# Impact of Income Inequality on Poverty Level in Nigeria: Evidence from ARDL

# Model

## ABSTRACT

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| --- |
| This study used the Auto Regressive Distributed Lag (ARDL) model to assess the impact of income disparity on the degree of poverty in Nigeria. Based on the association each income determinant showed with the Gini coefficient of income disparity in the Nigerian economy, the study's findings indicated that there is a long-run positive relationship between the poverty level and the Gini coefficient of income inequality in Nigeria. Therefore, it is advised to improve a more equitable distribution of wealth, which will effectively lessen income inequality and poverty in Nigeria. The government should concentrate its efforts on developing and implementing more realistic employment programs in Nigeria. Since the empirical results of this study have demonstrated that decreasing income disparity in Nigeria by an increase in the employment rate has not been sufficient. A more practical approach to employment would allow people to use their money to build wealth rather than just get by, which would increase the level of income distribution. |

*Keywords: ARDL modeling; income inequality; poverty; level.*

**JEL Classification Code:** E6; A1

## 1. INTRODUCTION

The Global and Shared Prosperity Index broadens the way we define and measure poverty [1]. There is widespread concern about global inequality and economic growth that the economic crisis has only widened the gap between the rich and the poor [2]. This has necessitated the need to examine the pattern of inequality in global and emerging countries. This has triggered a lot of concern to examine the multiple causes linked to growing income inequalities caused by globalization, technological change and changes in distribution policies. Over the years, the Organization for Economic Cooperation and Development (OEDC) has collected a significant level of evidence on the extent and drivers of inequalities, social mobility and equal opportunity including policy responses on how to address these issues. This evidence has shown how reducing inequality globally can be highly rewarding for the global society as a whole [3].

Income inequality and the level of poverty are critical indicators of economic development [4,5]. Developed nations are those with reduced economic inequality and low rates of poverty, whereas underdeveloped or developing nations are those with highly skewed or uneven income inequality and high rates of poverty [6]. Because of the impact that growing income inequality has on the political, social, and economic stability of a country, income inequality and poverty have remained hot topics in both local and global debates [7]. This is also evident in the sustainable development goals of the United Nations, which call for, among others, the eradication of poverty and an increase in the income of the poorest 40% of developing nations. Additionally, the International Monetary Fund (IMF) highlighted the significance of income inequality as an important factor in economic growth [8].

There have been widespread worries that economic progress is not being equitably distributed throughout the World. Poverty and income inequality are two interrelated and mutually reinforcing developmental issues in Africa [9,10]. Oxfam [11] noted that the continent remains afflicted by an alarmingly high and rising inequality and entrenched poverty. [12] stated that despite strong economic growth in many African countries, human development and poverty indicators have not progressed as expected, fueling a renewed interest in the study of inequality. It is centred on multidimensional poverty measures that are anchored on household consumption and the international poverty lines of 1.9 dollars per person a day but broaden the measures to include information on access to include education and basic infrastructures [13]. It also sheds more light on the poverty differences within households gender, and age. The weak poverty-alleviation elasticity of growth [14].

The National Bureau of Statistics (NBS) recently released the 2019 poverty and inequality in Nigeria. The report highlights that 40% of the population or almost 83 million people live below the country's poverty line of 137,430 naira ($381.75) per year. The NBS report is based on data from the latest round of Nigeria's Living Standard Survey, conducted from 2018 to 2019 with support from the World Bank's poverty global practices and technical assistance from the NLSS program. Nigerian Living Standard Survey (NLSS) is the official survey that is the basis for measuring poverty and living standards in the country and is used to estimate a wide range of socio-economic indicators including benchmarking of the sustainable development goals. Between September 2018 and October 2019, the National Bureau of Statistics conducted the latest round of NLSS a decade after the previous one. The World Bank provided technical support to the NBS throughout the implementation, introducing several methodological improvements that led to the availability of reliable data for the poverty estimate (NBS, 2020).

In Nigeria, the scale of inequality is quite extreme. [15,9,10] particularly noted that the paradox of growth in Nigeria is such that as the country gets richer, only a few benefits, while the majority continue to suffer from poverty and deprivation. Recent data according to Quartz Africa (2018) reveals that 86.9 million Nigerians now live in extreme poverty representing nearly 50% of its estimated 180 million population- Nigeria is multi-dimensionally poor. Furthermore, the Oxfam [11] report ranking African nations by their commitment to tackling inequality ranked Nigeria as the 45th out of 45 countries, stating that Nigeria has the unenviable distinction of being at the bottom of the African ranking, as well as its global ranking for two years running.

Achieving inclusive growth is hampered by negative income disparity. Income disparity is receiving more empirical attention as a result of the unfavourable environment it creates for growth. As a result, [16] argued that addressing poverty also necessitates policies to address inequality, asserting that it is conceivable to be worried about poverty while being indifferent to inequality. Therefore, the importance of being concerned about economic disparity cannot be overstated because it has a significant impact on poverty. This is because greater inequality will almost always imply greater levels of poverty for a given level of average income [17]. Similar to how rising levels of poverty will necessitate redistribution toward the poor, addressing the issue of pervasive income inequality will also require developing policies, which form the basis for this study, which covers the years 1980 to 2021.

