



# Maritime Sector Growth, Environmental Regulation and CO<sub>2</sub> Emission Nexus: Evidence from Nigeria

Obasi Catherine Chimma<sup>1</sup>  
\*Aruwei Porwekobowei<sup>2</sup>

<sup>1,2</sup>Nigeria Maritime University, Okerenkoko, Delta State, Nigeria

\*Corresponding Author: Aruwei Porwekobowei

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## Abstract

*This study investigates maritime sector growth, environmental regulation and CO<sub>2</sub> emission nexus in Nigeria. The study used secondary time series data from 1990 to 2023 sourced from the Central Bank of Nigeria statistical bulletin and the World Development Indicator. The data for the study was analysed by Vector autoregression (VAR) model, after employing some preliminary data analysis such as descriptive statistics, ADF unit root test and cointegration test. Findings from the study showed that in the VAR model estimates the single most important determinant of each variable is its one-period lagged value. The Impulse Response Function result showed that there is a positive and stable relationship between the maritime sector growth and CO<sub>2</sub> emission both in the long run and short run. Environmental regulation have an asymmetric effect on CO<sub>2</sub> emission both in the long run and short run. The variance decomposition results showed that the predominant sources of CO<sub>2</sub> emission variation are largely due to its own shocks and maritime sector growth innovations. The study therefore recommends that the government should make stiffer and more enforceable environmental rules to reduce CO<sub>2</sub> emission from the maritime industry, lower emissions by supporting the maritime industry's adoption of sustainable practices and greener technology like alternative fuels and energy-efficient ships, create policies to strike a balance between environmental sustainability and the expansion of the maritime industry so as to prevent disproportionate rise of CO<sub>2</sub> emission.*

**Keywords:** Maritime Sector Growth, Environmental Regulation, CO<sub>2</sub> Emission, VAR, Cointegration

## I. Introduction

The rise in commercial activities and business demand have made seaborne trade

experience an amazing growth to provide about 80% of world goods in 2019. It increased from 7,857 million tonnes (Mt) in 2009 to 11,005 Mt in 2018, the laden volume of international marine traffic has therefore seen ten years of continuous increase until the COVID-19 epidemic struck (UNCTAD, 2020). Developing economics especially show this predominance (UNCTAD, 2020). The COVID-19 epidemic caused about 4% decline in international seaborne trade in 2020; yet, as the world economy started to recover in 2021, the upward trajectory returned (UNCTAD, 2022). In return for amazing growths, the environmental effects of intense maritime transportation started to be underlined, particularly the climate effects from exhaust Greenhouse Gases (GHG) (IMO, 2021; Ytreberg et al., 2021).

Although marine transport is among the most effective means of transport for goods, in 2018 it produced 1,076 Mt of GHG emissions, of which 1,056 Mt were CO<sub>2</sub> emissions, therefore making around 3% of all anthropogenic CO<sub>2</sub> emissions. But by 2025, the rate is expected to grow to 17%, suggesting that, absent intervention, worldwide shipping CO<sub>2</sub> emissions are trending in the other direction (ISEMAR, 2020). This is so since oil accounted for more than 99% of overall energy needs historically (IEA, 2023), so the maritime sector is more fossilised than other sectors. From 977 million tonnes in 2012 to 1076 million tonnes in 2018, the fourth IMO GHG research indicated that the overall marine GHG emissions, both domestic and international, comprising CO<sub>2</sub>, CH<sub>4</sub> and N<sub>2</sub>O, expressed in CO<sub>2</sub> equivalent emissions (CO<sub>2</sub>e) have increased by 9.6%. Of these around 98% are CO<sub>2</sub> emissions (IMO, 2020).

In Nigeria for instance, the land use change and forestry, Energy, waste, agricultural, and industrial operations sector contributed 38.2%, 32.6%, 14.0, 13.0% and 2.1% correspondingly to GHG emissions in Nigeria (IEA, 2023). Burning



