP per capita has a positive and significant effect on Indonesia, both in the short term and in the long term. The income inequality variable has a positive and insignificant effect on CO<sub>2</sub> emissions in Indonesia in the short term. Meanwhile, in the long term, income inequality has a negative and insignificant effect on CO<sub>2</sub> emissions in Indonesia. The population variable has an insignificant negative effect on CO<sub>2</sub> emissions in Indonesia in the short term. However, in the long term, the population has a significant positive effect on CO<sub>2</sub> emissions in Indonesia.

*Key-Words:* - CO<sub>2</sub> Emissions, ECM, GDP per Capita, Income Inequality, Indonesia, Population.

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## 1 Introduction

Environmental degradation is a topic that is often raised because it is a serious problem at the world level. The most serious impact due to environmental degradation is global warming. Global warming is caused by the rise of Greenhouse Gases (GHG). Indonesia continues to experience an increase in GHG emissions from year to year. Based on data from [1], shows that the energy sector is the largest contributor to GHG emissions in Indonesia, which is 34%, followed by the waste sector (7%), agriculture (6%), and IPPU (3%). This shows that the energy sector contributes to national GHG emissions. Based on IPCC GL 2006 guidelines, [2], gases from the energy sector consist of CO<sub>2</sub>, CH<sub>4</sub>, and N<sub>2</sub>O. In 2019 in Indonesia the amount of CO<sub>2</sub> emissions amounted to 607,368 Gg CO<sub>2</sub>e, followed by CH<sub>4</sub> 27,181 Gg CO<sub>2</sub>e, and NO<sub>2</sub> (3,903 Gg CO<sub>2</sub>e), [3].

Environmental degradation is a decrease in environmental quality caused by natural and human factors. The main factor that causes environmental degradation is the human factor. Human factors that cause environmental degradation include industrial activities, land use change, the use of fossil energy, and others. Environmental degradation is driven by a country's need to promote economic growth and development and meet human needs, [4].

Economic growth can be used to improve the public welfare, [5]. Economic growth indicates an increase in the country's productivity to produce goods and services, [6]. Gross Domestic Product is the main indicator characterizing economic growth, [7]. To increase economic growth, carrying out economic activities and energy consumption is necessary. According to [8], energy is an important parameter to fulfill basic human needs from the food chain to carrying out various economic activities. The biggest challenge for developing countries is being able to maintain economic growth while maintaining environmental quality, [9].

Increasing economic growth without improving the structure of development causes problems of inequality in society. Income inequality occurs due to the gap in income distribution among community groups. According to [10], income inequality is a factor causing environmental pollution in developing countries. Income inequality causes the government to focus on economic growth policies only, without regard to environmental aspects, [11]. Economic growth efforts to reduce income inequality lead to increased resource use and energy consumption. This is a factor causing the increase in CO<sub>2</sub> emissions.

As an effort to increase GDP, it takes people or humans as development actors. The population

becomes an economic actor, both as a producer and a consumer. Currently, the population of Indonesia continues to increase from year to year. According to [12], an ever-increasing population will be followed by an increase in demand for goods and services which in turn increases the use of natural resources. The increase in demand for goods and services affects the increase in industrial activity. In addition, the growing population also causes an increase in the use of energy such as fossil fuels which results in environmental degradation in the form of CO<sub>2</sub> emissions. Efforts to reduce CO<sub>2</sub> emissions can be realized through poverty alleviation initiatives, which is highly prioritized in developing countries, [13].

The focus of this study is the rapid increase in GDP every year, income inequality, and increasing population that causes CO<sub>2</sub> emissions in Indonesia. This research combines economic, environmental, and social aspects contained in the concept of sustainable development. This study looks at the effect of GDP per capita, population, and income inequality on CO<sub>2</sub> emissions in Indonesia both in the short and long term.

Indonesia continues to experience rapid economic growth. The indicator that characterizes economic growth is Gross Domestic Product (GDP). GDP can be defined as the sum of all value added at every stage of production within a defined region, [14]. Indirectly, efforts to increase GDP encourage an increase in production and industrial activity. Economic growth is a parameter that determines the success of economic development, but on the other hand, can cause environmental degradation in the form of CO<sub>2</sub> emissions. According to [15], exist a trade-off between economic growth and environmental preservation.

