Environmental preservation and international trade are the two major issues related to human well-being and development [1]. This has generated a looming debate on the impact of trade on environmental quality. This debate has generated more heat than light as noted by [2]. Empirically, studies have thoroughly investigated the goods and overall trade impact on the environmental quality with little focus on the services. Therefore, a separate investigation of the environmental consequences of services trade is still unexplored. Our main objective in this study is to investigate the direct and indirect effect of services trade liberalization on environmental quality in the context of African countries. To achieve this objective, we used the Environmental Kuznets Curve (EKC) model and decomposed the trade effect into the scale, technique, and composition effects. This is because the environmental impact of free trade can be studied using this model while decomposing the effect of trade into scale, technique, and composition effects. The scale...
effect as elaborated by [2] shows the effect of increase economic activities (GDP) on carbon emissions. While the technique effect refers to the gainful effect of increased income on carbon emissions necessitated by strict environmental regulations and people’s demand for a cleaner environment. The composition effect, on the other hand, is determined by the comparative advantage derived from trade and it refers to the effect of change in the composition of output. The composition effect can assert a positive or negative effect on environmental pollution depending on the resource endowed by the country. Through the scale, technique, and composition effects, trade could affect the environment, this is because trade could increase income, productive activities, and composition of productive inputs which also increase emissions. A study by [2] among others, for instance, has found a positive scale effect, negative technique effect, and negative composition effect based on SO₂ data. While decomposing trade effect into scale, technique, and composition, most empirical studies report mixed findings for the goods and overall trade impact on the environment. The environmental impact of services trade may be different owing to the different modes of service supply. Therefore, there is a need to investigate the specific role of services in generating or mitigating CO₂ emissions.

Examining the impact of services trade on the environment is important owing to the growth in services over the last few decades. Growth in international trade has been much more attributed to the growth in the services sector and of services globally. Most countries, all over the world had experienced a tremendous increase in services trade. More importantly at the same time, countries around the world had also experienced environmental challenges owing to climate change, greenhouse gas emissions, air quality, and ozone layer depletion [3-5]. Services are an important input to different manufacturing industries and there is a need to understand its environmental consequences via the international trading system. As noted by [6] services help to reduce environmental degradation resulting from increased economic activities, though some services were found to increase emissions. For instance, in 2013 the disaggregated GHG emissions in Africa show that transport services account for 42% of the total GHG emissions which exceeds the global average [7]. In African countries, services trade has increased rapidly which cannot only be a driver of economic expansion but also a means of reducing emissions as most services are believed to be environment friendly. Theoretically, the turning point of the EKC is achieved by a transition from industry to services, so also services trade can play a role in reducing countries’ environmental degradations. In African countries that are mostly developing the impact of services trade on the environment may differ from the notion that services have a beneficial effect on environmental quality. This is because [8] have noted that foreign trade has a detrimental impact on the environment in developing countries. With this background it is debatable to conclude on the environmental impact of services trade in African countries. Therefore, this debate can only be clear via an empirical study of the services trade impact on environmental quality.

In African countries, there are alarming environmental issues that range from oil pollution, energy, air pollution, water pollution, loss of biodiversity. The continent is also faced with the problem of rising population, poverty, hunger, illiteracy, and political uprising. This has made the continent remained underdeveloped and more susceptible to environmental degradation. Despite these problems, the continent’s contribution to global emissions has remained low below 5% of the global CO₂ emissions [9].

As a result of the low contribution to global emissions by the continent, the environmental consequences of free trade had been largely ignored by both researchers and policymakers in the continent. Therefore, there is a need to understand the forces behind environmental degradation in African countries by taking the specific role of services trade. This is because the continent is developing, more open to foreign trade, and also facing many environmental challenges which may be the result of market opening and other impending factors related to human activity. This is because [10] in his speech on the Perspective on Global Warming mentioned that “human activity is a significant contributor to climate change” which can cause CO₂ emissions and environmental degradation.

With this in mind, therefore, this study contributes to the literature of trade and environment in many ways. Firstly, there is little emphasis given by researchers to services trade and the environment nexus. In this vein, our study contributes to the debate of trade and environment by solely investigating the effect of services trade on the environment in African countries. To the best of our knowledge, we are first to consider such an impact in the context of African countries. Secondly, the study also recognized the role of the turning point in the trade-environment nexus which has not been investigated by any empirical study in the context of African countries. This is important because trade may assert a different effect on the environment at a different level of trade openness. For instance, at a low level of openness, trade may be detrimental to the environment while at an advanced level it might have a beneficial effect resulting from technology spillover. So also if at a low level of openness, trade benefits the environment, it may degrade the environment at a high level because of pollution haven effect. Thirdly, we also incorporated GDP cubic components into the model of EKC to determine whether the first turning if any, is only temporary. To our knowledge and search for literature, this has not been investigated by any empirical study using a panel of African countries. Fourthly, the study also contributes to the existing knowledge by applying a new trade openness measure proposed by [11]. Besides, unlike previous studies, that
measure services openness by only considering cross border services i.e. modes 1 and 2. In constructing our services trade openness we incorporate all modes of services supply to have a broader measure of services trade openness. Fourthly, we considered the indirect channels via which trade can affect the environment by decomposing trade effect into the scale, technique, and composition effects. Lastly, we also investigate the pollution haven hypothesis (PHH) effect of trade based on income and factor abundance comparative advantage which has never been investigated by any empirical study in the context of African countries.

