

**Effects of income inequality on COVID-19 in Africa:  
Accounting for literacy and informal sector**

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**Abstract** - The empirical literature has explored little regarding the transmission channels in the study of the relationship between income inequality and the spread of COVID-19. The aim of this article is to fill this gap by assessing the direct and indirect effects of income inequality on the spread of the pandemic using a sample of 43 African countries during the period from 2020 to 2022. Two indicators of COVID-19 are utilized: the total number of cases and the severity of the disease and explanatory variables, such as urban population and population density, play a significant role. Two transmission channels are particularly examined: the size of the informal sector and literacy rates. The results, derived from a fixed-effects regression model or estimated using generalized least squares, indicate that income inequality has a significant effect on the spread of COVID-19. However, it is shown that these results underestimate the effects of inequality when considering the indirect effects due to the weight of the informal sector and illiteracy.

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**JEL Classification**

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## INTRODUCTION

The recent COVID-19 pandemic is arguably the most significant event of the second decade of the 19th century (Brunckhorst et al., 2024; Galanis and Georgiadis, 2024; Hangoma et al., 2024). She affected all social strata around the world, from adults (Lopes and Middleton, 2020) to children (Lopes et al., 2021). It has not only distinguished itself by disrupting the global health system, but also by disrupting both economic and sociological ecosystems (Ginsburgh et al., 2021). It has affected both developed and developing countries (Amate-Fortes and Guarnido-Rueda, 2022). The pandemic has led to containment and social distancing measures that have dealt a severe blow to the global economy (Bargain and Aminjonov, 2021; Henao-Cespedes, 2022).

Analyzing the annual data, the GDP of all developing countries is expected to remain below the pre-pandemic trend. This means a fall in economic activity of 2.8% in 2020 in a context of high uncertainty (World Bank, 2020). Some regions are facing output gaps that are significantly larger than their pre-pandemic performance. In the specific case of sub-Saharan Africa (SSA), for example, the gap in 2023 compared to the pre-pandemic trend is expected to be over 4% (World Bank, 2022). Furthermore, the situation of income inequality in Africa is complex. Seven of the ten most unequal countries in the world are located in Africa, mainly in Southern Africa (World Bank 2020). Oxfam's (2021) study shows that inequality could be on the increase worldwide. In the case of West Africa, the wealth of the region's three richest men rose from \$16.8 billion in March 2020 to \$23.2 billion in July 2021. In addition, these inequalities undermine countries' economic growth, reduce the impact of policies (Kuznets, 1955; Stiglitz, 2012), and undermine the achievement of the Sustainable Development Goals (World Bank, 2020). Their negative effects on economic performance are associated with increased health shocks, terrorism, political instability and crime leading to a mistrust climate (Wagstaff and Lindelow, 2014; Wilkinson and Pickett, 2011).

COVID-19 creates a situation where the supply and demand of goods and services temporarily stopped, bringing countries to the brink of economic recession (Henao-Cespedes, 2022). To better understand the evolution of the pandemic and the effects on aggregate supply and demand, it is important to look at the mechanisms by which the pandemic affects the economy and each other. In addition to the reduction in consumption of goods and services due to the prolonged duration of the pandemic and the social distancing measures that reduce spending and therefore household consumption (Dandonougbo et al., 2021). COVID-19 creates disruptions that can occur in the labour market. This is because COVID-19 keeps production low, which hampers supply chains, labour demand, leading to prolonged periods of layoffs and increased inequality which in turn would lead to the spread of the pandemic (Ahmed et al., 2020; Casarico and Lattanzio, 2022).

In addition, Acemoglu et al. (2020) show in their work that containment harms the economy and reduces the productivity of uninfected members. However, due to income inequalities within social strata, individuals from poor backgrounds do not always respect confinement measures. This in order to be able to satisfy their physiological needs such as food (Stiegler and Bouchard, 2020). Failure to comply with these containment and social distancing measures therefore encourages the COVID-19 spread, particularly in developing economies (Bargain and Aminjonov, 2021). The pandemic did not spread with the same intensity within different

countries and/or economies (Diop et al., 2020). Moreover, it could not be brought under control within these economies within the same timeframe, which more or less caused the severity of the pandemic to severely hamper the development of the economies (Pearce et al., 2020). This has led to a continuing debate about the causes of the spread of the disease. In response to this concern, the existing literature has focused on the determinants of the evolution of COVID-19 around the world.

In this vein, investigating the causes of the COVID-19 spread in Africa would enable the implementation of appropriate economic policies to address the major challenges they face, including the achievement of the Sustainable Development Goals and the post COVID-19 recovery that most African economies are facing (Barbier and Burgess, 2020; Roe et al., 2020; Leach et al., 2021). Thus, a review of the existing literature on the determinants of COVID-19 highlights several contributions from different disciplines that have proposed reasonable answers to the question of the spread of the disease worldwide (Diop et al., 2020; Coccia, 2021; Tavares and Betti, 2021; Bontempi and Coccia, 2021).

Specifically, several axes emerge, such as the role of direct transmission occurring from human to human through contact or proximity in the first instance (Chang et al., 2021). Secondly, demographic parameters ranging from population density in relation to socio-economic activities, age, gender, and urbanization (Bertuzzo et al., 2020; Copiello and Grillenzoni, 2020; Ehlert, 2021; Garenne, 2020). Other studies suggest the axis of environmental and climatic factors, such as pollution, humidity, wind speed and temperature associated with the transmission dynamics of COVID-19 (Copat et al., 2020; Ahmadi et al., 2020; Şahin, 2020). Subsequently, health factors, particularly through health inequalities, health expenditure and medical density were discussed (Okoi and Bwawa, 2020; Garenne, 2020). Finally, the last axis explored in the literature on the determinants of the spread and COVID-19 is that of economic factors such as GDP per capita, the level of tourist travel, the employment rate, poverty, inequalities and air transport (Diop et al., 2020; Banik et al., 2020; Farzanegan et al., 2021; Davies, 2021; World Bank, 2022; Garenne, 2020).

