




# THE FINANCIAL DEVELOPMENT-ENVIRONMENTAL DEGRADATION NEXUS IN BANGLADESH: A LONG-RUN CO-INTEGRATION APPROACH



 **Mahmuda Akter Khuky**

*School of Business and Economics Universiti Putra Malaysia (UPM), UPM Serdang Selangor Darul Ehsan Malaysia.  
Email: [mahmudaakterkhuky@yahoo.com](mailto:mahmudaakterkhuky@yahoo.com) Tel: +60142728446*



## ABSTRACT

### Article History

Received: 19 March 2021

Revised: 23 April 2021

Accepted: 25 May 2021

Published: 17 June 2021

### Keywords

Financial development  
Carbon dioxide emissions  
Ecological footprint  
Economic growth  
FMOLS  
DOLS  
Bangladesh.

### JEL Classification:

C1; C10; G20; Q53.

Using annual time series data from 1983 to 2019, this study examines the impact of economic growth, exports, energy consumption, and financial development on environmental deterioration in Bangladesh. The long-run cointegration of variables is investigated using Dynamic Ordinary Least Squares (DOLS) and Fully Modified OLS (FMOLS) approaches. The co-integration study demonstrates a long-term relationship between dependent variable and independent variables. The findings reveal that in the long run the primary determinants of carbon dioxide emissions are economic growth, energy consumption, export, and financial development are the primary determinants of carbon dioxide emissions. Except for energy consumption, the result shows that financial development, economic development, and export have a statistically significant long-run co-integration with environmental degradation (CO<sub>2</sub> emissions and ecological footprint). The Environmental Kuznets Curve, a quadratic term for economic expansion, demonstrates a negative impact on environmental degradation (EKC). The key findings show and emphasize the need for sound policies for more equal economic and financial growth, as well as, environmentally sustainable financial services. According to the findings, the government should prioritize programs that reduce carbon dioxide emissions by strengthening the financial sectors. Also government should consider the role it plays in slowing environmental degradation and hence, directly enhancing environmental quality in Bangladesh.

**Contribution/ Originality:** This study contributes to the existing works of literature by adding evidence that financial development is responsible for environmental degradation. The study is one of the few studies in the Bangladesh context that examine the effect of financial development on environment quality using both CO<sub>2</sub> emission and ecological footprint measures.

## 1. INTRODUCTION

Global warming is currently one of the most critical concerns affecting industrialized, emerging, and developing countries. As a result of the effects of degrading the environment's quality, The depletion of ozone layer has been proven in the literature as a term that refers to carbon dioxide (CO<sub>2</sub>) emissions as the primary contributor to global warming (Alam, Begum, Buysse, & Van Huylbroeck, 2012). Since the 1900s, global greenhouse gas (GHG) emissions from fossil fuels have grown dramatically. CO<sub>2</sub> emissions are known for their influence on environmental deterioration as many different forms of pollution. Several factors have been linked to CO<sub>2</sub> emissions, including a country's income level, financial development, use of renewable energy, and technical innovation, human capital, and gross domestic product (GDP). Several recent studies have found a link between

financial development, energy use, and pollution in diverse locales and countries. Tang and Tan (2014) and Islam, Shahbaz, Ahmed, and Alam (2013) in Malaysia, Zhang (2011) in China, and Coban and Topcu (2013) in the European Union have all shown direct links between energy use and pollution (EU).

This implies that economic growth and energy use exacerbate environmental deterioration. Other research, notably (Tamazian & Rao, 2010) across transition nations, Shahbaz, Khan, and Tahir (2013) in Indonesia, and Jalil and Feridun (2011) in China, showed the opposite conclusion, namely that financial growth improves environmental quality by reducing carbon dioxide emissions. As a result, Previous research has connected them to both economic development and foreign direct investment (Costantini & Monni, 2008; Inglesi-Lotz & Dogan, 2018; Zhang, Peng, Ma, & Shen, 2017).