**Literature Review**

On the theoretical ground, the EKC hypothesis postulate that at the initial stage of development increased income is positively correlated with environmental pollution and after reaching the threshold point, environmental quality improves with further increase in per capita income [12]. Studies have questioned whether this threshold can be attained domestically or by transferring pollution to poor and developing countries [13, 14] in which case developing countries will become a pollution haven. In this case, therefore, even EKC explained the PHH in which advanced countries use trade as a means of moving their polluting industries to poor and developing countries. The optimism and the extent to which EKC can explain the transfer of pollution from developed to developing countries is not well justified [15]. The PHH, therefore, described a scenario in which trade openness improves the environment in some open countries and degraded it in other countries [16]. Another hypothesis based on factor abundance postulates that resources owned by countries are the main determinants of trade and environmental pollution [17]. According to this hypothesis, developed countries with abundant capital will become dirtier as they possessed a comparative advantage in capital-intensive export and production. While labour abundance developing countries will become cleaner as they explore comparative advantage in labour-intensive export and production. This hypothesis, therefore, operates in contrast to the PHH. The present study will focus on investigating the impact of trade on the environment within the context of the theoretical underpinning of the above theories more especially the EKC hypothesis. Many empirical studies applied the EKC model while investigating the environmental impact of free trade and the extent to which countries are becoming pollution haven resulting from trade openness. These studies have reported mixed findings which may be due to the use of different methodologies, model specification problems, and the way the trade openness is measured.

On the empirical ground, studies on the environmental impact of services trade are scarcer, the few available works of literature include; [18-20] who find that commercial services trade, financial services trade, and broader services trade reduces carbon emissions while GDP, energy consumption, renewable energy, and industrialization increase carbon emissions in 25 major developing countries, BRIC and China. A cross-country analysis by [21] observed that services trade increases CO₂ emissions and reduces SO₂ emissions. Other empirical studies [22-28] have confirmed the beneficial effect of mode 3 services supply (FDI) on the environment as it reduces CO₂ emissions. On the contrary [20-30] have found mode 3 services to increase CO₂ emissions and detriment the environment.

The few available literatures in the context of African countries include; [27, 31-34], which finds trade openness, GDP, energy consumption, and capital-labour ratio to increase CO₂ emissions and reduce environmental quality. A study by [35] reports that in both the short-run and long-run trade openness, GDP, and renewable consumption asserts asymmetric effect on CO₂ emissions in Nigeria and South Africa. In the context of South Africa [36] observed that trade openness, human capital, and renewable energy consumption improve environmental quality by reducing ecological footprint while GDP degrades the environment by increasing ecological footprint. Empirical findings by [37] revealed no significant effect of democracy on CO₂ emissions while globalization reduces CO₂ emission and energy use increases emissions and degrade the environment in South Africa. An empirical finding by [38] established that Africa’s regional trade has contributed to reducing CO₂ emissions and PM₁₀ in the continent. A study by [23] revealed that trade and capital stock increase CO₂ emissions while GDP decreases CO₂ emissions in Morocco. A study by [39], observed that trade openness decreases CO₂ emissions while GDP, energy use, and urbanization increase emissions and degrade environmental quality in OPEC member countries. In the case of Nigeria [40] has found no evidence of trade effect on CO₂ emissions while GDP increases emissions and that energy consumption, manufacturing value-added reduces carbon emissions. In another study [41] revealed that GDP increases CO₂ emissions while renewable energy consumption reduces emissions in Nigeria.

Other studies used heterogeneous panels and explore the role of trade in generating environmental pollution. These studies include; [42, 43] who observed trade openness to increases CO₂ emissions and degrades the environment. So also [44] reports that trade openness, GDP, and capital labour-ratio significantly influence carbon emissions. Studies by [45, 46] also revealed that trade and GDP degrade the environment while democratic governments improve environmental quality by reducing emissions and increasing air quality. In another study by [47], democracy was found to improve environmental quality in the short-run by reducing CO₂ emissions while energy consumption degrades the environment by increasing CO₂ emissions.
In the context of Iran’s trade with the Organization of Economic Cooperation and Development (OECD), Middle East, and East Asian countries [48] report that emissions have increased in Iran resulting from its trade with these regions. A recent study by [49] has found trade, energy efficiency, and renewable energy to reduce emissions while GDP and industrialization increase GHG emissions. Another study [50] observed that for the global sample and low-income countries trade degrade the environment while for high-income countries it improves the environment by reducing PM$_{2.5}$.