In this respect, the diagnosis of a careful reading of these and related studies in the light of developments in the literature on the explanatory factors of the COVID-19 spread recommends further investigation for at least two reasons. The first concerns economic factors, and more specifically income inequalities. Indeed, although the literature has not yet reached a consensus on the possibility of a proven effect of income inequalities on pandemics (Davies, 2021), they have been very weakly associated with COVID-19, particularly in developing economies and specifically in Africa where these inequalities persist (Leach et al., 2021; Wildman, 2021). To the best of our knowledge, Wildman's (2021) study is the only one to explicitly address the effect of income inequality on COVID-19 within Organization for Economic Co-operation and Development (OECD) member countries. To this end, it demonstrates a positive and significant association between income inequality and COVID-19 deaths. This is justified by the fact that individuals facing income inequality face many socio-economic disadvantages that can contribute to COVID-19 spread and deaths associated with the disease. However, this work is limited to a direct analysis of this relationship, which could raise questions as the simultaneous consideration of direct and indirect effects would allow us to obtain an overall effect that best reflects the reality of the evolution of the relationship studied (Papyrakis and Gerlagh, 2004).

The second reason refers to the empirical assessment of this relationship, which to the best of our knowledge has not been sufficiently tested empirically, and even less in the context of African economies. Above all, the approach of the channels through which inequalities affect the COVID-19 spread have not yet been explored in the literature. More specifically, the channels of the informal economy and literacy have been left out of empirical assessments of the relationship between income inequality and COVID-19. With regard to the informal sector, studies show that formal sector is associated with the payment of taxes (Besley et al., 2012), something that seems unlikely for people without high incomes. Therefore, income inequality is associated with an increase in the informal sector. Similarly, the conditions in which informal sector activities take place, including crowded markets and congested streets, favour contact between people and therefore the COVID-19 spread (Nguimkeu and Okou, 2022). The rationale for the literacy channel is that income inequality hinders the ability to prevent and treat COVID-19 among the poor who are most vulnerable due to their high illiteracy rates, thus confounding the safety, eradication and control of the spread of the pandemic (Nanda, 2020).

This study therefore proposes to fill this gap in the existing literature by examining the question of the direct and indirect effects of income inequality on the COVID-19 spread in a sample of 43 African countries. Specifically, this study extends the existing literature on COVID-19 by examining how income inequality affects the COVID-19 spread. Although the literature analysing the effect of income inequality on the COVID-19 spread is very scarce, the relationship between the two concepts can be established through two channels, namely the size of informal sector and literacy. In this respect, apart from analysing the direct link between these variables, this study focuses particular emphasis on the role played by the literacy and the size of informal sector in explaining this relationship. Thus, we propose one of the first empirical assessments of transmission channels in the analysis of the relationship between income inequality and COVID-19 spread.

In order to analyze transmission channels, we refer to the work of Papyrakis and Gerlagh (2004) and Adams and Fotio (2022). The advantage of this method is that it makes it possible to separate the total effect from the indirect effect. Work based on interactions between variables may underestimate or overestimate the estimated coefficient due to the non-separation of these two effects (Papyrakis and Gerlagh, 2004; Adams and Fotio, 2022). Thus, the results of the direct analysis reveal that income inequality increases the COVID-19 spread in Africa. Although the indirect analysis supports these results, the coefficients estimated from the direct analysis are one the one hand underestimated (in the case of the number of COVID-19 cases) and the other hand overestimated (in the case of COVID-19 severity). The size of informal sector and literacy seems to be the channel through which income inequality affects COVID-19 spread.

The rest of the article is structured as follows: Section 1 presents the methodological approach. The results and discussions are presented in Section 2.

## **1. METHODOLOGICAL APPROACH**

### **1.1. Analysis model**

This study aims to examine the direct and the mediated effect of income inequality on COVID-19 spread on a sample of 43 Africa countries. The empirical model relates the income inequality of country  $i$  at the time  $t$  to the natural log of

COVID-19 total cases and COVID-19 severity. To choose between fixed or random effects we use Hausman test where the null hypothesis is that the preferred model is random effects versus the alternative hypothesis is the fixed effects (Upreti, 2019). Regardless of COVID-19 indicator chosen, the results of the Hausman test indicate that fixed effects model fits well to our data to explain the link between income inequality and COVID-19 spread.

$b$  = consistent under  $H_0$  and  $H_a$ ; obtained from xtreg  
 $B$  = inconsistent under  $H_a$ , efficient under  $H_0$ ; obtained from xtreg

when COVID-19 is captured by total cases:  
 $Chi^2(5) = (b - B)'[(V_b - V_B)^{-1}](b - B) = 62.28$   
 $Prob. > Chi^2 = 0.000$

when COVID-19 is captured by severity:

$Chi^2(5) = (b - B)'[(V_b - V_B)^{-1}](b - B) = 55.35$   
 $Prob. > Chi^2 = 0.000$

In either case, the  $Prob. > Chi^2$  is less than 1% level of significance and therefore, the null hypothesis (random effect is appropriate) is rejected. So, we use fixed effect estimation. Thus, the specification adopted is as follows:

$$LogY_{it} = \alpha + \beta \times Ineq_{it} + X'_{it} \times \gamma + \tau \times M'_{it} + \sigma_i + \omega_t + \varepsilon_{it} \quad (1)$$

where the subscripts  $i$  = country and  $t$  = year;  $LogY_i$  is the logarithm of the dependent variable which can be either the total number of confirmed COVID-19 cases or the severity of COVID-19;  $Ineq_i$  is our main variable of interest which captures income inequality;  $\sigma_i$  and  $\omega_t$  respectively represent country and time specific fixed effect. The reason for using fixed effect (FE) model is to control each of the 43 countries' own individual characteristics that may influence the predictor variables, income inequality. Heterogeneities across sample countries are controlled by  $\sigma_i$  and  $\omega_t$ . FE removes the effect of those time-invariant characteristics, so that the result can assess the net effect of the predictors on the outcome variables.