The increasing GDP figure does not guarantee that Indonesia is free from social problems in the form of income inequality. The richest 10% of people in Indonesia control 75.7% of the national wealth, and the richest 1% of people in Indonesia control 49.3% of the national wealth, [16]. The data proves that Indonesia still experiences income inequality, where inequality can affect CO<sub>2</sub> emissions.

According to Adam Smith's theory, one of the most important components of economic growth is population. The increase in population in Indonesia is also accompanied by an increase in economic growth. However, the increase in population causes an increase in the use of natural resources that cause pollution. The increasing population also has an impact on increasing energy use which causes CO<sub>2</sub> emissions. Based on this explanation, the researcher

determines the following problem formulation: (1) how did GDP per capita affect CO<sub>2</sub> emissions in Indonesia in 1990-2021?; (2) how does population affect CO<sub>2</sub> emissions in Indonesia in 1990-2021?; and (3) how did income inequality affect CO<sub>2</sub> emissions in Indonesia in 1990-2021?

## 2 Methodology and Variables

The research method used is a quantitative method with a descriptive approach. Quantitative methods are research methods in the form of numbers measured by statistical tests to provide conclusions. The descriptive approach used serves to describe the results of research by presenting, analyzing, and interpreting them. The scope of this research is Indonesia. The data used is annual data or time series for 32 years starting from 1990-2021—research data obtained from the World Bank and Our World in Data.

This study consisted of three independent variables and one dependent variable. The independent variables used are GDP per capita in units of US \$, variables in population with units of thousands of people and variables of income inequality measured using the Gini ratio. The dependent variable is CO<sub>2</sub> emissions in tons. CO<sub>2</sub> emissions taken into account in this study are only emissions derived from fossil and industrial energy.

Time series data requires stationary data. So before estimating data, it is necessary to perform a stationary test. Data is said to be stationary if the data does not have drastic changes. The first data analysis carried out was a stationary test. The stationarity test conducted in this study used the Dickey-Fuller Augmented method by comparing the t-statistical ADF with MacKinnon's critical value. If the ADF value of t-statistics is greater than the critical value of MacKinnon 5%, then the data is stationary. If the stationary test data shows results that are not yet stationary, then an integration test is carried out. Integration tests are performed to see to what degree the data will be stationary. Furthermore, the cointegration test uses the Engel Granger (EG) test. The Engel-Granger test can determine the cointegration of stationarity in its residuals.

Data estimation in this study uses an error correction model (ECM). ECM estimation aims to determine whether there are short-term and long-term influences on the variables tested, [17]. The data used in this study is time series data. The advantage of using the ECM method is to overcome the shortcomings of a common method, namely Ordinary Least Square (OLS) which cannot be used

when the variable is not stationary. Equation The ECM in this study is as follows:

$$\Delta CO_2 = \alpha_0 + \alpha_1 \Delta PDB_t + \alpha_2 \Delta GR_t + \alpha_3 \Delta P_t + \varepsilon_t$$

For the regression equation in the long term, it is written as follows:

$$CO_2 = \beta_0 + \beta_1 PDB_t + \beta_2 GR_t + \beta_3 P_t + \varepsilon_t$$

Information: CO<sub>2</sub>; CO<sub>2</sub> emissions; PDB: GDP per Capita; GR: income inequality (Gini ratio); P: population; α<sub>0</sub> and β<sub>0</sub>: constant; α<sub>1</sub>, α<sub>2</sub>, α<sub>3</sub> and β<sub>1</sub>, β<sub>2</sub>, β<sub>3</sub>: regression coefficient; and ε: error term.

The error correction model method is characterized by the presence of an element of error correction term (ECT). ECT is a residual that appears in the ECM model. If the value of the ECT coefficient < 1 and is significant at 5%, then the specification model used is valid.