Therefore, the main issue with this study is that, despite previous policy interventions such as the National Poverty Eradication Programme, Family Economic Advancement Programme, and Family Support Programme, among others to achieve inclusive growth, which is defined by increased GDP reduction in inequality and, consequently, poverty, the issue of poverty still exists, necessitating a critical evaluation of how inequality affects poverty. This study intends to look at the impact of income inequality on the Nigerian poverty level.

## 2. LITERATURE REVIEW

Kopp [18] defined income inequality as “an extreme disparity of income inequality with a high concentration of income usually in the hands of a small percentage of a population". Therefore, there is a big difference in wealth between different demographic groups when there is income inequality. Income inequality and income disparity segregations can be assessed by a variety of segmentations, including occupation, historical income, male versus female, ethnicity, and geographic area [19]. Because different types of income inequality are examined using different segmentations of income disparity analysis, income inequalities by demographic segmentation serve as the basis for research on income inequality and disparity.

From a country's perspective, a variety of internal and foreign factors may influence income disparity [20]. One such significant external factor is globalization [21]. The relationship between income inequality, growth, and globalization has long drawn the attention of academics. However, except for a partial view in studies like [22] who looked at the relationship between inequality and the economic components (trade, FDI and financial openness) of Globalisation, the lack of a Globalisation index has not allowed statistical estimation and testing of the relationship.

According to [23] the term poverty refers to the state or condition in which people or communities lack the financial resources and essentials for a minimum standard of living. As such, basic human needs cannot be met. Poverty-stricken people and families may go without proper housing, clean water, healthy food and medical attention. Each nation may have its criteria for determining the poverty line and counting how many of its people live in poverty. It's important to remember that poverty is a socioeconomic condition that is the result of multiple factors- not just income. These factors include race, sexual identity, sexual orientation and little to no access to education among others. Poverty is both an individual concern as well as a broader social problem on the individual or household level. Not being able to meet needs can lead to a range of physical and mental issues. At the societal level, a high poverty rate can be a damper on economic growth and be associated with problems like crime, unemployment, urban decay, education, and poor health. The government often put socio-economic welfare program in place to help lift individuals, families and community out of poverty [23]. Some countries have a stronger welfare state (social safety net) than others. The U.S. for instance, tends to be much more individualistic and sure welfare programs. European countries in comparison, have a much broader range of welfare programs and support for those in need. The Department of Health and Human Services (HHS) designates those who do not make it into a particular income household as being in poverty. The U.S. Census Bureau calculates the poverty rate in the United States or the proportion of the population that lives in poverty. The poverty line in 2022 is $27,750 annually for a family of four with two children under the age of 18. In 2022, the poverty line will be $18,310 per year for two persons over 65 without a child under the age of 18 [23].

Morduch & Sicular (2018) introduced a new regression-based method for breaking down inequality indices using household-level data. They also looked at the advantages and disadvantages of breaking down inequality by income source in comparison to how they are typically interpreted. The method breaks down aggregate inequality indices using estimated income flows from variables in linear income equations. In a multivariate environment, the integrated approach offers an effective and adaptable way to measure the contributions of variables including education, age, infrastructure, and social status. The evidence from China indicates the benefits of the suggested, integrated method and highlights the stark disparities that might occur when using decomposition techniques with different features. The empirical findings highlight the role that spatial segmentations play in escalating inequality: in the sample, the village of residence greatly influences inequality. The relatively equitable distribution of human capital, especially demographic factors, acts as a partial counterbalance to this force. Affiliation with the communist party and indicators of social position, in contrast to previous recent studies, play a very small influence in explaining inequality. This also was done in rural China and not Nigeria.

However, Alayande [24] decomposed income inequality and poverty in Nigeria using the regression-based decomposition approach created by Morduch & Sicular (2018). She found that while the number of unemployed people in a household had a positive impact on income inequality, primary and postsecondary educational attainment was important in reducing it. This investigation focuses on unemployment in households and not on improved knowledge or effective labour.

Additionally, Jacobs (2020) discovered through empirical research that the percentage of total income disparity in Japan, Taiwan, and South Korea can be attributed to inequalities in age groups is relatively low (less than 5%). When compared to the mean of different age groups, inequality is significantly more common among people who fall into the same age category. In other words, neither of the three countries' wealth inequality nor a large portion of it can be attributed to age. Also, Nigeria was not included in this study.

Moreover, Adesimi [25] analyzed the structure of rural-urban income inequality vis-à-vis occupational groups and surveyed the four major states in the western part of the country that is, Lagos, Ogun, Ondo and Oyo States in Nigeria. To weigh and evaluate the rural and urban sectors of the economy, factors such as population, key economic activity, services, and industrialization level were taken into consideration. He noted that in the three states of Oyo, Ondo, and Ogun for which data are available, the rural sector received 38.3% of the taxpayer's income.