$X'_i$  is a vector of control variables including population density, GDP per capita, government effectiveness and urban population proxy by urban population growth. Variables such as GDP per capita and population density are considered in logarithms. The use of the logarithm in the empirical literature is justified by the reduction of skewness in data (Saadi, 2020).  $M'_i$  represents the vector of other control variables that incorporates the informality rate and the literacy rate into the model;  $\varepsilon_i$  represents the error term. Finally,  $\beta$ ,  $\gamma$  and  $\tau$  are coefficients to be estimated and  $\alpha$  the constant.

All of these identified control variables are in line with the theoretical and empirical literature (Lopes and McKay 2020; Ngumkeu and Okou 2022; Ngumkeu and Tadadjeu 2021; Koudjom et al. 2022; Wildma, 2021). Indeed, high population density is likely to increase intercommunal contamination even if social distancing measures are respected (Amate-Fortes and Guarnido-Rueda, 2022; Ngumkeu and Tadadjeu, 2021; Koudjom et al., 2022; Garenne, 2021). Thus, just like previous authors, we expect a positive relationship between population density and COVID-19

spread. In the empirical literature, there is a lack of consensus about the effects of income on disease spread. While Nguimkeu and Tadadjeu (2020) and Nguimkeu and Okou (2022) find no significant effect, other authors find a positive and significant effect on disease spread (Wildman, 2022; Koudjom et al., 2022). Thus, as in previous studies, to capture income level, we use GDP per capita (Nguimkeu and Okou, 2021; Nguimkeu and Tadadjeu, 2020). To capture institutional quality, this research considers government effectiveness. The use of such a variable is justified by all the measures taken by the government to stop the progression of the disease (Desson et al., 2020). These measures include, among other things, partial or complete lockdown, compliance with barrier measures, detection and treatment centers and social transfer policies. Like previous studies (Koudjom et al., 2022; Amate-Fortes and Guarnido-Rueda, 2022), we expect a negative relationship between the two variables. For infectious diseases such as COVID-19, some authors show that cities are the main centers of transmission (Stillwaggon, 2002; Greif et al., 2011). So, like previous studies, we expected urbanization to increase the number of COVID-19 cases.

This study takes into account the size of informal sector and the literacy rate, which can lead to an increase in the number of COVID-19 cases thanks to non-compliance with barrier measures (Koudjom et al., 2022; Nguimkeu and Tadadjeu, 2021; Lopes and McKay, 2020). Indeed, the activities of the informal sector most often take place without respecting the barrier measures enacted by political decision-makers to stop the progression of the disease. Although the authors seem unanimous about a positive relationship, very few studies have used the informal sector as a determinant of COVID-19. To our knowledge, only Nguimkeu and Okou (2022) use the size of informal sector as a determinant of disease. For these authors, the size of informal sector increases the spread of the disease. Thus, we expect a positive relationship between the two variables.

Work by Lopes and McKay (2020) shows that people who are illiterate are less open to health education and are less likely to comply with the advice and guidelines of disease prevention services. Although the role of education is considered decisive in the fight against this disease, to our knowledge, no study has included this variable as a determinant of the spread of the disease. Based on theoretical literature, we expect a positive relationship between the two variables. Thus, the extended version of equation (1) can take the following form:

$$\text{Log}Y_{it} = \alpha + \beta \times \text{Ineq}_{it} + X'_{it} \times \gamma + \tau_1 \times \text{Inf}_{it} + \tau_2 \times \text{Lit}_{it} + \sigma_i + \omega_t + \varepsilon_{it} \quad (2)$$

where  $\text{Inf}_i$  and  $\text{Lit}_i$  respectively represent the size of informal sector, i.e., the share of informal employment as a percentage of total employment and the literacy rate measured by the share of the literacy rate of people aged 15 and over. The other variables and symbols remained the same.

## 1.2. Analysis data

To achieve the objective of this work, this study uses a sample of 43 African countries observed over the period 2020-2022. Five data sources are used to collect the variables necessary for the empirical analysis: (1) the total number of confirmed COVID-19 cases and disease severity are from the World Health Organization (WHO, 2020); (2) the Gini concentration index that captures income inequality is taken from the World Bank PovcalNet Report (PovcalNet-World Bank, 2019); (3) popu-

lation density, GDP per capita, official development assistance and literacy rate are taken from World Bank Indicators (World Bank 2020); (4) the informality rate is taken from the International Labor Organization database (ILO, 2020); (5) The institutional variable (government effectiveness) is taken from Worldwide Governance Indicators database (Kaufmann et al., 2023). Detailed definitions and data sources are summarized in Table A2 in the appendix along with the list of countries used (Table A1). The descriptive statistics of the variables used for the analysis of the direct and indirect effect of income inequality on COVID-19 spread in Africa are presented in Table 1.