In the context of developed countries studies have also established evidence of the beneficial and detrimental effect of trade on the environment. In the case of the U.S [51] find no evidence of trade impact on toxic releases while a study by [30] report that trade openness and capital-labour ratio contribute to reducing emissions and that GDP, energy consumption increase carbon emissions. In another study, [52, 53] revealed that trade openness, GDP, and energy consumption increases CO$_2$ emissions in Turkey. For the developing countries sample, a study by [54-56] reports that trade and economic globalization reduces CO$_2$ emissions. Studies by [28, 29, 57, 58] indicate that trade openness, GDP, energy consumption industrialization are detrimental to the environment. Findings by [59, 60] show that trade and renewable energy reduced CO$_2$ emissions and improve environmental quality in China, BRICS, and N-11 countries. Contrarily, [61, 62] show no evidence of trade effect on CO$_2$ emissions while GDP and energy consumption increases CO$_2$ emissions in the case of Pakistan and India. In the case of China and Indonesia [63, 64] observed that trade, energy use, and industrialization increase emissions while GDP growth was found to decrease emissions and improve environmental quality.

Some empirical studies have demonstrated the PHH effect using the input-output (I-O) model and econometric approach in the context of Africa and other regions. For instance, [15]’s findings revealed that U.S and U.K as advanced countries use trade as a mechanism to transfer their polluting industries to the South African Customs Union (SACU) countries. Similarly, [65] has found that the net CO$_2$ emissions transfer to China has resulted from its trading activities with Western Europe, North America, and other developed countries whereas its emissions outflow embedded in trade are transfer to Sub-Saharan Africa, America, and South Asia. Using computable general equilibrium [66] reports that advanced countries tend to shift their pollution to developing countries. A study by [25] confirmed that developed countries used to trade as a means of transferring their pollution to developing countries. [67], show that trade has contributed toward net emissions outflow in the Chinese Hebei province. A study by [68] supported the existence of PHH in China’s domestic trade. Similarly, [69] revealed that a bulk of Beijing’s fuel-related mercury pollutions were outsourced through trade. A study by [70] show that China’s trade with “Belt and Road Initiative” countries has resulted in the concentration of China’s export toward resource-intensive industries and increase pollution.

### Material and Methods

#### Model Specification

To examine the environmental impact of services trade, we follow empirical literature and incorporated variables related to the scale, technique, and composition effects. Besides, we augmented interaction terms between the explanatory variables into the function of the CO$_2$ emissions to verify the existence of income and factor endowment PHH. We formulate an empirical model within the context of standard EKC hypothesis with unobserved panel specification of the following form:

$$\ln CO_{2it} = \lambda_0 + \lambda_1 \ln GDP_{it} + \lambda_2 \ln GDP_{it}^2 + \lambda_3 \ln GDP_{it}^3 + \lambda_4 \ln STO_{it} + \lambda_5 \ln STO_{it}^2 + \lambda_6 \ln KL_{it} + \lambda_7 \ln EI_{it}$$

$$+ \lambda_8 \ln GDP_{it} \cdot \ln STO_{it} + \lambda_9 \ln KL_{it} \cdot \ln STO_{it} + \lambda_{10} \ln EI_{it} \cdot \ln STO_{it} + \lambda_{11} \ln AGR_{it} + \lambda_{12} \ln IND_{it}$$

$$+ \lambda_{13} \ln SER_{it} + \lambda_{14} \ln POL_{it} + \psi_i + \eta_t + \epsilon_{it}$$

(1)

...where: $i = 1, 2, 3,....\ldots$ represent the country dimension in the cross-section, $t = $ is the time period. All the variables were transformed into a natural log. $CO_{2i}$ (i.e. emissions measured in per capita metric tons) is a measure of environmental quality. $GDP_i$ is the real GDP per capita which measures the scale effect. If the coefficient of GDP is positive and significant we verify the scale effect. $GDP^2_i$ is squared of $GDP$ which measures the technique effect and the existence of the EKC hypothesis. The negative and statistically significant coefficient of $GDP$ squared verify the technique effect and the existence of the EKC hypothesis. $GDP^3_i$ is the GDP cubic component introduced to verify whether the first turning point of EKC if exists is only temporary. Positive and statistically significant $GDP^3$ will validate the existence of N-shape EKC. $STO_i$ is the services trade openness expected to assert either a positive or negative effect on emissions. $STO^2_i$ is the squared of trade openness, introduced to verify the non-linear nexus between trade and environmental pollution. Positive $STO_i$ and negative $STO^2_i$ would verify the hypothesis that, at a lower level of openness, trade may be harmful to the environment, and after reaching an advanced stage of trade liberalization it may have a beneficial effect on the environment as a result of technology spillover. $KL_i$, is the capita-labour ratio which measures the composition effect. Positive and statistically significant $KL_i$ would support the hypothesis that trade alters the composition of the industry towards dirtier production and increase pollution. $EI_{it}$ is the energy intensity measuring...
the energy effect and is hypothesized to increase emissions.