coal, oil, and natural gas among other fuels drives the great majority of CO<sub>2</sub> emissions in the energy industry. This constant implies that trade and movement of products and services in the maritime sector adds to CO<sub>2</sub> emissions. Moreover, 80% of Nigerian carbon emissions directly result from burning fossil fuels (Ogundipe, Okwara and Ogundipe, 2020). Given Nigeria's reliance on ports and shipping for trade, oil exports and importation of goods, the maritime industry is a vital part of its economy (Aruwei and Eko-Raphaels, 2022). More so, with Nigeria been the fourth biggest economy and significant involvement in world trade, the maritime industry is of great importance (Bloomberg, 2024). Industrial activity includes shipping aggravates the major environmental issues the nation faces: air pollution, oil spills, and deforestation (IEA, 2023). This study looks at how environmental rules, CO<sub>2</sub> emissions, and Nigeria's maritime industry growth interact. It emphasises the economic value of the industry, its environmental effect, and the part laws help to reduce emissions. Complementing worldwide SDGs and Nigeria's development objectives, the results provide policy insights for juggling economic growth with sustainability. Furthermore used in the study are sophisticated econometric methods vector autoregression (VAR) models to examine the relationship among CO<sub>2</sub> emissions, environmental policies, and maritime sector growth. Thus, offering a strong methodological approach with ideas that may be repeated in many situations. The study offers localised data that might guide national plans for environmentally friendly maritime growth.

It is therefore imperative to investigate what this mean for a developing country like Nigeria that is trying to diversify its economy by exploring the blue economy? It is therefore important to examine the link between maritime Sector Growth, environmental regulation and CO<sub>2</sub> Emission from Nigeria. The rest of the study is

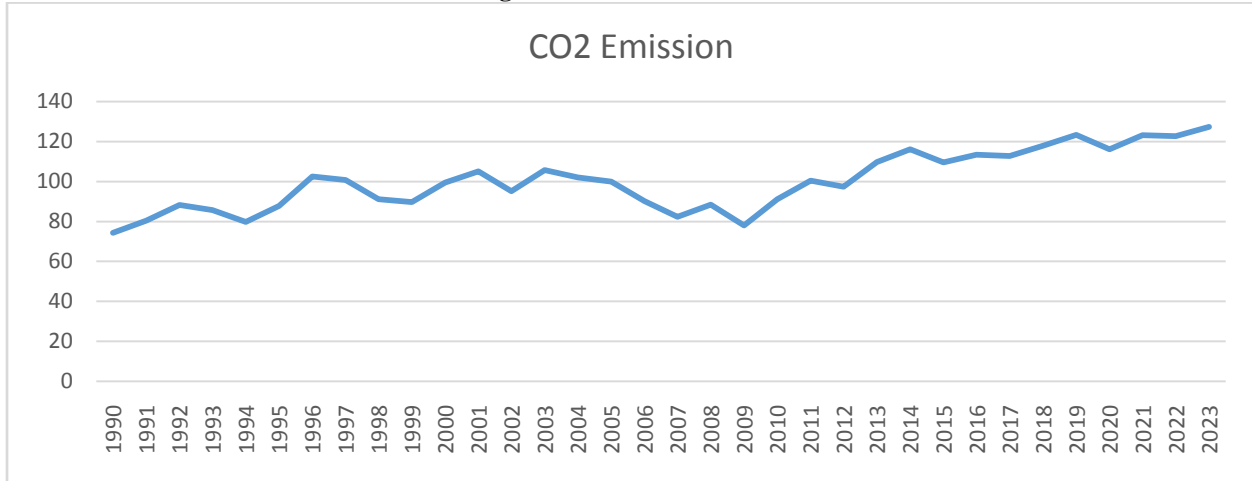
structured as follows; the second section focus on the literature review, the third section addresses the estimation methodologies. The fourth section is concerned with the results and discussion of the findings and the final section presents the conclusion.

#### **CO<sub>2</sub> Emission:**

Mostly byproducts of energy generation and consumption, carbon dioxide emissions contribute the most of the greenhouse gases linked to global warming. Main causes of anthropogenic carbon dioxide emissions are cement production and burning of fossil fuels. For the same level of energy consumption, different fossil fuels emit different amounts of carbon dioxide; coal releases almost twice as much as natural gas and oil releases roughly 50 percent more (WDI, 2025). The average emission rate of a particular pollutant from a given source expressed relative to the intensity of a given activity is known as emission intensity. Comparatively evaluating the environmental impact of various fuels or activities also relies on emission intensities (WDI, 2025). Often used instead are the related words emission factor and carbon intensity. A nation's carbon dioxide emissions tell only one greenhouse gas story. Gases like methane and nitrous oxide should be considered if one wants a more whole picture of how a nation shapes climate change. In agricultural economy especially, this is quite vital. Carbon dioxide's environmental consequences pique great attention. Comprising the most of the greenhouse gases causing global warming and climate change, carbon dioxide (CO<sub>2</sub>) By means of carbon dioxide (or CO<sub>2</sub>) equivalents, all other greenhouse gases (methane (CH<sub>4</sub>), nitrous oxide (N<sub>2</sub>O), hydrofluorocarbons (HFCs), perfluorocarbons (PFCs), sulphur hexafluoride (SF<sub>6</sub>) could be compared and their individual and total contributions to global warming (WDI, 2025) could be ascertain.