**Table 1. Descriptive statistics**

Variables	Observations	Mean	Std. Dev.	Min	Max
<b>Dependent variables</b>					
Total COVID-19 cases	129	143531.5	476722.3	226	4028160
COVID-19 severity	127	39.399	15.00179	7.638	70.92282
<b>Interest variables</b>					
Income inequality (%)	127	54.92898	7.495525	31.92	67.46
Informality (%)	120	80.52203	19.08803	15.06	98.62
Literacy rate (%)	128	72.9844	13.91951	27.28	96.2
<b>Control variables</b>					
GDP per capita	128	4.80 e+10	9.49 e+10	5.47 e+10	4.77 e+10
Government effectiveness	128	-0.6820138	0.6618482	-2.131399	0.8995878
Population density	120	122.2705	146.1064	3.0233	634.1182
Urban population (growth rate, %)	129	3.351867	1.220738	-0.21559	5.737879

Source: Authors.

Summary statistics for all these variables are presented in Table 1. Table 1 shows that, on average, confirmed cases and the severity of COVID-19 over the period 2020-2022 amount to 143531.5 cases and 39.339 % respectively. A careful analysis of our main variables of interest over the period 2020 to 2022 shows that in most African countries, income inequality, the informality rate and the literacy rate are respectively 54.93%, 80.52% and 72.98%. In terms of socio-economic variables, average per capita income and population density are respectively US\$4.80 e+10 and 122.2705 inhabitants/km<sup>2</sup>. Institutional quality, captured by government efficiency, averages -0.682, with a relatively high variance.

Table A4 in the Appendix shows the correlation matrix for the variables used. The correlation coefficients are low and below 0.7, leading to the absence of multicollinearity between the model's explanatory variables. The correlation coefficients between the explanatory variable and the explained variables are low, but Baltagi (2008) and Wooldridge (2010) for example show that introducing literature-guided control variables other than the variables of interest into the model can statistically and significantly improve the econometric results of the model.

### 1.3. Descriptive analyzes

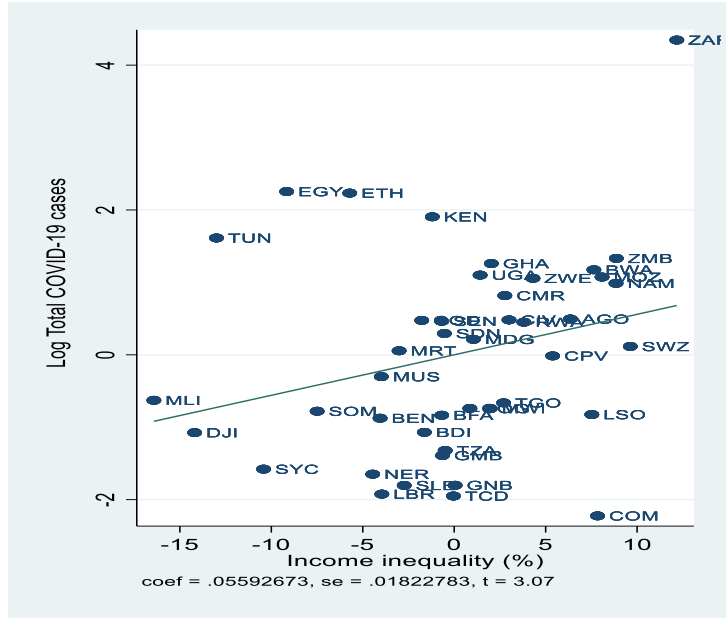
The first cases of COVID-19 were reported in most African countries in early March 2020 while the number of confirmed cases increased rapidly after 15 March 2020. As of April 10, 2020, some African countries Africa already had more than 6,000 confirmed cases. South Africa had the largest outbreak in Africa with 1,039,161 cases from March to December 2020 associated with 28,033 death cases, while Seychelles had the lowest number of confirmed cases (226) associated with 0 deaths over the same period. Considering the years 2021 (January-December) and 2022 (January-October), we find that South Africa is still recording more cases and deaths, i.e., 3,446,532 cases associated with 91,061 deaths in 2021. Also, we note in this country that the number of confirmed cases and deaths are respectively estimated at 4,028,160 and 102,311 in 2022. In addition, the statistics indicate that in 2021 it is rather Chad which experienced fewer cases (5,701) associated with 181 deaths. However, in 2022 we also see that it is rather Sao Tome and Principe which records fewer cases (6,266) associated with 77 deaths. Table A3 in the Appendix presents the multicollinearity test (variance inflation factor, VIF). The results show that  $VIF=1.51$ , this value indicates a moderate correlation between the explanatory variables of the model, but it is often not severe enough to require special attention. Furthermore, Figure 1 and 2 presents the scatter plot of the relationship between income inequality, the number of COVID-19 infections and the severity of the pandemic in Africa. It appears from these figures that the total number of cases was higher in South Africa, lower in countries such as Comores, Chad, Liberia, Guinea-Bissau, Sierra Leone, Niger, Seychelles, Gambia and Tanzania. On the other hand, the severity of the disease was lower in Burundi.

COVID-19 pandemic is an urban crisis – about 95% of total cases are in urban areas (United Nations Settlements Programme, 2020b; Smit, 2021). This can make it easier for the virus to spread quickly in areas where people live and work in close proximity to each other. Thus, South Africa is one of the most urbanized countries in Africa with around 67 % of its population living in urban areas (United Nations Settlements Programme, 2020a). Cape Town, the second largest city in South Africa, has been one of the cities most affected in Africa – it has had more than 70,000 cases (South African Government, 2020; Smit, 2021). Cities are considered the epicenter in infectious disease transmission, not only because of international travel and migration, but also because urbanization is associated with high numbers of active infected cases (Stillwaggon, 2002; Nguimkeu and Tadadjeu, 2021). World Bank data on urbanization as a percentage of total population for the least-affected countries in our sample show that over the period 2020-2022, Burundi has the lowest urbanization rates of 13.7%, 14.1% and 14.4%, for the years 2020, 2021 and 2022 respectively. Given that there is a positive relationship between urbanization and disease spread (Nguimkeu and Tadadjeu, 2020; Smit, 2021), such a situation may justify Burundi as the country with the lowest severity forms of the disease. Similarly, Figure 1 also shows that the slope of the relationship between income inequality and the total number of cases, and between income inequality and disease severity, is positive.