Kennedy [26] examined the impact of governance on income inequality and income inequality in Nigeria using the Federal Government's Trader Moni Social Intervention Policy Programme. The study used content analysis to examine the strategy between 2018 and 2019. The execution of Trader Moni's social intervention policy was deemed to have fallen short of achieving its stated goal of income redistribution. According to the study, Africa's level of economic disparity is comparable to that of Latin America or India. It demonstrated the extreme inequality in southern and central Africa. Additionally, the within-country component accounts for the majority of the continent's income inequality, while the between-country component has marginally decreased over the past two decades as a result of faster growth in developing nations. Furthermore, except in southern Africa, inequality was comparatively consistent across the period. The duality between agriculture and other industries, including mining rents, was another significant factor in determining inequality. This was focused on the Federal Government social policy intervention programme which is different from the parameter used in this study.

Based on the literature reviewed, there appeared to be a gap specifically in the relationship between income inequality and poverty level in Nigeria. This study investigates the mechanism through which income inequality affects the poverty level such as low level of income. Also, this research is required to understand the real relationship between income inequality and poverty level so that governments at various levels should bring up policies that will break down the level of inequality in income distribution [27-32].

## 3. METHODOLOGY

**3.1 Source and Analysis of Data**

This study made use of secondary data from CBN (2011) and WDI (2022). The study examined data from 1980 to 2021 for its analysis.

The Autoregressive Distributed Bounds Test (ARDL) was used to accomplish the goals of this investigation. The adoption of the ARDL technique suggested by Pesaran, Shin & Smith (2001) is the result of its advantages over traditional co-integration techniques when applied to small sample sizes, its ability to simultaneously test both short-run and long-run relationships while providing unbiased results, and the fact that it tests variables regardless of whether they are differences of order zero or order one.

**3.2 Theoretical Framework and Model Specification**

The social capital theory, which emphasizes that social connections are resources that can contribute to the growth and accumulation of human capital or not, forms the foundation of the study. The hypothesis is predicated on the idea that social interactions may make it easier for people to accrue benefits for themselves, whether they be financial or not. The impact of social ties, here represented by income disparity, on poverty, which symbolizes the condition of human capital, captures the theoretical relevance of the idea.

The research modified the Musa, Magaji, Eke & Yakeen [13] model. As a result, it models poverty (POVT) as a function of the unemployment rate (UNEMT), inflation rate (INFT) and Human Development Index (HDI). Equations [3.1] and [3.2] provide the functional and econometric forms of the model, respectively;

This is shown in econometrics format:

Therefore, this study modified its model as

POVR = … (3.3)

Where;

POVR = Poverty Rate

INQ = Income Inequality

UNEMP = Unemployment

INFL = Inflation

= Coefficient of the variables

ut = Error term

**3.3 A Priori Expectation**

Economic a-priori, which is used to explain the sign and size of the parameters in the model and as well as explain the movement of variables (independent and dependent variables) in the models, will be checked to determine whether they conform to economic theory.

The a-priori expectations are: <0; < 0; > 0.

**3.3 Estimation Technique**

**3.3.1 Stationarity test**

To test for stationarity, the unit-root method will be used and will take the form of an autoregressive model (ar process), with each variable regressed on its own lagged value without an intercept and a deterministic trend. To correct autocorrelation in the error term, the ADF unit root test will be applied. The model used is:

Δyt = β1 + β2 t + δyt−1 + ∑mi=1 αi δyt−i + εt (3.4)

Δ=ρ-1

Where; y represents all the variables under consideration, δ represents the coefficient of the lagged value of y, δ is the first difference operator, yt-i represents the lagged terms included, and εt represents the pure white noise error term.

**3.3.2 Co-integration test**

ARDL bound test of co-integration by Pesaran (2001) was carried out. The ARDL approach to co-integration is recommended in cases where the variables have different orders of integration that is i(0) and i(1).

𝛥𝐻𝑂𝑅𝑡 = 𝛼0 + ∑𝑚𝑗=1 𝛼1𝑗𝛥𝐻𝑂𝑅𝑡−𝑗 + ∑𝑚𝑗=1 𝛼2𝑗𝛥𝐼𝑁𝐹𝑡−𝑗 + ∑𝑚𝑗=1 𝛼3𝑗𝛥𝐻𝑃𝐼𝑡−𝑗 + ∑𝑚𝑗=1 𝛼4𝑗𝛥𝑃𝐶𝐼𝑡−𝑗 + ∑𝑚𝑗=1 𝛼5𝑗𝛥𝑃𝐻𝐸𝑡−𝑗 + 𝜃1𝐻𝑂𝑅𝑡−1 + 𝜃2𝐼𝑁𝐹𝑡−1 + 𝜃3𝐻𝑃𝐼𝑡−1 + 𝜃4𝑃𝐶𝐼𝑡−1 + 𝜃5𝑃𝐻𝐸𝑡−1 + ---------- (3.5)

Testing the co-integration relationship is based on the f-statistic. Since the asymptotic distribution of this f-statistic is non-standard irrespective of whether the variables are *I (0)*or *i(1)*, Narayan (2005) tabulated two sets of critical values which are appropriate for the studies with small sample sizes ranging from 30 to 80 observations. In this "sense, one set assumes that all variables are *i(0)* and the other set assumes that all variables are *i(1)*. This provides a bond covering all possible classifications of the variables. If the calculated f-statistic lies above the upper level of the bound, the h0 is rejected, supporting the co-integration relationship. If the calculated f-statistic lies below the lower level of the bound, then the h0 cannot be rejected, indicating lack of co-integration” Narayan (2005).