Figure 1: CO2 Emission



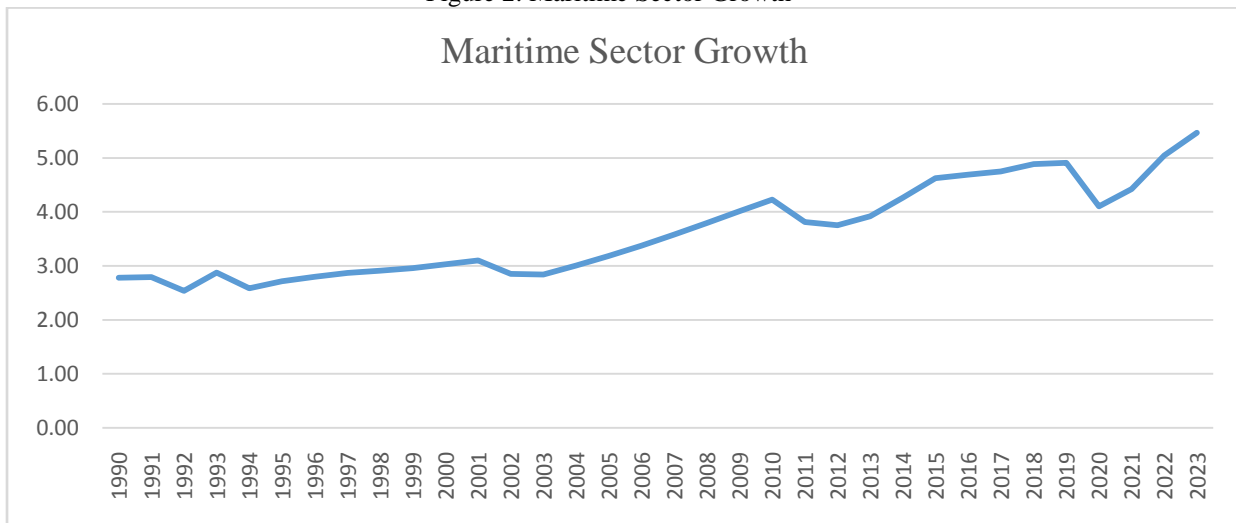
Source: Data from WDI, graphed by Author

### Maritime Sector:

The maritime sector is all industry and activity connected to sea-based trade, transportation, and resources. It covers shipping, ports, shipbuilding, marine engineering, fishing, offshore energy (coal, gas, and renewables), and maritime

tourism. Globally, trade depends on this industry since over eighty percent of commodities are shipped by sea. In environmental sustainability, resource exploitation and economic development it is also an important sector.

Figure 2: Maritime Sector Growth



Source: Data from WDI, graphed by Author

## II. Literature Review

### 2.1 Theoretical review

The Environmental Kuznets Curve theory provides the theoretical framework to clarify the link among energy, economic growth, and carbon emissions. According to the EKC theory, economic expansion affects the surroundings at every level of development. The theory holds that while the energy consumption and economic activities in the first

stage of economic development endanger the environment, the will will be a turnover when a particular degree of development is reached (Gyimah and Yao, 2022). This theory has been applied in several studies to explain the impact on economic growth and energy use on carbon emissions (Tachega, et al. 2021; Nyantakyi, et al. 2023). Given the industrial revolution period we live in, many nations-especially those in developing



nations-are implementing policies to accommodate the worldwide increase of development. Many nations find themselves in a position to exploit environmentally damaging energy sources under this scenario. Investigating how the kind of energy use supports economic development and environmental quality is relevant given the global struggle against climate change.

## 2.2 Empirical review

Obani et al., (2025) with data from 2005 to 2023 evaluated with the Granger causality and ordinary least square (OLS) technique, examined the effect of environmental rules on foreign direct investment in Nigeria. The results revealed that although foreign direct investment does not environmental regulation, environmental regulation grange causes. FDI and carbon emissions have a bi-directional causality. Although FDI does not grander cause GDP, GDP grander cause is seen to be FDI. Furthermore negatively and statistically insignificantly is environmental regulation's effect on FDI. One finds a negative, minor correlation between FDI and carbon emissions. GDP and FDI have a negative and significant association. With data from 38 SSA countries between 2002 and 2022, Oumarou and Nourou (2024) evaluated the effects of trade openness on carbon dioxide (CO<sub>2</sub>) emissions in Sub-Saharan Africa (SSA) estimated with the Generalised Method of Moments (GMM) system and the double ordinary least squares method. The primary findings revealed that trade openness in SSA helps to explain increasing CO<sub>2</sub> emissions there. Moreover, the results revealed that environmental quality is mostly driven by the consumption of renewable energy, thereby presumably lowering CO<sub>2</sub> emissions. Conversely, population rise, human capital, and the quality of institutions raise CO<sub>2</sub> emissions. Moreover, the interplay of openness with institutional excellence reduces CO<sub>2</sub> emissions. Using the ordinary least squares regression approach to estimate the data between 1990 and 2023, Obani et al., (2024) examined the impact of renewable energy on economic growth especially it empirically evaluated the link between solar power and GDP. Their results revealed that usage of renewable energy-a substitute for solar power-has a negligible and favourable effect on economic growth. Economic growth is favourably and significantly influenced by carbon emissions. Direct foreign investment has a major and favourable impact on economic development. Though it is not statistically significant, trade balance also shown a favourable effect on economic development. Furthermore, whereas carbon