However, the number of cases of infection with COVID-19 has probably been underestimated in Africa due to the lack of screening capacity in many countries (WHO, 2020). Although the number of confirmed cases remains low, when Africa is compared to other continents of the world, the negative effects are still noticeable, including the contraction of economic activity leading to a drastic decline in people's livelihoods, accentuating income inequalities in Africa, especially in sub-Saharan Africa (IMF, 2020; Nguimkeu and Okou, 2022).

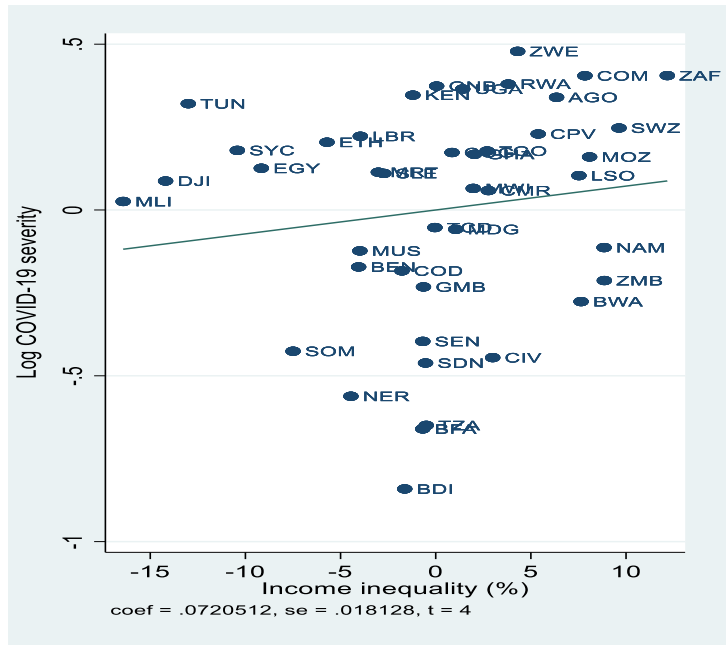


**Figure 1. Relationship between income inequality and total number of COVID-19 cases in Africa over 2020-2022**



Source: Authors.

**Figure 2. Relationship between income inequality and COVID-19 severity index in Africa over 2020-2022**



Source: Authors.

Furthermore, it should be recalled that most economies in African countries are informal. Consequently, people in these countries tend to face higher levels of informality, illiteracy and income inequality coupled with a lack of social protection (Nguimkeu and Okou, 2022; Amin and Okou, 2020) and have no replacement income or savings in case of unexpected external shocks, such as COVID-19. Therefore, social distancing measures to control the virus may be ineffective for African populations, as staying at home and not working implies the loss of income crucial to their livelihoods, potentially driving up income inequalities.

## 2. RESULTS AND DISCUSSION

### 2.1. Analysis of the direct effect of income inequality on COVID-19 spread

#### 2.1.1. Basic results

Table 2 presents the basic results. The first two columns of this table present the results where our dependent variable is the total number of cases while in the other two columns the dependent variable is the severity of the disease. In the second column of each block, we introduce the variable that captures literacy and the size of informal sector. In Table 2, all coefficients have the expected signs in terms of the direction of the relation between the independent and dependent variable. The coefficient of determination  $R^2$ , in all equations is 61% or more, decisively rejects the hypothesis of joint non-significance of the independent variables. With respect to our variable of interest, namely income inequality, the results show that it positively and significantly affects COVID-19 at the 10% (columns 3) and 5% (column 1, 2 and 4) thresholds. Therefore, all other things being equal, a one-percentage-point increase in income inequality is associated with 51.3% (column 1) and 77.5% (column 2), increase in the total number of COVID-19 cases. Similarly, a one-percentage-point increase in income inequality is associated with a 87.8% (column 3) and 65.7% (column 4) increase in disease severity. These results are consistent with those obtained in previous work. The work of Koudjom et al. (2022) shows that income inequality has a positive and significant effect on the total number of cases and the severity of COVID-19. Similarly, the work of Wildman (2021) shows that income inequality jointly increases the total number of cases and the number of deaths due to COVID-19 in OECD countries. The positive effect of income inequalities can be explained by the fact that the most disadvantaged social classes can not respect the barrier measures to limit the spread of the disease. Despite state social transfers, the disadvantaged classes are forced to continue going about their business, since they live on daily income (Cartlitz and Makhura, 2021; Durizzo et al., 2021; Bargain, 2021). In the specific case of Africa, the social transfers put in place to support populations were not sufficient to reduce the income gap between social classes (Koudjom et al., 2022). Consequently, populations with low incomes are forced to continue going about their business, thus promoting the spread of COVID-19.

As far as our control variables are concerned, the results are generally satisfactory. Table 2 shows that population density, urban population and the size of informal sector have a positive and significant effect on the spread of the disease, government effectiveness and literacy have a negative and significant effect. Regarding the variables with a positive effect, the positive effect of population density is also consistent with numerous studies in the empirical literature (Nguimkeu and Tadadjeu, 2021; Nguimkeu and Okou, 2022; Koudjom et al., 2022). Studies show that respiratory diseases such as COVID-19 are transmitted through contact (Alirol et al., 2011). High population density increases inter-community

contact and therefore the spread of COVID-19. The positive and significant effect of urbanization is also consistent with previous work (Nguimkeu and Tadadjeu, 2021; González-Val and Sanz-Gracia, 2022; Shen et al., 2021). Nguimkeu and Tadadjeu (2021) and González-Val and Sanz-Gracia (2022), show that urbanization favors the spread of COVID-19 in a sample of 182 and 40 countries respectively. In the case of China, Shen et al (2021) confirm the previous results by showing that this negative effect is driven by air pollution. The negative effect of urbanization can be explained by the fact that in urban areas, high population density makes it difficult to comply with barrier measures. The negative effect of government effectiveness is consistent with the work of Koudjom et al. (2022). This negative effect is explained by the fact that the appearance of the first cases in the various African countries was followed by measures to stop the spread of the disease.