Given the rationale provided above, the ARDL model employed in this work is as follows:

Where is intercepted,

T is the time dimension while

Δ is the difference operator and

µ is the error term.

The long-run co-integration is estimated using equation [3.6];

Based on the automatic lag length selection, the ARDL maximum lag (p q) is chosen. The study used the long-run estimate from equation [3.7] to estimate the short-run dynamic parameter using the error correction model (ECM).

In equation [3.7], are short-run dynamic coefficients convergent to long-run equilibrium, and is an error correction model's speed of adjustment parameter derived from the predicted equilibrium relationship.

**3.3.3 The bound test**

To evaluate if there is a long-term link between the variables, the bound test typically models the ARDL equation using the least squares method. The f-statistics test is then undertaken to determine the joint significance of the coefficient of lagged variables. H0:10 is in opposition to the alternative, h1: 10. The crucial value is examined in comparison to the derived f-statistics.

## 4. RESULTS AND DISCUSSION

**4.1 Descriptive Statistics**

These statistics are used to summarize the samples and observations that serve as the foundation for the description of the data set, including measures of central tendency (mean, median, and mode); measures of variability (standard deviation, variance); and the minimum and maximum values of variables (kurtosis and skewness).

The descriptive statistics in Table 1 above provided basic information about the variables to be estimated on their potential relationship. The mean which measures the average values of the data is 51.53268, 61.70927, 302.1915 and 3.336585 for unemployment (UNEMP), inflation (INFL), inequality (INQ) and the poverty rate (POVR) respectively. The median which is the middle score of the data set of the variables is 46.820000, 63.10000, 7.490000, 9.460000 and 3.600000 for the five variables respectively. The values of kurtosis of most variables are less than 3 while some are greater than 3, which means that the distribution is both Platykurtic and Leptokurtic. The Jarque-Bera statistic shows that the variables are normally distributed since the values of the variables are far from zero; while the skewness statistic showed that some variables are positively skewed, except poverty which is negatively skewed. Though, the outcome of the normality checks on the data set is inconsistent with the expected outcomes. However, this does not undermine the reliability of the analysis of the data set in economic decision-making and forecasting since this is just the means to the end and not the end itself.

**Table 1. Descriptive statistics**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
|  | **UNEMP** | **POVR** | **INQ** | **INFL** |
| Mean | 51.53268 | 61.70927 | 8.293659 | 3.336585 |
| Median | 46.82000 | 63.10000 | 7.490000 | 3.600000 |
| maximum | 86.00000 | 71.30000 | 27.04000 | 33.70000 |
| minimum | 43.00000 | 52.99000 | 0.130000 | -13.10000 |
| std. Dev. | 13.36866 | 6.012061 | 8.225641 | 7.199367 |
| skewness | 1.833326 | -0.125972 | 0.643629 | 1.320395 |
| kurtosis | 4.689072 | 1.676194 | 2.166594 | 9.594182 |
| jarque-bera | 27.84121 | 3.102228 | 4.017313 | 86.19739 |
| probability | 0.000001 | 0.212012 | 0.134169 | 0.000000 |
| Sum | 2112.840 | 2530.080 | 340.0400 | 136.8000 |
| sum sq. Dev. | 7148.843 | 1445.795 | 2706.447 | 2073.235 |
| Observation | 36 | 36 | 36 | 36 |

*Source: Author’s Computation using E-view 10, (2024)*

**Table 2. Unit root test result**

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
|  | **ADF** | | | **Philip-Perron** | | |
| **Variable** | **Level** | **1st difference** | **Order of integration** | **Level** | **1st difference** | **Order of integration** |
| UNEMP | -2.657570\* | - | I(0) | -2.635213\* | - | I(0) |
| INFL | -5.168269\*\*\* | - | I(0) | -5.161857\*\*\* | - | I(0) |
| POVR | -1.410505 | -5.846427 | I(i) | -1.681904 | -5.866561 | I(i) |
| INQ | -1.311675 | -7.196952 | I(1) | -1.226905 | -7.213212 | I(1) |

The unit root is conducted to ascertain the stationarity of the data used to prevent our work against spurious regression. Both Augmented Dicky Fuller (ADF) and Philip-Perron (PP) tests were conducted to be very sure of the stationarity of the data. The ADF result showed that unemployment (UNEMP) and inflation rate (INFL) are stationary at a level i(0); while poverty reduction (POVR) and inequality rate (INQ) are all stationary at the first difference i(i). In the same vein, the PP shows that two variables (POVR, and INQ) are stationary at the first difference (i(i); while two variables (INFL and UNEMP) are stationary at a level i(0). This means that we have an admixture of i(0) and i(i).