emission and GDP have bidirectional causality, solar power consumption has no statistically significant causal effect on GDP. Using data from 1990 to 2019 projected with Simultaneous equations approach, Bouchoucha (2024) investigated 40 African countries. Trade openness has a positive and significant effect on CO<sub>2</sub> emissions, hence CO<sub>2</sub> emissions have a positive correlation with infant mortality and under-five mortality whereas CO<sub>2</sub> emissions have a negative association with life expectancy.

Using data from 2003 to 2017 projected with the Bayesian technique, Pham & Nguyen (2024) averaged 64 developing nations. There was no evidence of a statistically significant impact of trade openness on environmental pollution in emerging nations. Still, the data contradict the polluted paradise theory. Apart from trade openness, the findings revealed that capital abundance, financial openness, and renewable energy use are main factors influencing environmental quality, thereby seeming to lower CO<sub>2</sub> emissions. Over a span of 150 years projected with rigorous cointegration approaches, Barkat et al. (2024) used 20 Organisation for Economic Cooperation and Development (OECD) countries. Trade liberalisation and CO<sub>2</sub> emissions revealed a positive direct effect; income growth mediated a negative influence in the indirect effect. These different effects help to establish the environmental Kuznets curve theory. Trade opening is clearly a cause of environmental damage in BRICS countries, according to Chhabra et al. (2023), who used data from 1991–2019 and projected with the modern method of dynamic common correlative effects verifying the pollution paradise hypothesis. Goswami et al. (2023) projected with the ARDL model and the random forest model data from 1981–2021. The ARDL model turned out that variables were co-integrated. Short-term correlations between CO<sub>2</sub> emissions and economic development and trade openness are found; energy consumption and urbanisation were positively correlated. Long term, CO<sub>2</sub> emissions are favourably connected with urbanisation and business opening as well as with energy consumption; economic development and CO<sub>2</sub> emissions are likewise favourably correlated.

Omri and Saadaoui (2023) investigated France using data from 1980–2020 projected with the NARDL (Non-linear Autoregressive Distributed Lag). Rising emissions are coming from unrestricted trade and fossil fuels. The study validates the inverted U-curve relating carbon emissions to economic development. Using Granger's non-causality estimate of the mean group of correlated



effects, Ghaderi et al. (2023) examined the Environmental Kuznets Curve Hypothesis in the MENA Region from 1995–2018. Main causes of CO<sub>2</sub> emissions are trade openness and energy usage. Ogundipe, (2020) looked at the impact of fossil fuel consumption on environmental quality in Nigeria. The results also revealed that although first-generation estimators confirmed the Environmental Kuznets curve (EKC) hypothesis, there is no inverted U-shaped association between economic progress and CO<sub>2</sub> emissions. The paper applies the Johansen co-integration technique using secondary data from 1970 to 2017. Their findings showed that direct consequence of fossil fuel combustion accounts for almost eighty percent of carbon emissions in Nigeria. They also discovered that pollution was a growing function of income; increases in pollution follow the increasing density of communities during the course of the study. According to Sulaiman et al. (2011), present and future engines will be compelled to adopt new technologies in the future based on their study on air emission from ship driving force for next generation marine technical and regulatory change.

Furthermore, they thought that future technologies will be a main driver of a sequence of research projects aiming at resolving present environmental and energy issues. Such research will produce data that will be used to enforce pertinent rules on climate change control and compliance. Launched recently towards global warming, climate change and ozone depletion control in the maritime industry, the developing technology discussed could help meet the current demand by IMO for the implementation of Energy Efficiency Design Index (EEDI), Ship Energy Efficient Management Plan (SEEMP) and Ship Energy Efficiency Operational Indicator (SEEOI).