The interactions among the variables of interest are introduced to measure and verify the existence of income and factor endowment PHH and to verify whether trade openness has enabled the use of energy-efficient technology in mitigating carbon emissions. A positive and statistically significant coefficient of GDP * STO would validate the hypothesis that developing countries like Africa are pollution havens in which case developed countries used trade openness to transfer their pollution to the continent. A negative and statistically significant coefficient KL * STO would support the hypothesis that trade allows labour abundance countries like Africa to specialize in labour-intensive production and exportation which is less polluting. A negative coefficient EI * STO would provide evidence that trade allows for the use of energy-efficient equipment that is less polluting. Other control variables include; Polity POL which measures the level of a country’s democracy and expected to lower emissions. AGR, IND and POLSER are the agriculture, industry, and services value-added as % of GDP with an a priori expected sign of negative, positive, and negative. \( \lambda_1, \ldots, \lambda_{1t} \) are the parameters to be estimated concerning the explanatory variables. \( \epsilon_{it} \) is the error term that is assumed to be independently and identically distributed. The term \( \psi_i \) is the unobserved effect specific to an individual country and doesn’t change with time, \( \eta_t \) is the time-specific effect that is time-variant and is common to all individual countries.

Model (1) can simply be estimated using Ordinary Least Square (OLS) or restricted OLS provided that across countries, the individual effects \( \psi_i \) are constant and there is no time effect such that \( \eta_t = 0 \). If country effects \( \psi_i \) are not constant but random with disturbances \( \epsilon_i \) and there is no time effect \( \eta_t = 0 \), such that \( \psi_i = \epsilon_i + \eta_i \). Where the expectations of \( \epsilon_i \) are zero i.e. \( E(\epsilon_i) = 0 \) with variance \( \epsilon_i = \sigma_i^2 \) and covariance \( \epsilon_i, \epsilon_i \) = 0. With this Model (1) can be estimated using Random Effect (RE) models or Generalized Least Squares (GLS). If we further assume country effects \( \psi_i \) to be constant and not equal across countries such that \( \psi_i = \eta_i \) and \( \eta_i \neq 0 \) then Model (1) can be estimated using the fixed effect (FE) model.

Given that most economic variables and their associations are dynamic and the potential endogeneity that cannot be addressed using POLS, FE, and RE models, we employed the dynamic panel data modelling. This approach is based on an estimate known as the Generalized Method of Moment (GMM) proposed by [71]. In a dynamic model the environmental pollution and its determinants which changes across time and space can be specified as follows:

\[
\ln(CO_{2it}) = \lambda_0 + \lambda_1 \ln(CO_{2i,t-1}) + \delta_1 \ln(X_{it}) + \delta_2 \ln(Y_{it}) + \psi_i + \eta_t + \epsilon_{it} \tag{2}
\]

...where \( CO_{2i,t} \) is the environmental quality indicator measured as carbon emissions. \( CO_{2i,t-1} \) is the lagged dependent variable introduced to capture the dynamic properties of the model. \( \lambda_1 \) is the adjustment coefficient. \( X_{it} \) and \( Y_{it} \) are the vectors of explanatory and control variables as defined in Model (1). \( \psi_i \) is the country-specific fixed effects, \( \eta_t \) is the time trend which captures global shocks, and \( \epsilon_{it} \) is the idiosyncratic time-way error component; \( i = 1, 2, \ldots, N \) are the number of countries and \( t = 1, 2, \ldots, T \) is the number of years. To eliminate the country effect in Model (2) Arellano and Bond’s estimation procedures required taking the first difference of Model (2) as follows:

\[
\ln(\Delta CO_{2it}) = \lambda_1 \ln(\Delta CO_{2i,t-1}) + \delta_1 \ln(\Delta X_{it}) + \delta_2 \ln(\Delta Y_{it}) + \Delta \eta_t + \Delta \epsilon_{it} \tag{3}
\]

...where: All variables are as defined in Model (2) except here they all entered the model as first differences. By differencing the individual effect is now automatically eliminated in Model (3). While differencing to remove the individual effect, the within estimator is also biased because the differences in the lagged dependent variable \( (\Delta CO_{2i,t} = CO_{2i,t-1} - CO_{2i,t-2}) \) are also correlated with the differences in the random disturbance term \( (\Delta \epsilon_{it} = \epsilon_{it} - \epsilon_{it-1}) \). As a result, [72] pointed out that with a small-time period and persistent time series the first difference GMM estimator may likely be biased because of weak instruments provided by lagged levels. If the first difference estimator is biased because of weak instrumentation resulting from the persistency of the lagged dependent variable, the most appropriate estimation technique is the system GMM estimator proposed by [72]. Despite the weakness of the first difference estimator, it is still considered relevant and less biased if the persistency of the data is weak. In this case, the first difference GMM would be more appropriate as it would possess some efficiency gains compared to the system GMM estimator. Owing to the efficiency gains from using either the difference or system GMM, we used the [73]’s rule of thumb to decide on the most appropriate estimator for our data set.

Data Source and Variable Measurement

The study utilized data for a sample of 47 African countries obtained from the World Bank, World Development Indicators, and the Polity IV Project at the University of Maryland over the period 2000-2014. The choice of this period is motivated by data availability and to capture the period African countries embarked on a massive trade liberalization strategy.