Table 3 showed the bound Cointegration test result to know if there is a long-run relationship among the variables.

The bound cointegration test result in Table 3 shows that the f-statistic (4.902071) is greater than the lower i(0) as well as the upper bound i(i) at a 5% level of significance meaning that the null hypothesis of no cointegration is rejected. This is an indication that there is long-run cointegration. That means that there exists a long-run relationship among the variables.

**Table 3. Cointegration test result**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **F-bound test** | | **Null hypothesis: no levels of relationship** | | |
| **Test of statistic** | **Value** | **Signif. Level (n>35)** | **I(0)** | **I(i)** |
| F-statistic | 4.902071 | 10% | 2.696 | 3.898 |
| K | 4 | 5% | 3.276 | 4.630 |
|  |  | 1% | 4.590 | 6.368 |

*Source: Author’s Computation Using E-view 10, (2024)*

**Table 4. Result estimation**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| ARDL error correction regression | | | |  |
| Dependent variable: d(POVR) | | |  |  |
| Selected model: ARDL(4, 3, 2, 3, 3) | | | |  |
| Case 3: unrestricted constant and no trend | | | |  |
| Date: 04/16/23 Time: 10:47 | | |  |  |
| Sample: 1980 2021 | | |  |  |
| Included observations: 41 | | |  |  |
| ECM Regression | | | | |
| Case 3: Unrestricted Constant and no Trend | | | | |
| **Variable** | **Coefficient** | **Std. Error** | **T-statistic** | **Prob.** |
| C | -84.86843 | 14.94118 | -5.680168 | 0.0001 |
| D(POVR) | 0.450157 | 0.510779 | 0.881314 | 0.3942 |
| D(POVR(-1)) | -1.073818 | 0.515711 | -2.082210 | 0.0576 |
| D(POVR(-2)) | 1.828434 | 0.483340 | 3.782918 | 0.0023 |
| D(UNEMP(-1)) | -0.409705 | 0.150781 | -2.717221 | 0.0176 |
| D(EMPG(-2)) | 1.043869 | 0.282757 | 3.691756 | 0.0027 |
| D(UNEMP(-3)) | -0.415720 | 0.211412 | -1.966396 | 0.0710 |
| D(INQ) | 0.896643 | 0.338269 | 2.650680 | 0.0200 |
| D(INQ(-1)) | 3.349744 | 0.799710 | 4.188696 | 0.0011 |
| D(INFL) | 0.896847 | 0.221406 | 4.050694 | 0.0014 |
| D(INFL(-1)) | -1.131064 | 0.244261 | -4.630549 | 0.0005 |
| D(INFL(-2)) | -0.752560 | 0.168948 | -4.454382 | 0.0006 |
| Cointeq(-1)\* | -0.165567 | 0.029245 | -5.661449 | 0.0001 |
| R-squared | 0.768909 | mean dependent var | | -0.385758 |
| Adjusted R-Squared | 0.565006 | s.d. dependent var | | 7.552471 |
| S.E. of Regression | 4.981164 | Akaike info criterion | | 6.355607 |
| Sum Squared Resid | 421.8040 | Schwarz criterion | | 7.081187 |
| Log-Likelihood | -88.86752 | Hannan-Quinn criteria. | | 6.599743 |
| F-Statistic | 3.770945 | Durbin-Watson stat | | 2.321042 |
| Prob(F-Statistic) | 0.005155 |  |  |  |

*\* p-value incompatible with t-bounds distribution*

Having undergone the representation and conducting the Wald test, the result is reduced to the table below:

**Table 5. Summary of the ARDL Result**

|  |  |  |  |
| --- | --- | --- | --- |
| ECM regression  Case 3: unrestricted constant and no trend | | | |
| **Variable** | **Coefficient** | **S.E** | **P-Value** |
| UNEMP | 0.834433 | 0.83443 | 0.0004 |
| POVR T-1 | 0.977535 | 0.604112 | 0.1296 |
| INQ | -1.487908 | 0.612780 | 0.0304 |
| INFL | 1.503575 | 0.715992 | 0.0558 |
| ECM(-1) | -0.165567 | 0.029245 | 0.0001 |

Table 5 above shows the calculated bound cointegration test where f lies above the upper level of the bound critical value (4.90 > 4.630) and is also greater than the lower bound of 3.276 at a 5% level of significance. This implies that there is long-run cointegration among the variables. However, the coefficient of 0.165567 ECM result, in the long run, shows an average of 17% speed of adjustment back from the short-run to the long-run equilibrium level. The positive nexus that exists between poverty and unemployment in the result is that; a unit increase in unemployment leads to a 98% increase in the poverty rate. This result complied with previous research, with a positive and significant relationship between SMEs, employment generation, poverty reduction and living standards in Nigeria.

