



# **Drivers of Carbon Emissions in Kenya: The Perspective of Technology**

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## **Authors' contributions**

*This work was carried out in collaboration among all authors. All authors read and approved the final manuscript.*

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## **ABSTRACT**

The empirical studies have portrayed from a different perspective that technology, foreign direct investment (FDI), and economic growth have diverse outcomes on carbon dioxide (CO<sub>2</sub>) emissions. African economies specifically Kenya are currently threatened with more CO<sub>2</sub> emissions for which proper strategies need to be adopted to reduce and mitigate this situation. To address this issue, the Autoregressive Distributed Lag (ARDL) technique was utilized, differentiating between the long-term and short-term effects of drivers of CO<sub>2</sub>. Granger causality was applied to analyze the causality between the series. The study uses time series data for Kenya from 1990 to 2022. The short-run and long-run results indicate a negative relationship between technology and carbon emissions. Specifically, an increase in technology reduces carbon emissions while an increase in economic growth and FDI inflow increases carbon emissions. The study has identified the negative nexus between technology and carbon emissions in Kenya. This means that technological innovation can increase labour productivity and utilization of resources, thereby reducing carbon emissions. Therefore, Kenya should increase research and utilization of low and efficient carbon technologies to decouple economic growth from environmental pollution. Based on the result, various strategies have been proposed, including using different clean

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technologies to produce renewable fuels. FDI that complies with environmental regulations, and the use of advanced green and clean technologies can help mitigate the negative impact of technological progress on the environment. Additionally, technological progress, particularly in information and communication technology, will reduce greenhouse gas emissions and improve environmental quality.

*Keywords: Technology; carbon emissions; environmental quality; patent.*

## 1. INTRODUCTION

Climate change and greenhouse gas (GHG) emissions are global issues that significantly influence the environment, ecosystems, technological progress, economic development, and human health. Climate change is an ongoing process that can lead to altered climatic conditions and has the potential to affect the stability of ecosystems and biodiversity. It also has significant impacts on human health and can contribute to the occurrence and distribution of disease patterns [1]. Global environmental policies to reduce CO<sub>2</sub> emissions include targeting the extreme emitters in the electricity industry, implementing carbon pricing and taxes, tightening environmental policies to reduce carbon leakage through international trade, and adopting sustainable climate change policies based on COVID-19 practices [2]. Global environmental issues, particularly the reduction of carbon dioxide (CO<sub>2</sub>) emissions, have become a significant concern worldwide [3]. The excessive consumption of fossil fuels has led to limited reserves and the release of large amounts of CO<sub>2</sub>, contributing to environmental problems such as global warming and climate change [4]. To address these issues, various strategies have been proposed, including using different technologies to produce renewable fuels or useful chemicals through the photocatalytic reduction of CO<sub>2</sub> [5].

Literature has emphasized the relevance of the environmental Kuznets curve (EKC) as proposed by Grossman and Krueger [6], who posit that environmental practices and quality are impacted by economic growth, energy consumption and production activities. The Environmental Kuznets Curve (EKC) theory provides theoretical contributions to the relationship between economic development and environmental degradation. It suggests an inverted U-shaped nexus, where environmental damage initially increases during the early stages of economic development but then decreases after a certain level of economic progress is reached [7,8]. Kuznets (1955) laid the foundation for this theory

and discussed the specific relationship between economic growth and environmental quality (Lau et al., 2023).

Muganyi et al. [9] showed that the interplay between technological advancement carbon emissions and environmental quality has become a topical issue among environmental economics and scientists. Technological innovations and institutional quality have a significant impact on environmental quality. Enhancing technological innovations and institutional effectiveness can help resolve environmental degradation issues and achieve sustainable development [10]. Prior research has shown that technology, trade openness, foreign direct investment, and economic growth can contribute to environmental pollution, particularly in CO<sub>2</sub> emissions [11]. The discussion so far has portrayed from a different perspective that Technology advancement, FDI, and economic growth have diverse outcomes on CO<sub>2</sub> emissions. African economies specifically Kenya are currently threatened with more CO<sub>2</sub> emissions for which proper strategies need to be adopted to reduce and mitigate this situation. Therefore, our study will assist the government, financial, technological, and environmental regulatory institutions, and other organizations as well as policymakers in pursuing more definite practical, pragmatic, and proper innovative initiatives related to CO<sub>2</sub> emissions in Kenya.