Bangladesh, a small South Asian country, has become one of Asia's best performers over the previous decade, with an average annual economic growth rate of more than 6%. It is a developing country with a large population and good demographics, with promising future progress in the economy. As a result of the constant economic growth, the electricity demand has been rising at an alarming rate in the last few years. Considering the environmental degradation is a worldwide issue, it is critical to examine the factual links between FD and environmental degradation from the standpoint of individual countries. The graph below Figure 1 shows that when financial development increases, CO<sub>2</sub> emissions grow as well.

The objective of this article is to look at the impact of exports, GDP per capita, financial development, and energy consumption on CO<sub>2</sub> emissions in Bangladesh using the Environmental Kuznets Curve (EKC) paradigm. This study, which uses ecological footprint as a measure of the environment, gives a better knowledge of environmental deterioration. Several studies examine the relationship between FD, energy, development, and emissions to produce better policy solutions for managing both development and the environment at the same time (Majeed, Samreen, Tauqir, & Mazhar, 2020). These studies provide diverse policy suggestions for different nations and locations based on inconclusive results. These studies have a fundamental flaw in that they use CO<sub>2</sub> emissions as a measure of environmental damage (Bekhet, Matar, & Yasmin, 2017; Maji, Habibullah, & Saari, 2017; Zakaria & Bibi, 2019). However, carbon emissions are only one aspect of the environmental damage produced by large-scale energy usage (Al-Mulali & Sheau-Ting, 2014).

The advantage of using EFP, on the other hand, is that it combines environmental data into a single measure that can be easily compared to the related productive capacity, emphasizing how much of the ecosystem surface we use for survival. The demand for natural resources is growing as the world's population grows. As a result, securing the long-term use of natural resources has turned into a worldwide concern. As a result, knowing the degree of natural resource usage is critical in order to protect them for future generations. Many studies consider the EFP to be a more complete and trustworthy indication of environmental impact (Al-Mulali & Sheau-Ting, 2014; Uddin, Salahuddin, Alam, & Gow, 2017). For the following reasons, this research recommends EFP as a comprehensive measure of environmental deterioration.

The implication of this study if financial institutions give loans and concessions to industries that embrace green technology and promote environmentally friendly initiatives, the overall environmental impact can be reduced. The contribution of this study is it will provide a valuable insight understanding of FD as a potential predictor of environmental degradation including its both measures CO<sub>2</sub> emission and ecological footprint. Thus, this study will also contribute to the FD-environment nexus literature. Also, no previous research on these factors in Bangladesh, thus the current study informs academia and policymakers about the current state of dynamic linkages among exports, financial development, energy consumption, economic growth, and environmental quality in Bangladesh. The remainder of the study is structured as follows: the following The review of related literature is presented in Section 2. The data, methods, and statistical analysis are discussed in Section 3. Section 4 presents the empirical findings as well as their interpretation. Section 5 shows the conclusion and makes policy suggestions.

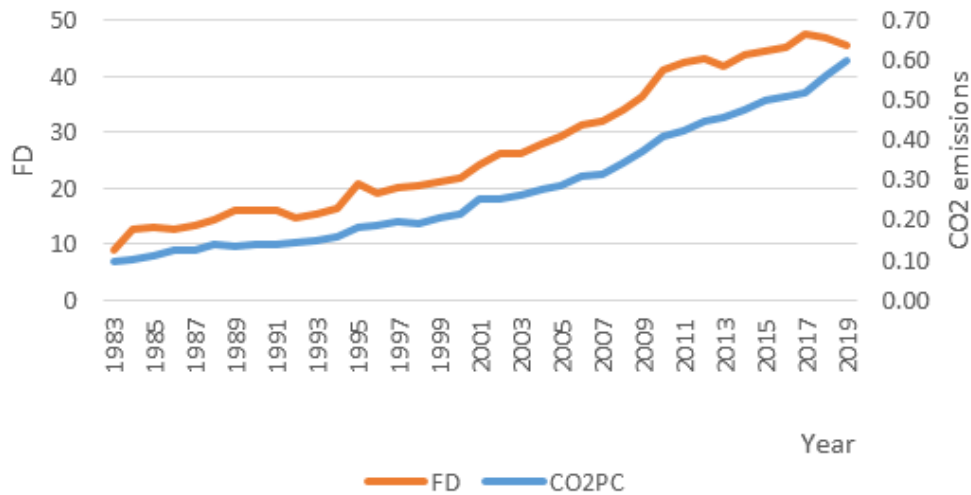


Figure-1. Financial development and CO2 emission of Bangladesh.