We construct a measure of services trade openness based on the [11]’s approach. [11], have argued that an economy is considered open if it has a high trade/GDP ratio and considerable trade share relative to the rest of the world. Our services openness measure based on [11]’s approach is given by:
\[ STO_i = \frac{(SX + SM)_i}{1/n \sum_{j=1}^{n} (SX + SM)_j} \times (SX + SM)_i / GDP_i \] (4)

...where \(i\) stands for country subscript, \(\phi\) stands for trading partners or the rest of the world. \(STO_i\) is the composite services trade openness, which is a composite measure of country \(i\)'s trade/GDP ratio and its trade share relative to the rest of the world. \(SX\) and \(SM\) are country \(i\)'s services export and import. \(n\) denotes the number of countries.

This measure has an advantage over the trade/GDP ratio more especially in cross-country and panel studies. In cross-country analysis, a high trade/GDP ratio may not reflect the actual degree of countries' openness. This is because smaller countries may tend to have a high trade/GDP ratio for a reason that their GDP is small relative to their trade. While larger economies whose trade constitutes a smaller fraction of their economic size would tend to have a low trade/GDP ratio even though, they are significant contributors to the global trade [11]. Also, we adopted the definition of services trade provided by the General Agreement on Trade in Services (GATS) in constructing the openness of our services trade. Therefore, our constructed \(STO\) is broader services openness that includes all modes of services supply as defined by GATS which include mode 1: cross border services trade, mode 2: consumption abroad, mode 3: commercial presence (FDI), mode 4: movement of the professional workers to provide services in the territory of another country. Data for remittance received and paid by African countries was used to proxy for services supplied via the movement of professional workers. To this end, in this study, \(SX\) and \(SM\) are broad services exports and imports that include all modes of services supply.

\(CO_2\) emissions in metric tons per capita represent the emission that every citizen is responsible for. GDP is measured as real per capita gross domestic product estimated at constant 2010 US$. The Capital-labour ratio is the ratio of capital stock to the economically active population (ages 15-65). Energy intensity presents the energy used per unit of output computed based on the level of primary energy (MJ/$2011 PPP GDP). Agriculture, industry, and services are sectoral value-added as a percentage of GDP. Democratic government is measured by the polity index with a score ranging from -10 (highly autocratic) to +10 (highly democratic).

**Results and Discussion**

The correlation coefficient of the dependent variable and GDP is 0.91 which is strong and highly significant at less than 1 percent level. Our analysis shows that services trade openness, energy intensity, services value-added, and polity are weakly but significantly correlated with the dependent variable with a correlation coefficient of 0.27, -0.43, 0.21, and -0.08 respectively. Our analysis further demonstrates that the capital-labour ratio and agriculture value-added are also strongly correlated with the dependent variable with 0.87 and 0.82 coefficients while industry value-added is moderately correlated with the dependent variable with a 0.58 coefficient. The correlation among the regressors does not reveal a high correlation that may result in a multicollinearity problem to our empirical analysis. We further conducted the variance inflation factor (VIF) test, to again check for multicollinearity. The result of the test for all the estimated linear and polynomial models also revealed that our estimated models did not have multicollinearity. This is because VIF values are all below the standard required value of 6 and the threshold value of 10. The estimated VIF values for the estimated models (1)-(9) lies between 2.55 and 3.98.

For our empirical analysis, we used [73] rules of thumb to decide on the difference and system GMM estimate. The rule of thumb is based on the estimate of the dynamic version of POLS, FE, and first difference GMM. The estimate of the lagged dependent variable from the pooled OLS is considered as an upper bound estimate while that of FE is a lower bound estimate. According to [73] if the estimate of the lagged dependent variable from the difference GMM is below the FE estimate, the system GMM should be used because the first difference estimate would be downward biased resulting from weak instrumentation [72]. If the first difference estimate is above the FE estimate, then the difference GMM should be used because it would provide a consistent, efficient, and less biased estimate. This rule of thumb when applied to our data has proven the need to use first difference GMM instead of system GMM. This is because we find that the lagged dependent variable of the difference GMM estimate is above that of the FE estimate. Therefore, in our case, the difference GMM should be preferred to system GMM as it is more appropriate for our data set. Table 1 reports the result of the two-step GMM estimate, where the lagged dependent variable from the difference GMM is preferred to the pooled OLS instead of system GMM. We further conducted the variance inflation factor (VIF) test for all the estimated linear and polynomial models also revealed that our estimated models did not have a multicollinearity problem to our empirical analysis. We estimated nine dynamic models that are different in two dimensions i.e. linear and cubic polynomial models for more robustness of the empirical findings. The p-values of AR(2), Sargan and Hansen tests show that there is no second-order serial correlation, and our instrument set is valid. From estimates (1)-(9) the coefficients of the lagged dependent variable are positive and highly statistically significant at less than 1% level. This finding suggests that the current emissions are affected by past emissions levels. This finding is supported by [22, 31, 57] among others who all found the current emissions to be positively influenced by past emissions. The baseline model (1) tells that except \(KL * STO\) all our variables of interest asserts a statistically significant
impact on CO₂ emissions. There is strong evidence that GDP increases CO₂ emissions in all estimated models. The result from the baseline model (1) suggests that a 1% increase in GDP is associated with a 0.538% increase in carbon emissions. This implies that the scale effect is observed in a panel of African countries. The technique effect measured by the squared of GDP is negative and statistically significant. The coefficient of GDP cubic is positive and statistically significant in (1)-(6) estimates. This finding is a disproved to EKC hypothesis as obtained by the positive scale effect and negative technique effect. This implies that there is an N-shaped nexus between GDP and emissions in African countries. Our finding is supported by recent studies like [40, 56, 58, 60].