### III. Methodology

#### 3.1 Data for the study

The study employs time series data from 1990 to 2023, sourced from the World Development Indicator (WDI) and the CBN statistical bulletin. Table 1 summarizes the information on the variables utilized in the study.

**Table 1:** Description of Variables

| Variables       | Description                                     | Unit of Measurement                                   | Source |
|-----------------|---|---|--------|
| CO <sub>2</sub> | CO <sub>2</sub> Emission                        | CO <sub>2</sub> Emission (MtCO <sub>2</sub> e)        | WDI    |
| MSG             | Maritime contribution to Gross Domestic Product | Maritime GDP (constant 2010 prices)                   | CBN    |
| ENVR            | Environmental Regulation                        | Environmental Sustainability rating (1=low to 6=high) | WDI    |

**Source:** Author's compilation, 2025

#### 3.2 Model of the study

The model of the study is specified by adopting that of Obani et al (2025) and Oumarou and Nourou (2024) with some modification. The functional form of the model is given as equation (1);

$$CO_2 = f(MSG, ENVR) \quad (1)$$

The study utilized the Vector Autoregressive (VAR) Model to analyse the data of the study. A VAR is an n-equation, n-variable linear model where each equation specifies a variable as a function of lagged values of all the n – variables in the model. Thus the VAR model considers all variables to be endogenous and captures the rich dynamics in multiple time series. Assuming there are n-variables

of interest, say,  $y_1, y_2, \dots, y_n$ ; thus the vector of variables in the VAR is  $(y_1, y_2, \dots, y_n)$ .

This implies that all the variables are endogenous. The VAR model is represented in its compact form as follows:

$$V_t = \alpha + \sum_{i=1}^p A_i V_{t-1} + \varepsilon_t \quad (2)$$

Where;

$V_t$  is the vector of CO<sub>2</sub> Emission, Maritime sector growth and Environmental Regulation.  $\alpha$  is the intercepts of autonomous variables.  $A_i$  is the matrix of coefficients of all the variables in the model.  $V_{t-1}$  is the vector of the lagged variables and  $\varepsilon_t$  is the vector of the stochastic error terms.



#### IV. Results and Discussion of Findings

**Table 2:** Descriptive Statistics

|              | CO2      | ENVR      | MSG      |
|--------------|----------|-----------|----------|
| Mean         | 100.2622 | 2.470588  | 3.630177 |
| Median       | 100.2112 | 3.000000  | 3.476799 |
| Maximum      | 127.3461 | 3.500000  | 5.463231 |
| Minimum      | 74.38070 | 0.000000  | 2.538185 |
| Std. Dev.    | 14.60043 | 1.408530  | 0.847494 |
| Skewness     | 0.126290 | -1.160937 | 0.479670 |
| Kurtosis     | 2.002427 | 2.479776  | 1.938457 |
| Jarque-Bera  | 1.500178 | 8.020785  | 2.900211 |
| Probability  | 0.472324 | 0.018126  | 0.234546 |
| Observations | 34       | 34        | 34       |

Source: Author's computation, 2025

The descriptive statistics is presented in table 2. From the information presented, the average CO<sub>2</sub> emission (CO2) for the study period stood at 100mt, environmental regulation (ENVR) was 2.47 percent and maritime sector growth (MSG) was 3.63 billion naira. Environmental regulation and

maritime sector growth are positively skewed while CO<sub>2</sub> emission is negatively skewed based on their skewness values. Their kurtosis values showed that all the variables are platykurtic distribution. Their Jarque-Bera probability value shows that only CO<sub>2</sub> and MSG are normally distributed.

**Table 3:** ADF unit root test

| Unit Root Test at Levels           |                    |         |                |
|------------------------------------|--------------------|---------|----------------|
| Variable                           | ADF Test Statistic | P-Value | Remark         |
| CO2                                | -1.480758          | 0.5307  | Non-Stationary |
| ENVR                               | -1.854951          | 0.3487  | Non-Stationary |
| MSG                                | 0.207992           | 0.9690  | Non-Stationary |
| Unit Root Test at First Difference |                    |         |                |
| Variable                           | ADF Test Statistic | P-Value | Remark         |
| CO2                                | -6.373444          | 0.0000  | Stationary     |
| ENVR                               | -5.660836          | 0.0000  | Stationary     |
| MSG                                | -4.342348          | 0.0018  | Stationary     |

Source: Author's computation, 2025

The result of the Augmented Dickey-Fuller Test is presented in table 3, from the result we observed that all the variables are stationary at first difference I(1). Therefore, we proceed to conduct the Johansen cointegration test to check for the existence of a long run relationship among the variables.