## 2. LITERATURE REVIEW

This part of the study delves into the details of prior literary works on the association between technology, economic growth, foreign direct investment, and carbon emissions.

### 2.1 Theoretical Literature

The Environmental Kuznets Curve (EKC) theory suggests that as per capita income increases, environmental degradation initially worsens but eventually improves. This theory is based on the idea that as income rises, societies have the resources and desire to invest in environmental protection [6,12]. Several studies have examined the EKC hypothesis in different countries and

regions. Wen and Dai [13] find evidence of the EKC in China, with pollution levels initially increasing and then decreasing as per capita income rises. Maranzano et al. [14] highlight the role of education in validating the EKC hypothesis and propose a new model called the Educational EKC. Boudinar [8] focuses on developed countries and investigates the existence of the EKC hypothesis. Dai et al. [15] examine the relationship between rural economic development and non-point source pollution in Jiangsu Province, China, and verify the EKC hypothesis. Lawn, [16] extends a theoretical model and argues that achieving improved environmental quality requires policies focused on sufficiency, equity, natural capital maintenance, and qualitative improvement of human-made goods.

Thus, this research provides empirical evidence on the role of technological innovation in mitigating CO<sub>2</sub> emissions. Environmental economists and scientists have identified technological advancements such as ICT infrastructure development, technological advances in the use of hydrogen fuel cells, carbon pricing, forest restoration, energy efficiency, and electric vehicles indicating that the efficiency of technology influences CO<sub>2</sub>. Therefore, in this research, we expect an inverse association between technology and CO<sub>2</sub> emission in Kenya.

## **2.2 Empirical Literature**

### **2.2.1 Technology and carbon emissions (CO<sub>2</sub>) nexus**

Technology plays a significant role in carbon emissions (CO<sub>2</sub>). Several studies have examined the relationship between technology and CO<sub>2</sub> emissions. Chaudhry and Chandni [17] suggest that innovation technology is a primary contributor to CO<sub>2</sub> emissions in India, both in the long term and short term. However, other research indicates that technological innovation can also decrease environmental degradation [18]. The impact of technology on environmental quality depends on various factors such as the type of technology, energy sources, and institutional quality [10]. Information and communication technology (ICT) usage also contributes to CO<sub>2</sub> emissions in the long run, although the relationship is not significant in the short run [19]. The impact of technology on carbon emissions also varies depending on the specific indicators used. For example, ICT

variables are found to be effective instruments for carbon reduction, while R&D expenditure and patents do not significantly influence carbon emissions [20]. Additionally, the expansion of renewable electricity and the promotion of trade openness can help mitigate CO<sub>2</sub> emissions [21]. Overall, technology and its various components have a complex relationship with carbon emissions, with both positive and negative effects depending on the specific context and indicators used [22]. The quality of governance and institutions plays a significant role in determining the relationship between technology and environmental quality. Countries with better institutional quality and governance tend to have lower carbon emissions and better environmental quality. This indicates that technology is a significant contributor to CO<sub>2</sub> emissions reduction [7].

### **2.2.2 Foreign direct investment (FDI) and carbon emissions (CO<sub>2</sub>) nexus**

Foreign direct investment (FDI) has a significant impact on carbon emissions in different countries and regions. The relationship between FDI and carbon emissions is complex and varies depending on factors such as the level of education, environmental management systems, and technological innovation [23]. Education levels play a moderating role, where FDI inflows may result in higher emissions in economies with low levels of schooling, but can reduce CO<sub>2</sub> emissions when education levels exceed certain thresholds [24]. Studies have shown that FDI can lead to increased carbon emissions in some cases, particularly in developing countries with low levels of education [25], which suggests that countries with weaker environmental regulations and lower environmental quality attract more FDI, leading to increased pollution levels [26,27,28]. However, in other cases, FDI has been found to harm carbon emissions, especially when combined with effective environmental management systems and higher levels of education [23,24]. Additionally, technological innovation plays a crucial role in mediating the relationship between FDI and carbon emissions, with FDI influencing carbon emission efficiency through its impact on technological innovation [29]. FDI quality upgrading can significantly reduce urban carbon emissions, especially in cities with low industrial rationalization [30], high investment in science and technology, strong environmental regulation, and high dependence on foreign capital [31]. Dong [32] identified direct impact and spatial spillover effect of FDI on