**Table 2. Effect of income inequality on COVID-19 spread in Africa**

	Log Total COVID-19 cases		Log COVID-19 severity	
	(1)	(2)	(3)	(4)
GINI concentration index	0.513** (0.224)	0.775** (0.336)	0.878* (0.521)	0.657** (0.319)
Log Population density	0.743*** (0.115)	0.311** (0.122)	0.317* (0.198)	0.292*** (0.087)
Log GDP per capita	-0.145 (0.096)	-0.692 (0.593)	-0.004 (0.031)	-0.078 (0.334)
Government effectiveness	-0.711*** (0.206)	-0.501** (0.224)	-0.046 (0.081)	-0.461*** (0.071)
Urban population	0.162 (0.137)	0.266** (0.114)	0.128*** (0.039)	0.174*** (0.048)
Informality		0.323** (0.161)		0.277*** (0.085)
Literacy		-0.169* (0.096)		-0.465** (0.189)
R <sup>2</sup>	0.621	0.823	0.751	0.712
Fisher	0.234	0.442	0.293	0.254
Number of countries	43	43	43	43

Note: Significance \*\*\*  $p < 0.01$ ; \*\*  $p < 0.05$ ; \*  $p < 0.1$ . (.) Standard deviations.

Source: Authors.

Finally, the positive effect of the size of informal sector on COVID-19 spread is also consistent with studies by Nguimkeu and Okou (2022). The informal sector in Africa being characterized by crowded markets, congested streets and communal sanitation facilities may be the cause of the spread of the disease. The negative effect of literacy also corroborates the results of the work of Lopes and Mckay (2020) which show the importance of education on improving health. Lack of education is sometimes associated with misinformation, non-compliance with barrier measures. Educated populations better understand the merits of respecting the barrier measures necessary to stop the progression of the disease (Lopes and Mckay, 2020).

### 2.1.2. Robustness analysis

The Fixed-effects approach minimizes the impact of experimental errors, while improving the information available in the measurement process. However, it is not always efficient and can generate high variances. So, with a view to using an unbiased estimator of the model parameters generating a lower variance, we also use the Generalized least squares estimator (proposed by Aitken, 1935) for robustness. The results obtained in Table 3 are consistent with those obtained in Table 2, namely a positive and significant effect of income inequality on the total

number of cases and disease severity. Similarly, the results obtained by the control variables are also consistent with those obtained in Table 2. We obtain a positive and significant effect of population density, urban population and informal sector on the one hand, a negative and significant effect of government effectiveness and literacy on the total number of cases and disease severity on the other.

**Table 3. Effect of income inequality on COVID-19 spread, a robustness analysis on GLS approach**

	Log Total COVID-19 cases		Log COVID-19 severity	
	(1)	(2)	(3)	(4)
GINI concentration index	0.1597*** (0.348)	1.029*** (0.267)	0.923* (0.547)	0.603** (0.253)
Log Population density	0.304*** (0.101)	0.147** (0.061)	0.637*** (0.127)	0.127* (0.074)
Log GDP per capita	-0.681 (0.454)	-0.688 (0.573)	-0.815 (0.543)	-0.362 (0.241)
Government effectiveness	-0.388** (0.184)	-0.275*** (0.091)	-0.0179 (0.0747)	-0.0173** (0.007)
Urban population	0.300*** (0.0938)	0.232** (0.107)	0.124*** (0.0352)	0.167*** (0.0439)
Informality		0.0154** (0.00685)		0.0247* (0.015)
Literacy		-0.0176** (0.00828)		-0.039** (0.019)
Prob (Chi2)	0.000	0.000	0.000	0.000
Number of countries	43	43	43	43

Note: Significance \*\*\*  $p < 0.01$ ; \*\*  $p < 0.05$ ; \*  $p < 0.1$ . (.) Standard deviations.

Source: Authors.

## 2.2. Analysis of the indirect effect of income inequality on COVID-19 spread

For the analysis of the indirect and total effect of income inequality on COVID-19 spread in Africa, we draw on the work of Papyrakis and Gerlagh (2004) and Adams and Fotio (2022). We adapt the described methodology to the context of our study. The main hypothesis of this study is that inequalities can have not only a direct, but also an indirect effect on the COVID-19 spread in such a way that they generate externalities that can either amplify or slow down the spread of COVID-19. Thus, Equation (1) may be underestimated (or overestimated) if inequalities indirectly affect COVID-19 spread. Two channels are identified in the literature (Gradstein, 2007; Gutiérrez-Romero, 2022), namely the size of informal sector and the literacy. To determine the existence of a mediated relationship, we specify a system of structural equations represented by Equations (1) and (3) in which the mediating variables are dependent. Equation (3) can be written as follows:

$$M_{its} = \delta_0 + \delta_1 \times Ineq_{it} + \chi_{its} \quad (3)$$

where  $M_{is}$  is the vector of the  $s$  transmission channels (size of informal sector and literacy) in country  $i$ .  $\delta_1$  is the elasticity of the  $s$  channel with respect to income inequality.  $\delta_0$  is the constant, and  $\chi_{is}$  is the error term. The mediation effect is only possible if  $\delta_1$  is significant. After replacing Equation (3) in Equation (1), we obtain the following Equation (4):

$$LogY_{it} = (\alpha + \tau\delta_0) + X'_{it} \times \gamma + (\beta + \delta_1\tau)Ineq_{it} + \tau\chi_{its} + \sigma_i + \omega_t + \varepsilon_{it} \quad (4)$$