**Table 4:** Cointegration test

| Unrestricted Cointegration Rank Test (Trace) |            |                 |                     |         |
|--|------------|-----------------|---------------------|---------|
| Hypothesized No. of CE(s)                    | Eigenvalue | Trace Statistic | 0.05 Critical Value | Prob.** |
| None   | 0.235714   | 14.35612        | 29.79707            | 0.8200  |
| At most 1                                    | 0.164497   | 5.754106        | 15.49471            | 0.7244  |
| At most 2                                    | 9.49E-05   | 0.003037        | 3.841465            | 0.9544  |

Trace test indicates no cointegration at the 0.05 level

\* denotes rejection of the hypothesis at the 0.05 level



\*\*MacKinnon-Haug-Michelis (1999) p-values

Unrestricted Cointegration Rank Test (Maximum Eigenvalue)

| Hypothesized<br>No. of CE(s) | Eigenvalue | Max-Eigen<br>Statistic | 0.05<br>Critical Value | Prob.** |
|------------------------------|------------|------------------------|------------------------|---------|
| None                         | 0.235714   | 8.602016               | 21.13162               | 0.8631  |
| At most 1                    | 0.164497   | 5.751069               | 14.26460               | 0.6454  |
| At most 2                    | 9.49E-05   | 0.003037               | 3.841465               | 0.9544  |

Max-eigenvalue test indicates no cointegration at the 0.05 level

\* denotes rejection of the hypothesis at the 0.05 level

\*\*MacKinnon-Haug-Michelis (1999) p-values

Source: Author's computation, 2025

From the results of the multivariate cointegration test presented in Table 4, the Johansen's cointegration technique revealed that both the trace statistic and maximum Eigen value statistic confirm no existence of a cointegrating equations among the variables. Since the variables are not cointegrated, the existence of a stable long run relationship between the variable can only be confirmed Vector Autoregressive (VAR) model.

**Table 5: Vector Autoregression Estimates**

Sample (adjusted): 3 34  
 Included observations: 32 after adjustments  
 Standard errors in ( ) & t-statistics in [ ]

|          | CO2                                  | ENVR                                 | MSG                                  |
|----------|--------------------------------------|--------------------------------------|--------------------------------------|
| CO2(-1)  | 0.681655<br>(0.20299)<br>[ 3.35800]  | -0.000465<br>(0.01446)<br>[-0.03215] | -0.002184<br>(0.00763)<br>[-0.28628] |
| CO2(-2)  | 0.006536<br>(0.20402)<br>[ 0.03204]  | 0.016040<br>(0.01453)<br>[ 1.10361]  | 0.007551<br>(0.00767)<br>[ 0.98482]  |
| ENVR(-1) | -0.399718<br>(2.81602)<br>[-0.14194] | 0.775217<br>(0.20061)<br>[ 3.86423]  | -0.011972<br>(0.10584)<br>[-0.11312] |
| ENVR(-2) | 0.482437<br>(2.72539)<br>[ 0.17702]  | 0.025243<br>(0.19416)<br>[ 0.13001]  | 0.052089<br>(0.10243)<br>[ 0.50853]  |
| MSG(-1)  | -2.274957<br>(5.41815)<br>[-0.41988] | -0.036832<br>(0.38599)<br>[-0.09542] | 1.023000<br>(0.20363)<br>[ 5.02370]  |
| MSG(-2)  | 6.953218<br>(5.50988)<br>[ 1.26195]  | -0.014193<br>(0.39253)<br>[-0.03616] | -0.137585<br>(0.20708)<br>[-0.66440] |
| C        | 16.22141<br>(11.5609)                | -0.740028<br>(0.82360)               | -0.135551<br>(0.43450)               |



|                |            |            |            |
|----------------|------------|------------|------------|
|                | [ 1.40313] | [-0.89853] | [-0.31197] |
| R-squared      | 0.760874   | 0.863385   | 0.910030   |
| Adj. R-squared | 0.703484   | 0.830597   | 0.888438   |
| F-statistic    | 13.25790   | 26.33262   | 42.14529   |
| Akaike AIC     | 7.063263   | 1.779867   | 0.500901   |
| Schwarz SC     | 7.383893   | 2.100497   | 0.821531   |

**Source:** Author's computation, 2025

In Table 5 the Vector Autoregression Estimates are presented. All the coefficient estimates are elasticities. A careful study of the results shows that the single most important determinant of each variable is its one-period lagged value. The elasticity of CO<sub>2</sub> emission with respect to its lagged value is 0.68; the elasticity of environmental regulation with respect to its

lagged value is 0.78 and the elasticity of maritime sector growth with respect to its lagged value is 1.02. These results coincide with those of IRF and Variance Decomposition analyses. The variables of the model also have a high explanatory value depicted by the R-squared value of 0.760874 which shows a 76% explanatory power.