After obtaining the research model, the next stage is to test classical assumptions. The classical assumption tests used in this study are normality tests, autocorrelation tests, heteroscedasticity tests, and multicollinearity tests. The normality test uses the Jarque-Bera test, if the JB value > α 5%, then the residual is normally distributed. Autocorrelation test using Durbin-Watson test, if DW value is between -2 to +2, then there is no autocorrelation problem. Test heteroscedasticity using the Breusch-Pagan-Godfrey test, if the value of Prob. Chi-Square is more than 0.05, so there is no heteroscedasticity problem. Multicollinearity test using VIF test, if the test result is below 10, then there is no multicollinearity problem.

### 3 Result and Discussion

#### 3.1 Result

Time series data requires stationary data. The stationarity test using the Dickey-Fuller Augmented method is shown in Table 1. Based on the results of the stationary test, it shows that the ADF test value on all variables is smaller than the MacKinnon critical value and the probability value is more than α 5%, so that all variables are not stationary at the level. Furthermore, a differentiation test is carried out to find out the degree of integration to how much the data will be stationary. Table 2, is the integration test in this study.

Based on the results of the integration test (Table 2), it shows that the ADF test value on all variables is greater than the MacKinnon critical value and the Probability value is less than α 5%, so all variables are stationary at the level of first difference. Because all variables are stationary at the first difference level, the next stage is to conduct a cointegration test to be able to perform ECM estimation. The cointegration test is shown in the following Table 3.

Based on the results of the cointegration test, it shows that the probability value is 0.0012 < α 5% and the ADF test value is more than the critical value. Thus, the equation tested has a long-term equilibrium relationship. So that the estimation model can be interpreted further.

This study uses the ECM Domowitz El-Badawi estimation model to determine the short-term and long-term effects of GDP per capita, income inequality, and population on CO<sub>2</sub> emissions. The results of regression in the short term are shown in Table 4.

Based on the results of estimates in the short term, the regression equation is obtained as follows:

$$CO_2 = 62045241 + 40630.62PDB + 698885.5GR - 16.79596P - 0.832239$$

The equation shows that the value of the constant is 62045241, meaning that if the value of all independent variables is zero, then the value of CO<sub>2</sub> emissions is 62045241 tons. The value of the coefficient in the variable GDP per capita is 40630.62, meaning that when GDP per capita increases by 1 US\$, CO<sub>2</sub> emissions will increase by 40630.62 tons (ceteris paribus). The coefficient in the income inequality variable is 698885.5, meaning that when inequality increases by 1%, CO<sub>2</sub> emissions will increase by 698885.5 tons (ceteris paribus). The coefficient on the variable population is -16.79596, meaning that when the population increases by 1 million, CO<sub>2</sub> emissions will decrease by 16.79596 tons (ceteris paribus). Meanwhile, the value of the coefficient in the ECT variable is -0.832239, because it has a negative sign (ECT < 1) and is significant at α 5%, the model specification used is valid. The R-square value has a coefficient of 0.593224, meaning that GDP per capita, income inequality, and population together can explain 59.3224% of CO<sub>2</sub> emissions. While the rest is explained by other variables outside the research model.

Table 1. Stationarity Test at Level

Variable	ADF Test Scores	McKinnon Critical Values			Prob.	Information
		1%	5%	10%		
CO <sub>2</sub>	-0.492203	-3.661661	-2.960411	-2.619160	0.8797	Non-stationary
PDB	0.248447	-3.661661	-2.960411	-2.619160	0.9714	Non-stationary
GR	-0.981690	-3.661661	-2.960411	-2.619160	0.7473	Non-stationary
P	-1.889931	-3.737853	-2.991878	-2.635542	0.3311	Non-stationary

Table 2. Integration Test on First Difference

Variable	ADF Test Scores	McKinnon Critical Values			Prob.	Information
		1%	5%	10%		
Co2	-5.405547	-3.679322	-2.967767	-2.622989	0.0001	Stationary
.PDB	-4.314652	-3.670170	-2.963972	-2.621007	0.0020	Stationary
GR	-4.396564	-3.670170	-2.963972	-2.621007	0.0016	Stationary
P	-3.036626	-3.699871	-2.976263	-2.627420	0.0441	Stationary

Table 3. Cointegration Test

Augmented Dickey-Fuller test Statistics	t-Statistic	Prob.*
	-4.507542	0.0012
Test Critical values: 1% level	-3.661661	
5% level	-2.960411	
10% level	-2.619160	