carbon emissions have temporal and regional differences in China, emphasizing the need for optimizing FDI quality and governance to achieve energy conservation and emission reduction goals. FDI can have both positive and negative impacts on environmental quality, with its effects depending on factors such as the rationalization degree of industrial structure, investment in science and technology, environmental regulation, and dependence on foreign capital. These findings highlight the importance of considering education, environmental management systems, and technological innovation when examining the impact of FDI on carbon emissions. According to the above arguments, FDI inflow increases carbon emissions tends to increase.

### 2.2.3 Economic growth and carbon emissions (CO2) nexus

Economic growth has been found to have a significant influence on CO2 emissions in several studies. In the European Union (EU) countries, economic growth was found to have a statistically significant effect on CO2 emissions, with a 1% change in GDP leading to a 0.072 change in CO2 emissions on average [33]. A study on Indonesia also found a positive relationship between economic growth and CO2 emissions in the long run [34]. In the Beijing-Tianjin-Hebei region, economic growth was found to reduce CO2 emissions, while CO2 emissions also contributed to economic growth to some extent. Additionally, economic growth facilitated industrial structure upgrading, and the advanced industrial structure promoted economic growth, but the rationalization of the industrial structure had a limiting effect on economic

growth [35]. Similarly, in West Africa, economic growth was found to increase CO2 emissions, following the Environmental Kuznet Curve hypothesis. However, in the presence of improved governance quality, economic growth did not significantly influence CO2 emissions in the region [36]. Other factors such as trade, urbanization, and renewable energy also influenced CO2 emissions in the region [30]. In Tunisia, economic growth was positively correlated with carbon emissions, indicating that the country has not yet reached the required level of per capita income for an inverted U-shaped Environmental Kuznets Curve. Energy consumption and financial development were also found to contribute to environmental damage in the long run [33]. These findings highlight the need for effective governance and environmental policies to mitigate the impact of economic growth on CO2 emissions.

## 3. METHODOLOGY

This chapter captures relevant data collection methods and data analysis techniques.

### 3.1 Research Design

The study employed a quantitative research design to investigate the relationship between technology progress and carbon emissions in Kenya for the period 1990 to 2022. African economies specifically Kenya are currently threatened with more CO2 emissions for which proper strategies and policies need to be adopted to reduce and mitigate this situation. Time series secondary data was obtained from the World Bank database. Table 1 provides a detailed description of the study variables.

**Table 1. Definition of Variables**

Variables	Unit of Measurement	Definition	Data source	Expected sign
<b>Dependent variable</b>				
Carbon emissions (CO2)	Metric tons	Carbon dioxide emissions	World Bank	No prediction [11]
<b>Independent variables</b>				
Technology (TECH)	Numbers	Patent applications (residents and non-residents)	World Bank	Negative [10]
Foreign Direct Investment (FDI)	Per cent	Foreign Direct Investment, net inflows (% GDP)	World Bank	Positive [26]
Economic Growth (ECG)	Constant 2015 US dollars	GDP per capita is Gross Domestic Product divided by midyear population	World Bank	Positive [11]

### 3.2 Model Establishment

The research employed the following model to investigate the role of technological progress on carbon dioxide emissions.

$$CO2_t = f(TECH_t, FDI_t, ECG_t) \quad (1)$$

To eliminate skewness and serial correlation problems the data variables were transformed into natural logs.