Source: WDI (2020).

## 2. REVIEW OF RELATED LITERATURE

Environmental degradation is the result of a variety of human actions that have resulted in a decline in environmental quality due to natural resource depletion, species expansion, weather fluctuations, and ecosystem loss (Majeed & Mazhar, 2019). Environmental deterioration has become a major focus of academic research in recent decades. As a result, both environmental ideas and empirical evidence have received a lot of attention and controversy. Environmental Kuznets Curve (EKC) theory provides the theoretical underpinning for this paper. According to EKC, environmental degradation rises in the early stages of economic growth, but at a certain degree of economic growth, society begins to ameliorate its connection with the environment. Grossman and Krueger (1995) were the first to observe the inverted U-shaped relationship between environmental pollutants and per capita income and introduced the name EKC. The empirical evidence strongly supports the existence of EKC in both developed and developing countries (Grossman & Krueger, 1995). The pioneering research on the financial development and environment nexus was began by Aufderheide and Rich (1988). They said that the World Bank's financial aid system frequently overlooks the environmental consequences of loanable taxes, resulting in major environmental issues. Furthermore, several research that used time-series data came to the same result. For example, Moghadam and Dehbashi (2018) observed negative consequences of financial development in Iran, Sehrawat, Giri, and Mohapatra (2015) in India, Mesagan and Nwachukwu (2018) in Nigeria, and Raza and Shah (2018) in Pakistan.

The influence of financial development (FD) on environmental deterioration has been noted in the literature (Destek & Sarkodie, 2019; Katircioglu & Katircioglu, 2018; Sarkodie & Strezov, 2019; Zhang, 2011; Zhang & Zhang, 2018). Zhang (2011) emphasized the importance of a well-developed institution in mitigating environmental stress (ecological footprint). Through research and development, financial development promotes environmental quality by attracting and bringing more environmentally friendly ideas to market. It may also make it easier to invest in renewable energies. Investment in renewable energy, example, is increasing, which is more environmentally friendly and helps to lessen the ecological impact (EFP). FD, on the other hand, can boost economic efficiency by minimizing capital risk and financial costs. In addition, through extending FDI inflows, stock market operations, and banking operations, FD may boost R&D and investment in clean technologies. FD, on the other hand, contributes to environmental damage by encouraging and enabling the use of credit to buy mechanical machinery, electrical equipment, vehicles, and residences. These facilities assist investors in broadening their company scopes and establishing new machinery and factories, which increases carbon emissions in the environment and deteriorates environmental quality (Zhang & Zhang, 2018).

Furthermore, as a result of a well-developed financial system, an increase in foreign direct investment adds to environmental damage (Sarkodie & Strezov, 2019). Based on the ARDL methodology Using quarterly data, Ali et al. (2019) explore the effects of financial development on energy usage in Nigeria between 1972Q1 and 2011Q4. According to the findings, as the null hypothesis is, variables are cointegrated. At the 1% threshold of significance, the hypothesis was rejected. The long-term outcome demonstrates that the financial impact is significant. The trend in energy usage is downwards, and while economic development has a detrimental impact, it is minimal even though it has a major influence on energy usage and energy prices have been positive and considerable for quite some time usage of energy. Farhani and Ozturk (2015) investigate taking data from 1971-2012 the causal linkage between CO<sub>2</sub> emissions, real GDP, energy consumption, financial development, trade openness, and urbanisation in Tunisia. They also used the distributed lag (ARDL) bounds testing methodology to assess the cointegration and error correction methods (ECM). The key empirical conclusion is that the variable of financial development has a positive and significant influence on pollution levels. This suggests that Tunisia's financial industry is growing at the expense of the country's environmental degradation. Salahuddin, Alam, Ozturk, and Sohag (2018) employed Kuwait data from 1980 to 2013 to use an autoregressive distributed lag (ARDL) bounds testing methodology. The purpose is to investigate how economic development, energy consumption, FDI, and financial development impact environmental quality. According to the core research findings, economic development, electricity consumption, and foreign direct investment all raise carbon dioxide emissions in the short and long term.