Table 4. Short-Term Estimation Results

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	62045241	77047907	0.805281	0.4280
D(GDP)	40630.62	17211.78	2.360629	0.0260
D(INEQUALITY)	698885.5	3124697.	0.223665	0.8248
D(POPULATION)	-16.79596	25.13333	-0.668274	0.5098
ECT(-1)	-0.832239	0.195686	-4.252934	0.0002
R-squared	0.593224			
Adjusted R-squared	0.530643			
F-statistic	9.479298			
Prob(F-statistic)	0.000073			

Table 5. Long-term Estimation Results

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-3.19E+08	1.38E+08	-2.319187	0.0279
.PDB	50154.20	12143.09	4.130268	0.0003
INEQUALITY	-4716203.	2911658.	-1.619765	0.1165
POPULATION	3.344576	0.395793	8.450317	0.0000
R-squared	0.983160			
Adjusted R-squared	0.981355			
F-statistic	544.8930			
Prob(F-statistic)	0.000000			

Table 4 also shows that based on the t-test the variable GDP per capita has a significant positive effect (Prob < 0.05), the income inequality variable has a positive effect is not significant (Prob > 0.05), and the population variable has a negative effect is not significant (Prob > 0.05) on CO<sub>2</sub> emissions in Indonesia in the short term. Meanwhile, simultaneously (test f) all variables together have a significant effect on CO<sub>2</sub> emissions in Indonesia (Prob f-statistics < 0.05).

Next is the classical assumption test which aims to find out whether the estimated results violate classical assumptions or not. The first classical assumption test is the normality test. Based on the results of the normality test, a Jarque-Bera value of 0.900505 > 0.05 was obtained, so that the data can be concluded as normally distributed. Then based on the results of the autocorrelation test showed that the Durbin-Watson value in this study was 1.920450, it can be concluded that there is no autocorrelation problem in the regression model. Based on the results of the heteroscedasticity test show that the value of Prob. ChiSquare is 0.4816 > 0.05, so it can be concluded that there is no heteroscedasticity problem. Based on the results of the multicollinearity test, a VIF value of less than 10 is obtained on each variable, it can be concluded that there is no problem with heteroscedasticity in regression models.

Based on the results of the estimation in the long term (Table 5), the regression equation is obtained as follows:

$$\text{CO}_2 = -3.19 + 50154.20\text{PDB} - 4716203\text{GR} + 3.344576\text{P}$$

The regression equation shows that the constant value is -3.19, meaning that in the long-term if all independent variables are zero, the CO<sub>2</sub> emission value is -3.19 tons. The value of the coefficient in the variable GDP per capita is 50154.20, meaning that if GDP per capita increases by 1 US \$ then CO<sub>2</sub> emissions will increase by 50154.20 tons (ceteris paribus). The value of the variable coefficient of income inequality is -4716203, meaning that if income inequality increases by 1%, CO<sub>2</sub> emissions will decrease by 4716203 tons (ceteris paribus). The value of the coefficient on the population variable is 3.344576, meaning that if the population increases by 1 million people, CO<sub>2</sub> emissions will increase by 3.344576 tons (ceteris paribus). Meanwhile, the R-squared coefficient of 0.983160 means that GDP per capita, income inequality, and population together can explain 98.3160% of CO<sub>2</sub> emissions. While the rest

is explained by other variables outside the research model. Table 4 shows that based on the t-test the variable GDP per capita has a significant positive effect (Prob < 0.05), the income inequality variable has a negative effect is not significant (Prob > 0.05), and the population variable has a significant positive effect (Prob < 0.05) on CO<sub>2</sub> emissions in Indonesia in the long run. Meanwhile, based on simultaneous tests (test f) show that all independent variables have a significant effect on CO<sub>2</sub> emissions in Indonesia in the long term. This can be seen from the statistical probability value f amounting to 0.000000 < 0.05.