Hence, we can re-write the above equation as follows:

$$\ln CO2_t = \beta_0 + \beta_1 \ln FDI_t + \beta_2 \ln TECH_t + \beta_3 \ln ECG + \varepsilon_t \quad (2)$$

Where  $\varepsilon_t$  represents the error term,  $t$  represents time dimensions,  $\beta_0$  shows the constant term, and  $\beta_1$  to  $\beta_3$  are coefficients variables.

ARDL estimation model was applied as shown in Equation 3

$$\Delta \ln CO2_t = \alpha_0 + \sum_{i=1}^p \alpha_{1i} \Delta \ln CO2_{t-i} + \sum_{i=0}^w \alpha_{2i} \Delta \ln TECH_{t-i} + \sum_{i=0}^w \alpha_{3i} \Delta \ln COV_{t-i} + \beta_{11} \ln CO2_{t-1} + \beta_{21} \ln TECH_{t-1} + \beta_{31} \ln COV_{t-1} + \varepsilon_{1t} \quad (3)$$

Equation 4 was used to analyze the long-run relationship between the target variables.

$$\ln CO2_t = \alpha_0 + \sum_{i=1}^p \alpha_{1i} \ln CO2_{t-i} + \sum_{i=0}^w \alpha_{2i} \ln TECH_{t-i} + \sum_{i=0}^w \alpha_{3i} \ln COV_{t-i} + \varepsilon_{it} \quad (4)$$

To check for causality association error correction term equation 5 was used.

$$\Delta \ln CO2_t = \alpha_0 + \sum_{i=1}^p \alpha_{1i} \Delta \ln CO2_{t-i} + \sum_{i=0}^w \alpha_{2i} \Delta \ln TECH_{t-i} + \sum_{i=0}^w \alpha_{3i} \Delta \ln COV_{t-i} + \phi_1 ECT_{t-1} + \varepsilon_{1t} \quad (5)$$

The lagged error correction term  $ECT_{t-1}$ , in equation 5 measures the speed of adjustment to the long-run equilibrium and also the long-run causality relationship.

Where  $COV_t$  – Matrix of Control variables (FDI and ECG) and  $ECT_{t-1}$  is the error correction term.

### 3.3 Estimation Procedure

Following the empirical works of Thomi et al. [37], the Phillips –Peron (PP) unit root test was employed to scrutinize stationarity in the times series data because it is a non-parametric test and robust to the general form of heteroscedasticity in the error term. The F-bounds co-integration test was applied to confirm the presence of a long-run relationship between study variables. Before the estimation, lag length and best model estimator were identified. ARDL estimation method was used to analyze the determinants of carbon emission. ARDL is highly robust regardless of the stationarity and endogeneity of the target variables [38]. Granger causality test was conducted to check for association between study variables. All models were subjected to diagnostic tests such as autocorrelation (Breusch-Godfrey test) and heteroscedasticity (Breusch-Pagan test) to avoid misleading inferences.

## 4 RESULTS AND DISCUSSION

This chapter reports the analysis results and discussion.

### 4.1 Unit Root Test

The research applied the Phillips-Perron (PP) unit root test to investigate the unit root problem. The PP result is reported in Table 2.

Based on the unit root result, carbon emission and economic growth are non-stationary and only become stationary after the first difference. FDI and technology were stationary at the level. Considering most variables are integrated of order I (1) including dependent variables the study proceeded to investigate the long-run relationship. Investigation of lag length is important to reduce serial correlation problems. Vector autoregression (VAR) lag selection result (-7.927380\*) has confirmed that the appropriate lag length is 5 and the best estimation method is the Akaike information criterion (AIC), as supported by the minimum value.

**Table 2. Unit root test**

Variable	Level		Variable	First difference		Integration order
	t-Statistics	P-Value		t-Statistics	P-Value	
CO2	-0.751313	0.8192	$\Delta$ CO2	-6.071111	0.0000	I(1)
ECG	1.700120	0.9994	$\Delta$ ECG	-3.910589	0.0054	I(1)
TECH	-3.153463	0.0325				I(0)
FDI	-4.936749	0.0003				I(0)

Note: Null Hypothesis: data series has a unit root

## 4.2 Cointegration Test

The study adopted the Pesaran et al. [38] cointegration test to investigate the long-run relationship between the study variables. Table 3 shows the F-bound cointegration test result.