The mixed results of previous studies on this topic are the motivation to investigate this issue in the context of Bangladesh. In both the short and long periods, Alam et al. (2012) and Ghosh, Alam, and Osmani (2014) found evidence of one-way causation running from energy use to financial development and energy consumption to CO<sub>2</sub> emissions. Uddin and Wadud (2014) analyzed seven SAARC nations and discovered that the variables of CO<sub>2</sub> emissions and economic growth had a co-integration connection. CO<sub>2</sub> emissions have a long-term favorable and considerable influence on economic growth. According to Alam et al. (2012), in the short run, unidirectional causation occurs between energy consumption and economic growth, however, in the long run, bidirectional causation exists between electricity consumption and economic growth. Nonetheless, Alam, Uddin, and Islam (2017) identified a causal association between energy consumption and CO<sub>2</sub> emissions, CO<sub>2</sub> emissions, and economic development in the short run, but no long-term cointegrating association in Bangladesh. According to the research of Bangladesh utilizing data from 1985 to 2013, bidirectional causation exists between the variables of financial development, power consumption, and fossil fuel energy consumption with economic growth in the short term. Therefore, financial progress is strongly linked to environmental quality, and vice versa. The greater the amount of energy consumed, the greater the chance of more carbon emissions will lead to a worsening of the situation increased pollution, and, as a result, a reduction in environmental quality.

### 3. EMPIRICAL MODEL AND METHODOLOGY

Since the study has used two different measures (CO<sub>2</sub> and Ecological footprint) of environment degradation. These two measures are run by separate models. Equation 1 is for CO<sub>2</sub> emissions model and Equation 2 is for the ecological footprint model. The log-linear regression model to gain the objective of this study is follows:

$$\ln CO2_t = \alpha_0 + \alpha_1 \ln EX_t + \alpha_2 \ln GDP_t + \alpha_3 \ln GDP_t^2 + \alpha_4 \ln FD_t + \alpha_5 \ln EU_t + \mu_t \quad (1)$$

$$\ln EFP_t = \alpha_0 + \alpha_1 \ln EX_t + \alpha_2 \ln GDP_t + \alpha_3 \ln GDP_t^2 + \alpha_4 \ln FD_t + \alpha_5 \ln EU_t + \mu_t \quad (2)$$

Firstly, this study has done the Unit root tests to check the unit root property of the variables by employing test Augmented Dickey-Fuller (ADF) and Phillips-Perron (PP) tests to know the stationary quality of the variables. the ADF and PP test was carried out based on the following model:

$$\Delta X_t = \alpha_0 + \gamma X_{t-1} + \beta_t + \sum_{i=1}^k \beta_i \Delta X_{t-1} + \epsilon_t$$

Here,  $\Delta$  represents the difference operator,  $t$  portrays the time trend,  $\epsilon$  denotes the error expression termed as a white noise error, signifies the series and  $k$  shows the number of lags used.

The Akaike Information Criterion (AIC) or the Schwarz Bayesian Criterion can be used to determine the optimal lag duration (SBC). The obtained F-statistic is compared to the critical limits established by Pesaran, Shin, and Smith (2001) or Narayan (2005) the upper critical bound and lower critical bound in the first phase of the bounds testing methodology. Several diagnostic tests, such as autocorrelation, functional form, normality of error term, and heteroskedasticity, can be used to assess the model's robustness. Brown, Durbin, and Evans (1975) proposed the cumulative sum of recursive residuals (CUSUM) and cumulative sum of squares of recursive residuals (CUSUMsq) tests to assess the parameters' stability.