### 3.2 Discussion

#### The Effect of GDP per Capita on CO<sub>2</sub> Emissions in Indonesia

The variable GDP per capita has a positive and significant influence on CO<sub>2</sub> emissions in Indonesia in 1990-2021, both in the long and short term. The results of this test are the same as the research conducted by [18]. The research provides results that in the short and long-term GDP per capita has a positive effect on CO<sub>2</sub> emissions in Indonesia. Efforts to increase GDP require economic activities such as consumption and production. The ever-increasing GDP shows that people's purchasing power is getting bigger. The higher the consumption, the higher the production in industries that require the use of fossil energy. This is a trigger for CO<sub>2</sub> emissions. So it can be concluded that the increase in GDP per capita in Indonesia causes an increase in CO<sub>2</sub> emissions through increased consumption of fossil energy and industrial activities.

#### The Effect of Income Inequality on CO<sub>2</sub> Emissions in Indonesia

The estimation results in this study show that income inequality variables have a positive and insignificant influence on CO<sub>2</sub> emissions in Indonesia in the short term. Meanwhile, in the long run, income inequality has a negative insignificant influence on CO<sub>2</sub> emissions in Indonesia. The results of this study are the same as the research conducted by [19], [20]. The study showed that there was no significant effect between income inequality and CO<sub>2</sub> emissions. The mechanism of the effect of income inequality on CO<sub>2</sub> emissions in the short term can be explained through efforts to increase economic growth. Income inequality drives up GDP through increased production that requires energy use. This is the main trigger for CO<sub>2</sub> emissions. This reason is also supported by the

focus on development that is only concerned with economic growth rather than environmental sustainability. The relationship between income inequality and CO<sub>2</sub> in the long run is negative. This condition can occur due to efforts to reduce income inequality and develop environmentally friendly technological innovations.

### **The Effect of Population on CO<sub>2</sub> Emissions in Indonesia**

Based on the results of the study, in the short term, the population has an insignificant negative influence on CO<sub>2</sub> emissions. However, in the long term, the population has a significant positive impact on CO<sub>2</sub> emissions. The results of the study are the same as the research conducted by [21]. The increase in population leads to an increase in people's energy needs. In addition, it led to an increase in production from various aspects. So if not accompanied by environmentally friendly policies, the population has a major influence on increasing CO<sub>2</sub> emissions. In the short term, the population can be negatively affected due to efforts to reduce fossil energy and reduce industrial activities.

## **4 Conclusion**

Economic growth, which in this case uses the variable GDP per capita has a significant positive influence both in the short and long term on CO<sub>2</sub> emissions in Indonesia in 1990-2021. Based on this, it is concluded that economic growth in Indonesia causes an increase in CO<sub>2</sub> emissions. The increase can come from an increase in fossil energy consumption and industrial activities.

Likewise, income inequality in the short term has an insignificant positive influence on CO<sub>2</sub> emissions in Indonesia. Meanwhile, in the long run, income inequality has a negative insignificant influence on CO<sub>2</sub> emissions in Indonesia. Income inequality drives up GDP through increased production that requires energy use. This is the main trigger for CO<sub>2</sub> emissions.

The estimation results in this study show that population has a negative insignificant influence on CO<sub>2</sub> emissions in the short term. While in the long run, the effect becomes positive and significant. The increasing population has led to an increase in people's need for energy and increased production, causing an increase in CO<sub>2</sub> emissions. However, in the short term, the influence of population on CO<sub>2</sub> emissions becomes negative due to the effect of decreasing economic growth which reduces the use of fossil energy and reduced industrial activities.

Although GDP increases CO<sub>2</sub> emissions, we must still build the economy while considering the environment. One way is through green economic transformation, which requires support from all parties, including the government, private sector, and the wider community. Development should be more evenly distributed and not concentrated in certain areas specifically for areas that are lagging, policies implemented must be different (acceleration development strategies). This is done so that equality can be achieved soon. Not only that, the government also socializes to the public to increase awareness of the environment. In addition, it can also be done through policies that provide easy access to general transformation for the public to reduce CO<sub>2</sub> emissions.

The limitation of this research is not including variables related to the industry, where it is known that the industry is one of the contributors to CO<sub>2</sub> emissions. Therefore, future research can include the number of large and medium industries as well as other significant variables. This aims to identify the policy interference needed.

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Heru Wahyudi made a research framework and collected literature reviews, wrote the research, proposed policy recommendations, and collected and processed research data.

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The author has no conflicts of interest to declare.

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