From the result, F statistics (5.299) is greater than the upper bound critical value (5.070), and thus cointegration or long-run relationship is present. The study proceeded to check for long-run and short-run effects.

## 4.3 Estimation Results

ARDL estimation was applied to investigate the long-run and short-run relationship between target variables based on the Akaike information criterion (AIC) (3, 4, 3, 4). Table 4 represents the ARDL estimation of the parameter result.

Based on Table 4 result, an increase in technology development will reduce carbon emissions in Kenya, in the long run. Specifically, an increase in technology and innovation by 1 % will reduce carbon emissions by 0.079%. Technology can enhance the efficiency of carbon emissions and as such reduce carbon emissions. Improved technology can significantly reduce carbon emission intensity and improve carbon emission efficiency. This means that technological innovation can increase labour productivity and utilization of resources, thereby reducing carbon emissions. Furthermore, additional energy consumption pollution is offset by the improvement of technological innovations. Also, software technology reduces the number of physical resources and thus uses less energy. Hence, enhancing technological innovations can help resolve environmental degradation issues and achieve sustainable development [10]. The findings agree with research by Gao and Fan [10] and Nawaz et al. [11] empirical studies. The short-run findings agree with long-run findings that an increase in technology development will translate to a reduction in carbon dioxide emissions and improve environmental quality.

An increase in GDP will increase carbon emissions in Kenya. In Kenya increase in GDP by 1% will increase CO<sub>2</sub> emission by 0.654%. Economic growth has been found to have a significant influence on CO<sub>2</sub> emissions in several

studies. Carbon emission increases as energy consumption and economic activities are enhanced in the economy. Increased labour force and capital formation will increase further carbon emissions. These findings support the Environmental Kuznets Curve (EKC) hypothesis. The study finding is similar to Nawaz et al. [11] empirical study. In the EU countries, economic growth was found to have a statistically significant effect on CO<sub>2</sub> emissions, with a 1% change in GDP leading to a 0.072 change in CO<sub>2</sub> emissions on average [33]. A study on Indonesia also found a positive relationship between economic growth and CO<sub>2</sub> emissions in the long run [34]. In the Beijing-Tianjin-Hebei region, economic growth was found to reduce CO<sub>2</sub> emissions, while CO<sub>2</sub> emissions also contributed to economic growth to some extent. Economic growth can improve CO<sub>2</sub> emissions by increasing carbon dioxide emissions levels. The studies conducted in China, Indonesia, Nepal, and the EU member states [34] supported positive results. In contrast, the short-run result indicates a negative relationship between economic growth and CO<sub>2</sub> emissions. This is possible with enhanced GDP the economy may adopt better technology and cleaner sources of energy and thus reduce carbon emissions. Economic growth facilitated industrial structure upgrading, and the advanced industrial structure promoted economic growth, but the rationalization of the industrial structure had a limiting effect on economic growth [35].

An increase in foreign direct investment (FDI) inflow will increase carbon emissions in Kenya. The result indicates FDI inflow is positively related to carbon emission. Specifically, increase in FDI level by 1 will increase carbon emission by 0.109 %. FDI will expand the production capacity of the host country via new investment, which results in carbon emission in the production unit involved and as such increases the carbon dioxide level [39]. The findings agree with Serge et al. [25] research. Studies have shown that FDI can lead to increased carbon emissions in some cases, particularly in developing countries with low levels of education [25], which suggests that countries with weaker environmental regulations and lower environmental quality attract more FDI, leading to increased pollution levels [26,27,28]. However, in other cases, FDI has been found to harm carbon emissions, especially when combined with effective environmental management systems and higher levels of education [23,24]. The findings agree with short-run results.

**Table 3. Cointegration Results**

Narayan	Value	Significance Level	Bounds Critical values	
F-Statistics	5.299641		I(0)	I(1)
		1%	5.17	6.36
K	3	5%	4.01	5.07
		10%	3.47	4.45

Note: Null hypothesis: No level relationship.