In Equation (4),  $\beta$  indicates the direct effect, while  $\tau\delta_1$  tells us about the indirect effect. Similarly,  $(\beta + \tau\delta_1)$  reflects the total effect of income inequality on the spread of COVID-19. We estimate the direct and indirect effects of income inequality on the COVID-19 spread using structural equation modeling. In addition, following the work of Adams and Fotio (2022), the indirect effect is obtained using the product of the Sobel coefficients. This effect is obtained from the following Equation (5):

$$\frac{\partial \text{Log}Y_i}{\partial M'_{is}} \times \frac{\partial M'_{is}}{\partial \text{Ineq}_i} = \tau\delta_1 \tag{5}$$

Table 2 shows that income inequality increases COVID-19 spread in Africa. Although these results are interesting, they may not reflect reality. We hypothesize that in addition to this direct effect, income inequality may indirectly affect the spread of COVID-19. In this case, the net effect of income inequality may be different from its direct effect. To test this hypothesis, we estimate the effect of income inequality on transmission channels. The results are contained in Table 4. They show that income inequality indirectly affects COVID-19 spread through its effects on the informal sector and literacy. All things being equal, a one-percentage-point increase in income inequality is associated with a 91.4 % increase in size of informal sector and a 59.1% reduction in literacy in Africa. These results are consistent with those obtained by Chong and Gradstein (2007) and Gutiérrez-Romero (2022). These results can be explained by the fact that due to low and/or unevenly distributed incomes, individuals prefer to enter the informal sector, as the formal sector requires the payment of taxes and fixed fees for registration with the authorities (Besley et al., 2012). Therefore, migrating to the formal sector requires sufficient wealth or many assets (Gutiérrez-Romero, 2022). Furthermore, the negative effect of income inequality on literacy rate can't be directly compared to previous work. Nevertheless, studies show that income inequalities lead to inequalities in consumption expenditure (Krueger and Perri, 2002). This disparity in consumption expenditure may be driven fundamentally by lower spending on education.

**Table 4. Effect of income inequality on the mediators**

	Informality (1)	Literacy (2)
GINI concentration index	0,914*** (0,168)	-0,591** (0,224)
Constant	7,305** (3,616)	18,06*** (6,225)
Number of countries	43	43

Note: Significance \*\*\*  $p < 0.01$ ; \*\*  $p < 0.05$ ; \*  $p < 0.1$ . (.) Standard deviations.

Source: Authors.

Table 5 assesses the total effect of income inequality on COVID-19 spread in Africa. The results indicate that the positive indirect effect of income inequality on the spread of COVID-19 is more pronounced in terms of the number of cases, while it is less pronounced regarding the severity of the disease, when compared to the findings presented in Table 2. Therefore, all other things being equal, a one-percentage-point increase in income inequality is associated with 97 % (column 1, compared to 77.5% in Table 2) increase in the total number of COVID-19 cases and 63.5% increase in disease severity (column 2, compared to 65.7% in Table 2). These results indicate that the direct effect alone underestimates the effect of income

inequality on COVID-19 spread in Africa in the case of the number of COVID-19 cases and overestimates in the case of COVID-19 severity.

**Table 5. Total effect of income inequality on COVID-19 spread**

	Log Total COVID-19 cases		Log COVID-19 severity	
	(1)	(2)	(1)	(2)
GINI concentration index	0.970** (0.421)	0.635** (0.317)		
Log Population density	0.311** (0.122)	0.292*** (0.087)		
Log GDP per capita	-0.692 (0.593)	-0.078 (0.334)		
Government effectiveness	-0.501** (0.224)	-0.461*** (0.071)		
Urban population	0.266** (0.114)	0.174*** (0.048)		
$\chi_{i1}$ Informality	0.323** (0.161)	0.277*** (0.085)		
$\chi_{i2}$ Literacy	-0.169* (0.096)	-0.465** (0.189)		
R <sup>2</sup>	0.823	0.713		
Fisher	0.442	0.254		
Number of countries	43	43		

Note: Significance \*\*\*  $p < 0.01$ ; \*\*  $p < 0.05$ ; \*  $p < 0.1$ . (.) Standard deviations.

Source: Authors.

**Table 6: Indirect transmission channels and their relative contribution**

	Log total COVID-19 cases		Log COVID-19 severity	
	Coef.	Relative contribution	Coef.	Relative contribution
Informality	0.295	74.87 %	0.253	48.01 %
Literacy	0.099	25.13 %	0.274	51.99 %
Total	0.394	100 %	0.527	100 %

Source: Authors.

Finally, based on Table 2 (columns 2 and 4) and Table 4, we calculate the contribution of each transmission channel to the total indirect effect of income inequality on COVID-19 spread. Similar to previous work by Adams and Fotio (2022) or more recently by Yeyouomo and Asongu (2023), we rely on the Sobel product coefficient approach. The results contained in Table 6 reveal that the size of the informal sector is the channel that contributes most to the total effect. Considering the total number of cases, the size of informal sector explains 74.87% against 25.13% for literacy of the positive indirect effect of income inequality on the spread of COVID-19. In the case of disease severity, the size of informal sector explains 48.01% and literacy 51.9 % of the positive indirect effect of income inequality on the spread of the disease.