In previous researches, an increase in employees' earnings will boost their purchasing power and hence their standard of living, thereby reducing the rate of poverty in Nigeria. Moreover, a unit change in INFL rate will lead to about 150% increase in poverty in Nigeria. This result is reflected in the Nigerian economy with double-digit inflation and a high rate of poverty in 2021 of 33.3% where there is a high decline in GDP growth (NBS, 2021). Finally, the result shows that income inequality has a negative and significant relationship with the poverty rate. This means that a unit change in inequality leads to about a 148% decline in the poverty rate.

**4.2 Causality Test**

This study aims to determine the impact of income inequality on the poverty level in Nigeria. In other words, is it income inequality that causes poverty in Nigeria? To do this, the Granger Causality Test was carried out between income inequality and poverty level in Nigeria. The expectation was that using Granger Causality Test, variable under consideration does not Granger Cause the other. The results of the Granger Causality is reported in Table 6.

**Please note:** the Granger causality results presented above are designed only to show the causality results between the explanatory and endogenous variables.

The results in Table 6 show the Granger causality test between income and poverty rate in Nigeria. It is instructive to point out here that the cointegration test carried out earlier indicates the existence of a long-run relationship between variables but says nothing about the direction of the causal relationship. Execution of the Granger causality test makes it possible for us to determine the direction of the Granger causality. In the Granger causality approach, causality exists if the f-statistic is statistically significant given its associated probability value. Thus, in this study, causality is established up to a 5% level.

The results reported in Table 6 revealed the absence of bi-directional causality between income inequality and poverty rate. This means that causality does not run between the main researcher's variables and vice-versa.

**4.3 Serial Correlation Test**

The Lagrange multiple (lm) tests is a diagnostics test used to identify the presence or absence of serial correlation to prevent the model from spurious results.

**Table 6. Granger causality results**

|  |  |  |  |
| --- | --- | --- | --- |
| Pairwise Granger causality tests | | | |
| Date: 04/16/23 Time: 16:53 | | | |
| Sample: 1980 2021 | | |  |
| Lags: 2 | |  |  |
| **Granger Causality** | **Obs** | **F-statistic** | **Prob.** |
| POVR does not granger cause INFL | 37 | 0.18190 | 0.8345 |
| INFL does not granger cause POVR | | 0.46263 | 0.6338 |
| UNEMP does not granger cause INFL | 37 | 1.95737 | 0.1578 |
| INFL does not granger cause UNEMP | | 1.13967 | 0.3326 |
| INQ does not granger cause POVR | 37 | 0.26232 | 0.7709 |
| POVR does not granger cause INQ | | 0.19119 | 0.8269 |
| UNEMP does not granger cause INQ | 37 | 0.14230 | 0.8679 |
| INQ does not granger cause UNEMP | | 1.15954 | 0.3265 |

*Nb: \* means rejection of the null hypothesis of non-granger causality.*

*Source: Author’s Computation, E-views, 10.0, 2024*

**Table 7. Breusch-godfrey serial correlation lm test**

|  |  |  |  |
| --- | --- | --- | --- |
| **Breusch-Godfrey Serial Correlation lm Test:** | | |  |
| F-statistic | 0.637390 | prob. F(2, 20) | 0.5391 |
| Obs\*r-squared | 1.617833 | prob. Chi-square(2) | 0.4452 |

*Source: Authors Computation Using E-views, 10.0, 2024*

Table 7 shows that the f-statistic and obs\*r-square values of 0.64 and 1.62 with p-values of 0.54 and 0.45 respectively indicate the absence of autocorrelation in the model since the f-statistic and obs\*r-square with p-values of 0.54 and 0.45 are greater than the critical values at 5% level of significance. Thus, we can conclude that there is no presence of autocorrelation in the model.

**4.4 Heteroscedasticity Test**

This test is directed to check whether the variability of error terms is steady or not. The presence of heteroscedasticity signifies the instability of the residuals and could affect the inferences.

Table 8 below, shows that the f-statistic and obs\*r-square values of 1.18 and 4.77 with p-values of 0.35 and 0.31 respectively indicate the absence of heteroskedasticity in the model since the f-statistic and obs\*r-square with p-values of 0.35 and 0.31 are greater than the critical values at 1% level of significance. Thus, we can conclude that there is no presence of heteroskedasticity in the model.

**4.5 Ramsey Reset Test**

In statistics, the Ramsey regression equation specification error test (reset) is a general specification test for the linear regression model. More specifically, it tests whether non-linear combinations for the fitted values help explain the response variables.

Table 9 above, shows that the f-statistic of 4.409036 with p-values of 0.0504 indicates that the model is correctly specified since the f-statistic with p-values of 0.0504 is greater than the critical values at p > 0.05 level of significance. We can conclude that there is no misspecification of the model and thus, the model was correctly specified.