Finding also revealed that an increase in trade openness deteriorates the environment by increasing CO₂ emissions. From the baseline models (1) of Table 1, the finding suggests that a 1% increase in trade openness increases carbon emissions by 0.0175%. This finding is robust to different linear and cubic polynomial model estimates and supported by [18, 19, 21] among others. The non-linear component of trade openness is positive and statistically significant in most estimates. This implies a monotonic increase as there is no turning point for trade to decrease carbon emissions. A plausible explanation is that services flow especially

| Table 1. Dynamic GMM estimate for services trade effect on environmental pollution (CO₂). |
|----------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|---------------------------------|
| L.lnCO₂                          | 0.694*** (0.193)                | 0.675*** (0.134)                | 0.635*** (0.114)                | 0.606*** (0.138)                | 0.662*** (0.115)                | 0.611*** (0.149)                | 0.638*** (0.163)                | 0.690*** (0.241)                | 0.632*** (0.178)                |
| lnGDP                            | 0.538* (0.283)                  | 0.593* (0.328)                  | 0.509* (0.307)                  | 0.537** (0.270)                 | 0.515** (0.202)                 | 0.523* (0.268)                  | 0.585** (0.274)                 | 0.577* (0.328)                  | 0.517*** (0.199)                |
| lnGDP²                           | -0.181** (0.0796)               | -0.233** (0.0952)               | -0.262** (0.0905)               | -0.222** (0.0864)               | -0.220** (0.0853)               | -0.270*** (0.112)              |                                  |                                  |                                  |
| lnGDP³                           | 0.0728* (0.0386)                | 0.0824* (0.0464)                | 0.0833* (0.0439)                | 0.0710* (0.0371)                | 0.0805** (0.0350)               | 0.0879* (0.0467)                |                                  |                                  |                                  |
| lnSTO                            | 0.0175** (0.00716)              | 0.0234* (0.0139)                | 0.0234* (0.0134)                | 0.0137* (0.0070)                | 0.0245*** (0.00644)             | 0.0134* (0.00804)               | 0.0383* (0.0216)                | 0.0117* (0.00683)               | 0.0140* (0.00828)               |
| lnSTO²                           | 8.56e-05 (0.000164)             | 0.000563* (0.000301)            | 0.000532* (0.000270)            | 0.000303* (0.000126)           | 1.17e-05 (0.000201)             | 0.000262* (0.000151)           |                                  |                                  |                                  |
| lnKL                             | 0.227* (0.134)                  | 0.254** (0.125)                 | 0.243* (0.139)                  | 0.247* (0.145)                  | 0.236* (0.131)                  | 0.216* (0.121)                  | 0.316* (0.170)                  | 0.436** (0.214)                  | 0.877** (0.371)                  |
| lnEI                             | 0.598*** (0.166)                | 0.593*** (0.170)                | 0.562*** (0.187)                | 0.701*** (0.210)                | 0.656*** (0.176)                | 0.641*** (0.198)                | 0.603*** (0.190)                | 0.538*** (0.258)                | 0.583*** (0.143)                |
| lnGDP*lnSTO                      | 0.0227* (0.0135)                | 0.0232*** (0.00827)             | 0.0126* (0.00743)               | 0.0199* (0.0118)                | 0.0128** (0.00508)              | 0.0223* (0.0114)                |                                  |                                  |                                  |
| lnKL*lnSTO                       | -0.00518 (0.0155)               | -0.00970 (0.0100)               | -0.00351 (0.0144)               | -0.00810* (0.00467)            | -0.00810* (0.00467)            | -0.000438 (0.00108)            |                                  |                                  |                                  |
| lnEI*lnSTO                       | 0.0312** (0.0144)               |                                  | 0.0478* (0.0196)                | 0.105*** (0.0348)              | 0.029** (0.0170)                |                                  |                                  |                                  |                                  |
| lnAGR                            | -0.0211 (0.107)                 | -0.00262 (0.223)                | -0.0820 (0.179)                 | -0.121 (0.185)                 | -0.0268 (0.167)                 | -0.277** (0.137)                |                                  |                                  |                                  |
| lnIND                            | 0.00124 (0.0773)                | 0.0197 (0.0758)                 | 0.00202 (0.0614)                | 0.298** (0.136)                | 0.0989 (0.112)                  | 0.403* (0.218)                  |                                  |                                  |                                  |
| lnSER                            | 0.123 (0.297)                   | 0.190* (0.106)                  | 0.300** (0.127)                 | 0.336* (0.183)                 | 0.0326 (0.139)                  | 0.0334 (0.183)                  |                                  |                                  |                                  |
| lnPOL                            | -0.209* (0.117)                 | -0.340*** (0.129)               | -0.315** (0.129)                | -0.217*** (0.0876)             | -0.0683 (0.0861)                | -0.0695 (0.142)                 |                                  |                                  |                                  |
| Sargan p-value                   | (0.560) (0.705)                 | (0.693) (0.691)                 | (0.389) (0.348)                 | (0.325) (0.281)                 | (0.143)                        |                                  |                                  |                                  |                                  |
| Hansen p-value                   | (0.629) (0.349)                 | (0.282) (0.509)                 | (0.473) (0.459)                 | (0.314) (0.389)                 | (0.552)                        |                                  |                                  |                                  |                                  |
| AR(2) p-value                    | (0.505) (0.624)                 | (0.604) (0.238)                 | (0.379) (0.405)                 | (0.474) (0.465)                 | (0.371)                        |                                  |                                  |                                  |                                  |
| Observations                     | 539                            | 536                            | 536                            | 534                            | 534                            | 536                            | 536                            | 516                            |                                  |