**Table 4. Regression results**

Long run results				Short run results			
Variable	Coefficient	t-statistics	P-value	Variable	Coefficient	t-Statistic	P-value
TECH	-0.079	-3.032**	0.0126	ΔTECH	-0.032	-4.145***	0.0020
ECG	0.654	4.097***	0.0022	ΔECG	-1.226	-2.023*	0.0706
FDI	0.109	7.313***	0.0000	ΔFDI	0.020	2.032*	0.0695
Cons	-2.525	-2.204*	0.0520	Cons	-2.525	-5.250***	0.0004
				ECT(-1)	-0.975	-5.249***	0.0004
Goodness of fit		R-Squared		0.708097			
Econometrics Problems		Test Statistic		F-statistics		P-Value	
Serial Correlation		Breusch-Godfrey LM		0.343513		0.7192	
Heteroscedasticity		Breusch-Pagan-Godfrey		0.470490		0.9213	

Note: \*  $p < 0.1$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$  are significance levels, in which the null hypothesis is rejected.

Dependent variable: CO2

**Table 5. Pairwise causality result**

Direction	F-Statistic	P-Value	Relationship
TECH → CO2	3.18451	0.0354	One way causality
CO2 → TECH	0.37815	0.8215	

The coefficient of error term (ECT) was significant and negative in Kenya, implying a long-run relationship exists between carbon emissions and technology. It also implies that 0.957 will be corrected in case of any disequilibrium in Kenya in the current year; disequilibrium will exist for a short time. Having confirmed long-run causality using ECT, the study proceeded to check for short-run causality using a pairwise causality test. The coefficient of determination is 0.708, implying about 70 % of the variation in carbon emission is explained in the model. The study also conducted diagnostic tests and controlled for the effect. Based on the diagnostic result, serial correlation and heteroscedasticity were not a problem in the regression model.

#### 4.4 Causality Test

A pairwise causality test was applied to confirm the association between technology and carbon emission in the short run. Table 5 shows the pairwise causality result.

The study result has identified one-way causation runs from technology development to carbon emission. This means technological progress has a relationship with carbon emissions. Technology can enhance the efficiency of carbon emissions and as such reduce carbon emissions.

#### 5. CONCLUSION

African economies specifically Kenya are currently threatened with more carbon dioxide (CO2) emissions for which proper strategies need to be adopted to reduce and mitigate this situation. To achieve this objective, the study employed advanced econometric methods to investigate the role of technology and innovations on carbon emissions. The study used the Phillips Peron unit root test, F-bound cointegration test, ARDL estimation approach, and causality test. Secondary data for Kenya for the period 1990 to 2022 was obtained from the World Bank database. The results indicate a negative relationship between technology and carbon emissions. This means that technological

innovation can increase labour productivity and utilization of resources, thereby reducing carbon emissions. In contrast, economic growth and FDI inflow have a positive sign on carbon emissions. Technology, economic growth and FDI have been identified as the main drivers of carbon emissions in Kenya. The study has also established a causal relationship running from technology to carbon emissions. Based on the result, various strategies have been proposed, including using different technologies and innovations to produce renewable fuels. Kenya needs to adopt the use of biofuels and wind turbines to reduce carbon emissions by adopting energy sources that do not burn fossil fuels. Also designing and building electric cars can slow fossil energy use and carbon emissions. Several carbon reduction technologies and innovations should include practices such as adopting new forms of agriculture and implementing reforestation. The use of green and clean technologies and encouraging energy efficiency can help mitigate the negative impact of technological progress on the environment. Additionally, technological progress, particularly in the information and communication technology (ICT) sector, will reduce greenhouse gas emissions. Policymakers and stakeholders should emphasize research and development for updated production technology while raising economic growth helps to reduce CO<sub>2</sub> emissions. Additionally, environmental policies should aim to minimize emissions, FDI should be environmentally friendly and green technology advanced, and incentives needed to redirect private credits toward green projects and renewable energy development. To achieve sustainable development, Kenya needs to ensure that FDI inflows comply with environmental regulations and standards to avoid transferring technologies, innovations and industrial activities that increase pollution problems.

### COMPETING INTERESTS

Authors have declared that no competing interests exist.

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