**Table 8. Heteroskedasticity test: Breusch-pagan-godfrey**

|  |  |  |  |
| --- | --- | --- | --- |
| F-statistic | 1.181472 | prob. F(4,22) | 0.3463 |
| Obs\*r-squared | 4.774357 | prob. Chi-square(4) | 0.3112 |
| Scaled Explained ss | 2.398449 | prob. Chi-square(4) | 0.6629 |

*Source: Authors Computation Using E-views, 10.0, 2024*

**Table 9. Ramsey reset test**

|  |  |  |  |
| --- | --- | --- | --- |
| Equation: untitled | | |  |
| Specification: POVR UNEMP POVTT-1 INQ INFL | | | |
| Omitted variables: squares of fitted values | | | |
|  | **Value** | **Df** | **Probability** |
| T-statistic | 2.075822 | 21 | 0.0504 |
| F-statistic | 4.309036 | (1, 21) | 0.0504 |
| Likelihood ratio | 5.039254 | 1 | 0.0248 |

*Source: Authors Computation Using E-views, 10.0, 2024*

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**Fig. 1. Cumulative Sum (Cusum)**

*Source: Authors Computation Using E-views, 10.0, 2023*



**Fig. 2. cumulative sum of squares (Cusumsq)**

*Source: Authors Computation Using E-views, 10, 2024*

**4.6 Stability Tests**

The cumulative sum (cusum) and cumulative sum of squares (cusumsq) are used to test the stability of the long-run coefficients alone with the short-run dynamics.

In Figs. 1 and 2 above, Cusum and Cusumsq are graphically shown. The statistics demonstrate the stability of the long-run coefficients of the repressors that affect inclusive growth in Nigeria because the plots of both the Cusum and the Cusumsq are within the borders. Given that none of the two test statistics deviate from the limits of the 5% level of significance, the model looks to be stable and well-described.

## 5. CONCLUSION

Sequel to the results and the findings in this study, the following logical, coherent and sequential conclusions are made:

There is a long-run positive relationship between the employment rate and the Gini coefficient of income inequality in Nigeria. That is, the employment rate increases income inequality in Nigeria. There is a long-run positive relationship between the inflation rate and the Gini coefficient of income inequality in Nigeria, since, the inflation rate widens the gap between the rich and the poor. Also, the inflation rate reduces the purchasing power of income of the junior workers thereby widening the income gap in Nigeria. There is a long-run, inverse relationship between inequality and the Gini coefficient of income inequality in Nigeria. Inequality of income inequality increases the rate of poverty. This is premised on the fact that a large portion of income is concentrated in a few hands (about 3%) which makes the poverty rate high. The study also concluded that employment rate, inflation rate, and inequality are true determinants of income inequality in Nigeria. A negative coefficient of the over-parameterized ECM which shows that changes in the Gini coefficient of income inequality depend on changes in all the variables and also on the equilibrium error term further justifies the existence of the long-run relationship between income inequality and its determinants.

From the empirical findings in this research work and based on the relationship each determinant of income exhibited with the Gini coefficient of income inequality in Nigeria's economy, the following recommendations are made to enhance a more evenly distribution of income which would in effect, reduce income gap and poverty in Nigeria. Efforts of the government should be mobilized towards the formulation and implementation of more pragmatic employment policies in Nigeria. Since the empirical findings in this research work have shown that a rise in the employment rate has not been sufficient to reduce inequality of income inequality in Nigeria. A more pragmatic employment policy would enable workers to create wealth from their income (and not just for sustenance) which enhances a more evenly distributed income. If the efforts of the Nigerian government on wage increase are going to yield more positive results, then, a policy to reduce the current inflation rate of 2-digits to 1-digit must be prioritized. A considerable rate of inflation would help increase the real value of income in the hands of workers especially the low-income earners thereby closing the gap between the rich and the poor. Our findings in this empirical work indicated that a rise in the inflation rate widens the income gap in Nigeria, therefore, simultaneous policies for wage increase and that which controls the growth rate of inflation should be given more priority in Nigeria. In addition to higher growth rates, development policies that address severe and pervasive income inequality must be implemented to lift people out of poverty. Examples of such policies include better distribution of human capital, well-targeted social protection, better distribution of socio-economic amenities like roads, electricity, schools, and hospitals, more progressive taxation, and implementation of the increased minimum wage and a set of policies designed to bring about a more equitable distribution of income, equal access to education and associated income earning opportunities should be given priority in Nigeria.