Note: The statistical significance of the estimates at <1%, <5% and <10% are denoted by ***, **, and * respectively. Robust standard errors were in parenthesis except for Sargan, Hansen, and AR(2) which are p-values.
of mode 3 come from less-advanced countries that have less sophisticated technology which increases CO\textsubscript{2} emissions. This finding contradicts [43]'s finding who found an inverted U-shaped for trade-pollution nexus. The composition effects worsen the environment by increasing CO\textsubscript{2} emissions. From the baseline model (1) a percentage increase in capital-labour ratio is associated with a 0.227% increase in CO\textsubscript{2} emissions at less than 10% significance level. This finding is robust to different estimates in Table 1 and is supported by [27, 44] but contradicts [30] among others. Evidence revealed that emissions increases with an increase in energy intensity. From the baseline model, a 1% increase in energy use is associated with a 0.598% increase in CO\textsubscript{2} emissions. This finding is supported by many recent studies [18, 31, 60] and contradicts [40] among others. The estimated coefficient of energy use is high because over 90% of countries in Africa relied on conventional non-renewable energy sources [22] which significantly increases emissions. Since non-renewable energy use dominates over renewable energy use, this implies more emissions with increased energy consumption.

The interaction term between trade and GDP which measures the PHH yields a positive and statistically significant coefficient in all estimates. This implies that African countries are pollution haven when open to foreign trade. This result confirmed that developed countries use trade as a means of transferring their polluting activities to African countries and this is supported by [25] but contradicts [49] who confirmed a negative but statistically insignificant effect of

| VARIABLES | (1) | (2) | (3) | (4) |
|-----------|-----|-----|-----|-----|
| lnGDP     | 0.845*** (0.0684) | 0.977*** (0.0466) | 1.093*** (0.117) | 1.042*** (0.0224) |
| lnGDP\textsuperscript{2} | -0.0936*** (0.0214) |               |               |               |
| lnGDP\textsuperscript{3} | 0.0356*** (0.0138) |               |               |               |
| lnSTO     | 0.0264*** (0.00888) | 0.0116* (0.00682) | 0.0121** (0.00571) | 0.0118** (0.00592) |
| lnSTO\textsuperscript{2} | 0.000800* (0.000461) |               |               |               |
| lnKL      | 0.353*** (0.0365) | 0.352*** (0.0227) | 0.279*** (0.0558) | 0.364*** (0.0150) |
| lnEI      | 0.551*** (0.0364) | 0.490*** (0.0253) | 0.534*** (0.0414) | 0.479*** (0.0190) |
| lnGDP*lnSTO | 0.0685*** (0.0164) | 0.0708*** (0.0166) |               | 0.0807*** (0.0203) |
| lnKL*lnSTO | -0.0221 (0.0138) | -0.0239** (0.0113) |               | -0.0279** (0.0137) |
| lnEI*lnSTO | 0.0416*** (0.0130) | 0.0562*** (0.0109) |               | 0.0628*** (0.0103) |
| lnAGR     | -0.192*** (0.0430) | -0.0999*** (0.0227) | -0.0588** (0.0245) |               |
| lnIND     | 0.00349 (0.0325) | 0.0190 (0.0225) | 0.0414* (0.0216) |               |
| lnSER     | 0.164** (0.0689) | 0.243*** (0.0498) | 0.229*** (0.0612) |               |
| lnPOL     | 0.0363 (0.0232) | 0.0375 (0.0274) | 0.0440 (0.0326) |               |
| Constant  | -1.052*** (0.0377) | -1.137*** (0.0304) | -1.152*** (0.0227) | -1.162*** (0.0124) |
| Time effect | Yes | Yes | Yes | Yes |
| Observations | 642 | 642 | 642 | 642 |
| R\textsuperscript{2} | 0.799 | 0.749 | 0.726 | 0.741 |

Note: The statistical significance of the estimates at <1%, <5% and <10% are denoted by ***, **, and * respectively. Robust standard errors were in parenthesis.
trade via GDP on environmental pollution. On the factor abundance effect \((KL \ast STO)\), finding revealed less strong evidence supporting this hypothesis. The coefficient in all estimates has the expected theoretical sign but only statistically significant in model (7). This implies that there is no robust evidence that with open trade African countries enjoy comparative advantage in labour-intensive export and production which are less polluting. The indirect effect of trade via energy intensity is positive and statistically significant which implies that trade does not allow the use of energy-efficient technology that lowers \(CO_2\) emissions. This finding is robust to different linear and cubic polynomial models. The three mediation effects have further confirmed that services trade is more polluting in African countries. As expected the coefficient of agriculture is negative in all estimates but only statistically significant in model (8). Finding revealed little evidence that increasing the share of industry increases emissions. Contrary to our theoretical a priori we found an increasing share of services to increase carbon emissions. This finding further confirmed that services trade and the activities of the services sector are polluting in African countries. Consistent with the theoretical a priori, democratic government has a negative and statistically significant impact on carbon emissions and improves environmental quality.