**Table 6:** Diagnostics of the Vector Autoregression Estimates  
 VAR Residual Serial Correlation LM Tests  
 Null hypothesis: No serial Correlation n at lag h

| Lag | LRE* stat | df | Prob.  | Rao F-stat | df         | Prob.  |
|-----|-----------|----|--------|------------|------------|--------|
| 1   | 4.766780  | 9  | 0.8541 | 0.517426   | (9, 48.8)  | 0.8548 |
| 2   | 7.628200  | 9  | 0.5720 | 0.851440   | (9, 48.8)  | 0.5736 |
| Lag | LRE* stat | df | Prob.  | Rao F-stat | df         | Prob.  |
| 1   | 4.766780  | 9  | 0.8541 | 0.517426   | (9, 48.8)  | 0.8548 |
| 2   | 14.76325  | 18 | 0.6782 | 0.804623   | (18, 48.6) | 0.6854 |

\*Edgeworth expansion corrected likelihood ratio statistic.

**Source:** Author's computation, 2025

**VAR Residual Heteroskedasticity Tests (Levels and Squares)**

Joint test:

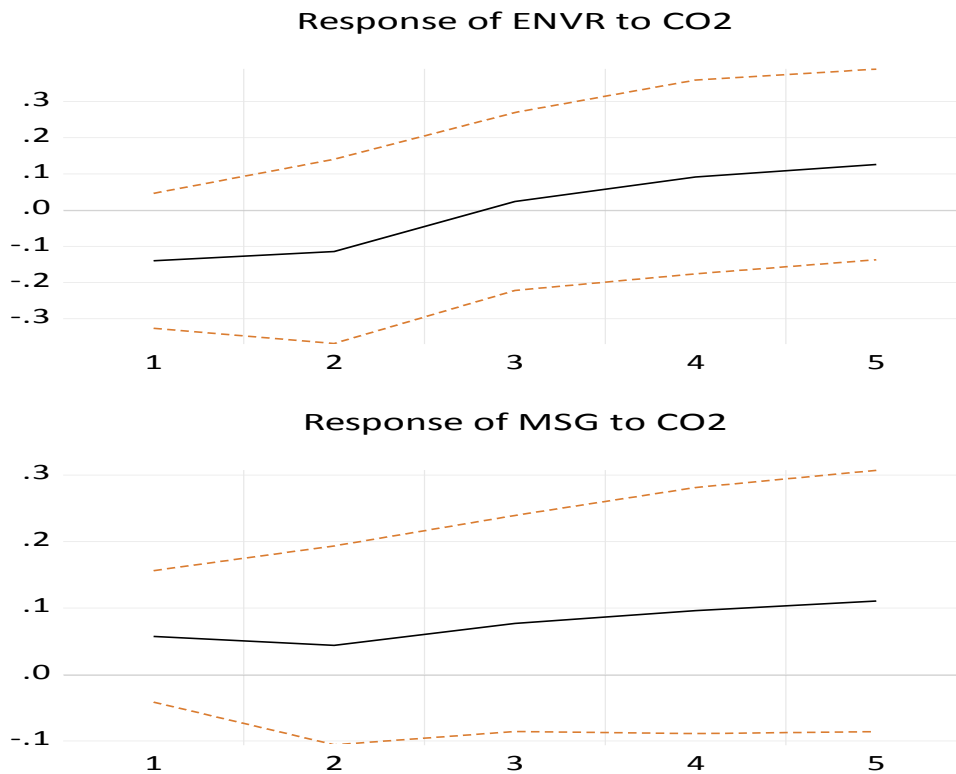
| Chi-sq   | Df | Prob.  |
|----------|----|--------|
| 69.49009 | 72 | 0.5619 |

**Source:** Author's computation, 2025

The model errors are free from serial correlation based on the VAR Residual Serial Correlation LM Tests result. The VAR Residual Heteroskedasticity Tests shows that the errors are homoskedastic.



**Figure 3: Impulse Response Function**  
 Response to Cholesky One S.D. (d.f. adjusted) Innovations  $\pm 2$  S.E.



Source: Author's computation, 2025

The impulse response function of environmental regulation and maritime sector growth to the one standard deviation innovations of CO<sub>2</sub> emission. The Impulse Response Function result showed that there is no initial response from the maritime sector growth to CO<sub>2</sub> emission but it gradually increased over the study period and

maintaining a positive effect both in the long run and short run. Environmental regulation in the short run exhibit a negative impact, it reached a steady state and start to rise. Thus, it has an asymmetric effect on CO<sub>2</sub> emission both in the long run and short run.