## CONCLUSION

This study empirically assesses direct and indirect effects of income inequality on COVID-19 spread on a sample of 43 African countries over the period 2020-2022, by using a fixed effects regression model and the general least squares (GLS) method. The results suggest that income inequality increases COVID-19 spread in Africa. This effect is made possible by the proportion of the informal sector and the level of literacy. The size of informal sector is the main channel, accounting for 74.9% of the total number of cases and 48% of the severity of the disease. These results show that direct analysis underestimates (in the case of the number of COVID-19 cases) and overestimated (in the case of COVID-19 severity) the effect of income inequality on COVID-19 spread. These results show that income inequalities are an obstacle to tackling health shocks through the negative externalities they generate on literacy and the informal sector. As indicated by Legido-Quigley et al. (2020), COVID-19 highlights the need for a certain level of knowledge to ensure that citizens have sufficient information to change their behaviour to limit the spread of the disease. More broadly, the pandemic underscores the need for more effective engagement within the informal sector, given its significant role in the spread of the disease.

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## APPENDIX

TableA1. List of countries

Angola (AGO)	Congo (COD)	Guinea-Bissau (GNB)	Mozambique (MOZ)	Sudan (SDN)
Benin (BEN)	Ivory Coast (CIV)	Kenya (KEN)	Namibia (NAM)	Tanzania (TZA)
Botswana (BWA)	Dem. Rep. of Congo (COG)	Lesotho (LSO)	Niger (NER)	Togo (TGO)
Burkina Faso (BFA)	Djibouti (DJJ)	Liberia (LBR)	Rwanda (RWA)	Tunisia (TUN)
Burundi (BDI)	Egypt (EGY)	Madagascar (MDG)	Senegal (SEN)	Uganda (UGA)
Cameroon (CMR)	Eswatini (SWZ)	Malawi (MWI)	Seychelles (SYC)	Zambia (ZMB)
Cape Verde (CPV)	Ethiopia (ETH)	Mali (MLI)	Sierra Leone (SLE)	Zimbabwe (ZWE)
Chad (TCD)	Gambia (GMB)	Mauritania (MRT)	Somalia (SOM)	
Comoros (COM)	Ghana (GHA)	Mauritius (MUS)	South Africa (ZAF)	

Table A2. Variables definition

Variables	Definitions	Sources
Log Total COVID-19 cases	Number of confirmed COVID-19 cases in Africa	WHO (2022)
Log COVID-19 severity (%)	Also known as severity of infection, this is the extent of illness in people infected with the corona virus	WHO (2022)
GINI concentration index (%)	This index highlights income inequality at the aggregate population level	PovcalNet-World Bank (2022)
Population density	Average number of inhabitants in the given area per square kilometer	WDI (2022)
Log GDP per capita	Level of income per capital in each African country	WDI (2022)
Health expenditure per capita	Level of health expenditure per capita in each African country	WDI (2022)
Informality (%)	Share of informal employment as a percentage of total employment	ILO (2022)
Literacy rate (%)	Percentage of people ages 15 and above who can both read and write with understanding a short simple statement about their everyday life	WDI (2022)
Government effectiveness	Measures institutional quality: ranging from -2.5 to 2.5	WGI (2023)
Urban population	Urban population growth (annual %). People living in urban areas defined by national statistical offices	WDI (2022)

Source: Authors.

**Table A3. Multicollinearity test**

	VIF	1/VIF
GINI concentration index	1.16	0.8592
Log GDP per capita	1.2	0.8313
Log Population density	1.28	0.7842
Informality	2.32	0.4304
Government effectiveness	1.56	0.6413
Literacy rate	1.25	0.7982
Urban population	1.79	0.5590
Mean VIF	1.51	

*Source: Authors.***Table A4. Correlation matrix**

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
(1) Log Total COVID-19 cases	1.000								
(2) Log COVID-19 severity	0.130	1.000							
(3) Income inequality (%)	0.042	0.131	1.000						
(4) Informality (%)	-0.391	-0.141	0.221	1.000					
(5) Literacy rate (%)	0.180	-0.046	0.074	-0.295	1.000				
(6) Log GDP per capita	0.586	-0.042	-0.021	0.048	0.280	1.000			
(7) Government effectiveness	0.396	0.069	-0.108	-0.554	0.103	0.032	1.000		
(8) Population density	-0.117	0.055	-0.119	-0.119	0.062	-0.212	0.239	1.000	
(9) Urban population (%)	0.252	0.360	-0.002	0.658	0.197	0.071	0.396	0.617	1.000

*Source: Authors.*

## **Les effets des inégalités de revenu sur la transmission de la Covid-19 en Afrique : le rôle du secteur informel et de l'analphabétisme**

**Résumé** - La littérature empirique a peu exploré les canaux de transmission dans l'étude de la relation entre les inégalités de revenu et la propagation de la Covid-19. L'objectif de cet article est de combler cette lacune en évaluant les effets directs et indirects des inégalités de revenu sur la propagation de la pandémie sur un échantillon de 43 pays africains au cours de la période 2020-2022. Deux indicateurs de la Covid-19 sont utilisés, à savoir le nombre total de cas et la sévérité de la maladie, et des variables explicatives jouent un rôle sensible comme la population urbaine ou la densité de population. Deux canaux de transmission sont particulièrement étudiés, la taille du secteur informel et l'alphabétisation. Les résultats, à partir d'un modèle de régression à effets fixes ou estimé par les moindres carrés généralisés, montrent que les inégalités de revenu ont un effet significatif sur la propagation de la Covid-19. Toutefois, il apparaît que ces résultats sous-estiment les effets des inégalités si l'on considère les effets indirects dus au poids du secteur informel et de l'analphabétisme.

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### **Mots clés**

COVID-19  
Inégalités de revenus  
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