Robustness Check

As a robustness check to our baseline cubic polynomial model (1) and the linear models (7)-(9) of Table 1, we used random coefficient and re-estimate the models as presented in Table 2. The RE model was chosen based on the Hausman test because the FE model cannot be consistently estimated for our panel. The empirical finding obtained from the alternative estimates is consistent with the result of the GMM estimate. This further implies that our finding is robust to different dynamic and static estimates as well as to linear and cubic polynomial models.

Conclusions

In the literature on trade and environment, several studies have thoroughly investigated the impact of goods and overall trade on environmental quality. To this end, this study deviates from previous studies and investigates both the direct and indirect effect of services trade on the environment in a panel of African countries. The empirical findings provided in this study are new to the literature. This is because the study adopted an openness index that is rarely used in trade literature which was constructed based on [11]'s approach. Findings revealed that the scale effect increases emissions while the technique effect reduces emissions and validate the EKC hypothesis. The decrease in emissions necessitated by the technique effect is only temporary as evidenced by the positive GDP cubic component. This implies that there is an N-shaped nexus between GDP and carbon emissions in African countries. Services trade was found to increase \(CO_2\) emissions and impede environmental quality and there is no evidence of a turning point in the services trade-\(CO_2\) emissions nexus. Finding also revealed that the composition effect increases \(CO_2\) emissions and degrades the environment. It is also found that the energy effect is positive and significant for the average African country. There is robust evidence that African countries are pollution haven when open to trade as developed countries used trade to transfer their polluting activities to the continent. Our finding revealed little evidence supporting the factor abundance effect. The mediation effect of trade via energy use is positive, indicating that trade has not allowed African countries to have access to energy-efficient technology which lowers emissions. For the control variables like democratic government, agriculture, industry, and services value-added findings revealed less strong evidence of their impact on the environmental quality indicator.

The policy implication of these findings is that, since the scale effect increases \(CO_2\) emissions and degrades the environment we, therefore, recommend the need to upgrade or replace the existing production techniques that are less efficient and more polluting with an up-to-date, more efficient, and less polluting technique of production. There is also the need to reduce pressure on resource use in meeting increasing domestic and foreign demand through “reduce” “recycle” and “re-use” of material when and wherever possible. This will help in striking a balance between sustainable growth and environmental pollution in the continent. Trade openness detrimental the environment in which failure to account for its specific role in generating \(CO_2\) emissions may result in poor \(CO_2\) emissions mitigation policies. We, therefore, recommend that policymakers should encourage the flow of services that are environment-friendly and capable of decreasing \(CO_2\) emissions. The positive and detrimental effect of services trade on the environment as obtained in this study is a clear indication that environmental problems arising from increased \(CO_2\) emissions are a purely global phenomenon. In this regards an international treaty is required in addressing the carbon emissions problem rather than relying on only national and regional policies.

The composition effect degrades the environment by increasing \(CO_2\) emissions but when interacted with trade openness it reduces emissions and improves environmental quality. Therefore, this finding is more appealing to policymakers. In this regard, we acknowledged the need for policymakers to implement policies aimed at reducing protection concerning the flow of less polluting services. Since energy use increases \(CO_2\) emissions a concerted effort should
be made to increase investment in renewable energy that is less polluting. This will reduce over-reliance on conventional non-renewable energy and help in mitigating environmental pollution. While shifting from non-renewable to renewable energy, policymakers should also make sure that renewable energy prices are reasonably lower for wider access and to reduce CO₂ emissions. This can be achieved by way of renewable energy consumption subsidies, lower import duty for solar panels, electric cars, etc. Environmental policies should be implemented to make sure that activities of the services sector are less prone to environmental damages by appraising the performance of the services sector’s environmentally-friendly adoption capacity. To mitigate CO₂ emissions in African countries, policymakers should also adopt a mechanism of pay as you emit on polluting activities. To mitigate the flow of pollution from advanced countries environmental regulations need to be strengthened to augur well for sustainable development and prevent the activities of polluting industries. Carbon emissions taxed should be imposed on foreign affiliate companies using energy-intensive equipment and subsidy should be granted to less polluting foreign affiliate’s activities. The role of a democratic government in reducing CO₂ emissions is less satisfactory in African countries. Therefore, policymakers should make effort in bringing active democratic freedom so that citizens’ political rights can lead to an increase in people’s awareness and demand for environmental quality.

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Conflict of Interest

The authors declare no conflict of interest.

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