## REFERENCES

1. World Bank Report. Attacking poverty published by the World Bank Oxford University Press; 2019.
2. Magaji S. Towards a gender model of poverty alleviation in Sub-Sharan Africa. Journal of Research and Development in Africa. 2002;1(2):81-85.
3. OECD. Better policies and better lives, the global recovery is strong and imbalanced; 2021.
4. Silver H. Social exclusion and social solidarity: three paradigms. International labour review. 2013;133(5-6):531-578.
5. El Agri F, Jerry M, Qafas A, Manzah Y. Determinants of the export function in Morocco: Evidence from ARDL and dynamic ARDL models. Global Journal of Emerging Market Economies. 2023;09749101231161231.
6. Aluko OO, Magaji S. Stagflation and poverty incidence in West Africa Sub-Region: A perspective journal of advanced research in social sciences, environmental studies and technology. 2020;5(1):38-59.
7. Shaba NY, Obansa SAJ, Magaji S, Yelwa M. Analysis of the relationship between income inequality and poverty prevalence in selected North Central States of Nigeria: 1991-2013. Applied Economics and Finance. 2018;5(3):22-33.
8. Ostry JD, Berg A, Tsangarides CG. Redistribution, Inequality and Growth, IMF Staff Discussion Note; 2014.
9. Magaji S, Musa I. Endemic Corruption and Nigeria’s underdevelopment, Abuja Journal of Business and Management. 2015;1(4):119–125.
10. Magaji S, Musa I. Impact of household income on child labour in Abuja Municipal Area Council (AMAC). International Journal of Management and Social Science. 2015;8(1):32–56.
11. Oxfam. A tale of two continents: fighting inequality in Africa. Cowley, Oxford, U.K: author; 2019.
12. Chancel I, Cogneau D, Gethin A, Myczkowski A. How large are African inequalities? Towards distributional national accounts in Africa. Wid. World working paper, no. 2019/13; 2019.
13. Musa I, Magaji S, Tsauni A. Analysis of the basic infrastructure affecting child labour in Northeastern Nigeria, Applied Journal of Economics, Management and Social Sciences. 2022;3(4):13–22.
14. Ibrahim M, Sule M. Nexus between household income and child labor in Northeastern Nigeria. African Journal of Social Sciences and Humanities Research. 2023;6(3):57-70.

DOI: 10.52589/AJSSHRYLWQGNJ2

1. Oxfam. Inequality in Nigeria: exploring the drivers; 2017.

Available:<https://www>[-cdn.oxfam.org/s3fspublic/file\_attachments/cr-inequality-in-nigeria-170517-en.pdf](https://www-cdn.oxfam.org/s3fs-public/file_attachments/cr-inequality-in-nigeria-170517-en.pdf)

1. Yunqian AC. The hidden costs of a successful developmental state: prosperity and paucity in Singapore (Master’s thesis). Department of international development, London School of Economics and political science Houghton Street, London; 2017.
2. Azam M, Uddin I, Saqib N. The determinants of life expectancy and environmental degradation in Pakistan: Evidence from ARDL bounds test approach. Environmental Science and Pollution Research. 2023;30(1):2233-2246.
3. Kopp CM. Income inequality; 2019.

Available:<https://www.investopedia.com/terms/i/income-inequality.asp>

1. Musa I, Magaji S. Analysis of the impact of household income on child labour in Bauchi State, Nigeria, Journal of Arid Zone Economy. 2023;1(1):130-141.
2. Adam AJ, Magaji S, Ayo AA, Musa I. The impact of domestic debt on economic performance in Nigeria (1970 – 2013), Journal of Economics and Sustainable Development. 2016;7(8):54-64.
3. Magaji S, Adamu AM. Youth employment in Nigeria since independence, Nigeria at Fifty: Issues and Challenges in Governance, University of Abuja. 2011;246–260.
4. Magaji S, Abubakar MM, Temitope YA. Impact of international trade on economic growth: The granger causality test approach. International Journal of Accounting, Management and Science. 2022;1(2):113–130.
5. Chen D, James U. [Structural basis for substrate selection by the SARS-CoV-2 replicate](https://www.nature.com/articles/s41586-022-05664-3). 2023;781–787.

Available:https://doi.org/10.1038/s41586-022-05664-3

1. Alayande B. Decomposition of Inequality Reconsidered: some evidence from Nigeria: paper presented at the unu/wider conference on inequality, poverty and human well-being in Helsinki, Finland between 29th and 31st of May, 2021; 2020.
2. Adesimi AA. Economic analysis of small scale Tamba, Global Scientific Journal. 2020;8(8):1168-1179.
3. Kennedy O. Income inequality and income inequality in Nigeria: bridging the gap through trader money (the Buhari model), issues, challenges and prospects. International Journal of Social Science. 2019;3(15):100-121.
4. Chen-Wiegart YCK, Catalano J, Williams GJ, Murphy A, Yao Y, Zumbulyadis N, Centeno SA, Dybowski C, Thieme J. Elemental and molecular segregation in oil paintings due to lead soap degradation. Scientific Reports. 2017;7(1):[11656].

Available:https://doi.org/10.1038/s41598-017-11525-1

1. Emmanuel IJ, Willie EE. Small and Medium Scale Enterprises (SMEs) and Economic Growth in Nigeria. IAR J Bus Mng. 2020;2(1):38-43.
2. Jehoel G, Vrooman I. Explaining social exclusion; A theoretical model tested in the Netherlands, The Netherlands Institute for Social Research; 2017.
3. Musa I, Magaji S, Eke CI, Yakeen OA. Poverty and its interactivity: Causes and consequences, Inclusive Society and Sustainability Studies (ISSUES). 2022;2(2):48-58.
4. Oguntuase A. An empirical investigation into the determinants of income inequality in the Nigerian manufacturing sector (unpublished M.Sc. thesis). Ekiti State University; 2017.
5. Pierre-Richard A, Koray A, Luiz A, Pereira da S. [capital regulation, monetary policy, and financial stability](https://econpapers.repec.org/article/ijcijcjou/y_3a2013_3aq_3a3_3aa_3a6.htm). International