This study used the fully modified OLS (FMOLS) and dynamic OLS estimators to further assess the robustness of the aforesaid ARDL models (DOLS). The literature has proposed several long-term estimated coefficient models, including Fully Modified OLS (FMOLS) and Dynamic OLS (DOLS). Long-term carbon emissions, economic growth, energy consumption, financial developments, and export in FMOLS and DOLS models are all included in the study.

The FMOLS model coefficient estimates are consistent with relatively small samples and also control the endogenous problem among retro-generators. Based on the theoretical framework of study Altıntaş and Kassouri (2020) and Nathaniel (2020) who used FMOL and DOLS taking the same variables in their analysis, this study following the FMOLS estimator is shown in Equation 3:

$$\beta_i^* = (X_i' X_i)^{-1} (X_i' y_i^* - T\delta) \quad (3)$$

The dynamic OLS (DOLS), which means that the co-integrating regression is increased by lags and leads. The following equation is used to obtain the Dynamic OLS model is shown in Equation 4:

$$Y_{it} = \alpha_i + X_{it}' \beta + \sum_{j=-q_1}^{j=q_2} C_{ij} \Delta X_{it+j} + v_{it}' \quad (4)$$

In the above-mentioned equation  $X$  indicates all the independent variables i.e. EX, GDP, FD, EU, and OP.  $C_{ij}$  indicates the coefficients for the lead or lag of first differenced independent variables. Hence, the estimated coefficients using DOLS is given in Equation 5:

$$\hat{\beta}_{DOLS} = \sum_{i=1}^N \left( \sum_{t=1}^T Z_{it} Z_{it}' \right)^{-1} \left( \sum_{t=1}^T Z_{it} \hat{y}_{it}^+ \right) \quad (5)$$

Where  $z_{it} = [X_{it} - \hat{X}_{it}, \Delta X_{it-q}, \dots, \Delta X_{it+q}]$  is a vector of regressors, and  $\hat{y}_{it}^+ (\hat{y}_{it}^+ = y_{it} - \hat{y}_{it})$  is the carbon emission variable.

### 3.1. Data

The sample period starts in 1983 and ends in 2019. The time-series data has been included. The variables names and their sources are shown in Table 1. The ecological footprint data is taken from Global Footprint

Network. The remaining data such as CO<sub>2</sub> emission, exports, GDP, energy consumption, and financial development are obtained from World Bank World Development Indicators. All the variables are transferred into a natural logarithm.

Table-1. Variables, measurements and data sources.

| Variables  | Measurement                                      | Source                       |
|--|--|------------------------------|
| Carbon dioxide emissions ( <i>CO<sub>2</sub></i> ) | Carbon emission (metric tons per capita (kt))    | World Development Indicators |
| Ecological footprint ( <i>EFP</i> )                | Ecological footprint per capita (hectares (gha)) | Global Footprint Network     |
| Export ( <i>EX</i> )                               | Total merchandise export in USD                  | World Development Indicators |
| Gross domestic product per capita ( <i>GDP</i> )   | Gross domestic product per capita (current US\$) | World Development Indicators |
| Financial development ( <i>FD</i> )                | Domestic credit to the private sector (% of GDP) | World Development Indicators |
| Energy consumption ( <i>EU</i> )                   | Fossil fuels consumption (% of total energy)     | World Development Indicators |

Source: WDI (2020).

Table 2 shows the descriptive statistics of the variables those are incorporated in the current study. The statistics show that the minimum value of ecological footprint is 0.066 while the maximum value for export is 9828.257.

Table 3 shows the correlation between dependent variable along with other explanatory variables. All variables are positively correlated with each other.