**Table 7: Variance Decomposition**

| (i)Variance Decomposition of CO <sub>2</sub> : |          |                 |          |          |
|--|----------|-----------------|----------|----------|
| Period   | S.E.     | CO <sub>2</sub> | ENVR     | MSG      |
| 1  | 7.518549 | 100.0000        | 0.000000 | 0.000000 |
| 2  | 9.081792 | 99.46617        | 0.053381 | 0.480454 |
| 3  | 9.869800 | 98.79641        | 0.046333 | 1.157257 |
| 4  | 10.41300 | 95.31941        | 0.042641 | 4.637952 |
| 5  | 10.95060 | 89.98365        | 0.043241 | 9.973106 |

| (ii)Variance Decomposition of ENVR: |          |                 |          |          |
|-------------------------------------|----------|-----------------|----------|----------|
| Period                              | S.E.     | CO <sub>2</sub> | ENVR     | MSG      |
| 1                                   | 0.535623 | 6.831839        | 93.16816 | 0.000000 |
| 2                                   | 0.678685 | 7.083315        | 92.89413 | 0.022551 |
| 3                                   | 0.752738 | 5.858245        | 94.03830 | 0.103453 |



| 4                                   | 0.801795 | 6.462520 | 93.18337 | 0.354113 |
|-------------------------------------|----------|----------|----------|----------|
| 5                                   | 0.837861 | 8.180499 | 91.35774 | 0.461757 |
| (iii)Variance Decomposition of MSG: |          |          |          |          |
| Period                              | S.E.     | CO2      | ENVR     | MSG      |
| 1                                   | 0.282576 | 4.106490 | 0.002436 | 95.89107 |
| 2                                   | 0.402398 | 3.211644 | 0.015212 | 96.77314 |
| 3                                   | 0.481810 | 4.779814 | 0.142930 | 95.07726 |
| 4                                   | 0.536584 | 7.067765 | 0.518348 | 92.41389 |
| 5                                   | 0.579955 | 9.688499 | 1.074356 | 89.23714 |
| Cholesky Ordering: CO2 ENVR MSG     |          |          |          |          |

Source: Author's computation, 2025

The forecast error variance decomposition was employed to further examine the short run dynamic properties of maritime sector growth, environmental regulation and CO<sub>2</sub> emission in Nigeria. The results for the three variables are presented in Table 4. An examination of the variance decomposition of CO<sub>2</sub> emission in Table4(i)shows that the lion's share of the variation experienced by CO<sub>2</sub> emission is attributed to its own shock. The contribution of own shock is 100% in the first period and declines slightly to 89.9% at the end of the 5-period horizon. The other 2 variables contributed marginally to it. The highest been maritime sector growth, which contributes 9.97% in the fifth period.

In Table 4(ii) environmental regulation exhibited a strong endogenous impact on its own shocks. The contribution of own shock is 93.1% in the first period but declines to 91.3% in the fifth period. The contribution of the other variables shows a strong exogenous influence with environmental regulation accounting. CO<sub>2</sub> emission have the highest among them contributing 8.1%, showing that environmental regulation as essentially driven by CO<sub>2</sub> emission.

In Table 4(iii) we observed that CO<sub>2</sub> emission have a strong exogenous impact on maritime sector growth both in the short run and long run. In the fifth period, CO<sub>2</sub> emission accounted for 9.6% of variation in maritime sector growth in the long run. Thus, displaying similar results as the impulse response function.

## V. Conclusion

This study investigates maritime sector growth, environmental regulation and CO<sub>2</sub> emission nexus in Nigeria from 1990 to 2023. The study concludes that the single most important determinant of each variable is its one-period lagged value. The Impulse Response Function and forecast error variance decomposition showed that the maritime

sector growth and CO<sub>2</sub> emission exhibit a positive effect both in the long run and short run. Environmental regulation have an asymmetric effect on CO<sub>2</sub> emission both in the long run and short run.

Based on the findings the government should put in place stiffer and more enforceable environmental rules will help to reduce CO<sub>2</sub> emissions from the maritime industry, so guaranteeing their efficacy in the long run as well as in the present. To lower emissions, support the marine industry's adoption of sustainable practices and greener technology like alternative fuels and energy-efficient ships. Create policies to strike a balance between environmental sustainability and the expansion of the maritime industry so as to prevent disproportionate rise of CO<sub>2</sub> emissions. Encourage innovation to decouple sector development from environmental damage by supporting research into sustainable maritime practices and emissions reduction solutions.

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