Table-2. Descriptive statistic.

| Description  | CO <sub>2</sub> | EFP   | EX       | GDP      | EU      | FD     |
|--------------|-----------------|-------|----------|----------|---------|--------|
| Mean         | 0.278           | 0.491 | 9828.257 | 635.330  | 162.012 | 27.046 |
| Median       | 0.250           | 0.466 | 4859.830 | 541.291  | 153.560 | 24.179 |
| Maximum      | 0.600           | 0.603 | 34133.27 | 1287.822 | 229.246 | 47.583 |
| Minimum      | 0.095           | 0.409 | 31.570   | 378.092  | 107.672 | 9.034  |
| Std. Dev.    | 0.149           | 0.066 | 10644.79 | 258.009  | 42.959  | 12.340 |
| Observations | 37              | 37    | 37       | 37       | 37      | 37     |

Table-3. Pair-wise correlations.

| Variable          | LNCO <sub>2</sub> | LNEFP  | LNEX   | LNGDP  | LNEU   | LNFD |
|-------------------|-------------------|--------|--------|--------|--------|------|
| LNCO <sub>2</sub> | 1                 |        |        |        |        |      |
| LNEFP             | 0.9100            | 1      |        |        |        |      |
| LNEX              | 0.9470            | 0.7653 | 1      |        |        |      |
| LNGDP             | 0.9789            | 0.9455 | 0.8893 | 1      |        |      |
| LNEU              | 0.9924            | 0.9260 | 0.9306 | 0.9765 | 1      |      |
| LNFD              | 0.9914            | 0.8821 | 0.9629 | 0.9527 | 0.9857 | 1    |

Note: Correlation analysis of the variables shown represents their natural logarithmic values.

## 4. ESTIMATION RESULTS

### 4.1. Unit Root Test

To determine the order of integration, it is important to estimate the unit-roots test using Augmented Dickey Fuller (ADF) and Phillips Perron (PP) for all variables. The results of the ADF and PP are shown in Table 4. To fulfil the requirement all variables must be stationary either at level I(0) or first difference I(1) to apply the long-run cointegration methods (Pesaran et al., 2001). All the variables are stationary at first difference which means all variables are I(1).



Table-4. ADF and PP unit root tests.

| Variables        | Models   | Augmented Dickey Fuller Test (ADF) |                            | Phillips Perron Test (PP) |                            | Order of Integration |
|------------------|----------|------------------------------------|----------------------------|---------------------------|----------------------------|----------------------|
|                  |          | Level                              | 1 <sup>st</sup> Difference | Level                     | 1 <sup>st</sup> Difference |                      |
| LCO <sub>2</sub> | Constant | 3.8407<br>(1.0000)                 | -4.1592<br>(0.0025)***     | 7.1920<br>(1.0000)        | -4.0956<br>(0.0030)***     | I(1)                 |
|                  | Trend    | -0.0596<br>(0.9937)                | -6.0575<br>(0.0001)***     | 1.3037<br>(0.9999)        | -7.8937<br>(0.0001)***     |                      |
| LEFP             | Constant | 1.0775<br>(0.9965)                 | -4.7468<br>(0.0006)***     | 0.8214<br>(0.9929)        | -4.8275<br>(0.0005)***     | I(1)                 |
|                  | Trend    | -1.3894<br>(0.8455)                | -5.4666<br>(0.0005)***     | -1.3895<br>(0.8454)       | -5.4650<br>(0.0005)***     |                      |
| LEX              | Constant | 1.8562<br>(0.9996)                 | -3.7456<br>(0.0090)***     | 0.6259<br>(0.9885)        | -5.5069<br>(0.0001)***     | I(1)                 |
|                  | Trend    | -4.2223<br>(0.0126)**              | -4.0362<br>(0.0195)**      | -1.6420<br>(0.7558)       | -5.9240<br>(0.0001)***     |                      |
| LGDP             | Constant | 11.7070<br>(1.0000)                | -0.9385<br>(0.7638)        | 10.6692<br>(1.0000)       | -0.4172<br>(0.8953)        | I(1)                 |
|                  | Trend    | 5.4086<br>(1.0000)                 | -2.6394<br>(0.0002)***     | 4.8174<br>(1.0000)        | -2.5012<br>(0.0012)***     |                      |
| LEU              | Constant | 0.3113<br>(0.9757)                 | -2.7478<br>(0.0766)*       | 0.1861<br>(0.9679)        | -5.8045<br>(0.0001)***     | I(1)                 |
|                  | Trend    | -1.8256<br>(0.6713)                | -2.6126<br>(0.2774)        | -1.9450<br>(0.6105)       | -5.7763<br>(0.0002)***     |                      |
| LFD              | Constant | -1.9497<br>(0.3068)                | -7.7154<br>(0.0001)***     | -2.1787<br>(0.2171)       | -13.4098<br>(0.0001)***    | I(1)                 |
|                  | Trend    | -3.9104<br>(0.0218)**              | -7.5865<br>(0.0001)***     | -3.9104<br>(0.0218)**     | -14.2603<br>(0.0001)***    |                      |

Note: The ADF and PP test equations include both constant and trend terms. The Schwarz information criterion (SIC) is used to select the optimal lag order in the ADF test equation. The values in brackets are corresponding p values. \*, \*\*, \*\*\*Significance level at 1, 5, and 10% respectively.

The cointegration test results are reported in Table 5. The calculated F-statistics must be larger than the upper bound critical values. The F-statistics for the CO<sub>2</sub> emission model is 8.101 and for the ecological model, the F-statistics is 4.886.

Table-5. Results of ARDL bounds test.

| Dependent variable: Carbon emission model                             |                   |                   |                         |
|---|-------------------|-------------------|-------------------------|
| Model 1: (LCO <sub>2</sub> = LEX, LGDP, LGDP <sub>2</sub> , LEU, LFD) |                   |                   |                         |
|   | Lower bound I (0) | Upper bound I (1) | Significance levels (%) |
|   | 2.08              | 3                 | 10                      |
| <b>F= 8.101</b>   | 2.39              | 3.38              | 5                       |
| K=5   | 3.06              | 4.15              | 1                       |
| Dependent variable: Ecological footprint Model                        |                   |                   |                         |
| Model 2: (LEFP= LEX, LGDP, LGDP <sub>2</sub> , LEU, LFD)              |                   |                   |                         |
|   | 2.49              | 3.38              | 10                      |
| <b>F= 4.886</b>   | 2.81              | 3.76              | 5                       |
| K=5   | 3.5               | 4.63              | 1                       |

Note: k is the number of regressors. The model selection method is the Akaike info criteria (AIC).

Before the estimation, all models need to pass several diagnostic tests as given in Table 6. Brown et al. (1975) developed the cumulative sum of the residual square test residual square (CUSUMQ) to assess the model's median stability and variance. When two straight lines are represented as critical value lines, the CUSUM track decreases below 5%, indicating that the coefficients are stable. A comparable metric was used in the CUSUMQ test. The CUSUM and CUSUMQ tests' graphs. As a result, the coefficients can be determined to be stable. Tests of serial correlation and normality. The models pass the tests since their p-values are greater than 0.05, indicating that the null hypothesis that there is no serial correlation or normality issue is not rejected. Their residuals are serially uncorrelated and normally distributed, according to them. The Jarque-Bera test, the Lagrange Multiplier test for

residual serial correlation, and the CUSUM and CUSUM square tests are all shown in Figure 2 and Figure 3 separately in each model to show the stability test outcomes.

Table-6. Diagnostic check.

|                          | Model 1        | Model 2        |
|--------------------------|----------------|----------------|
| Normality (Jarque-Bera)  | 0.46<br>(0.79) | 0.34<br>(0.84) |
| Serial Correlation LM    | 1.53<br>(0.23) | 1.04<br>(0.59) |
| Heteroscedasticity (BPG) | 5.01<br>(0.75) | 3.7<br>(0.80)  |
| <b>Stability Test</b>    |                |                |
| CUSUM (5% sig.)          | Stable         | Stable         |
| CUSUM Sq (5% sig.)       | Stable         | Stable         |

Note: figures in parenthesis show p-values.

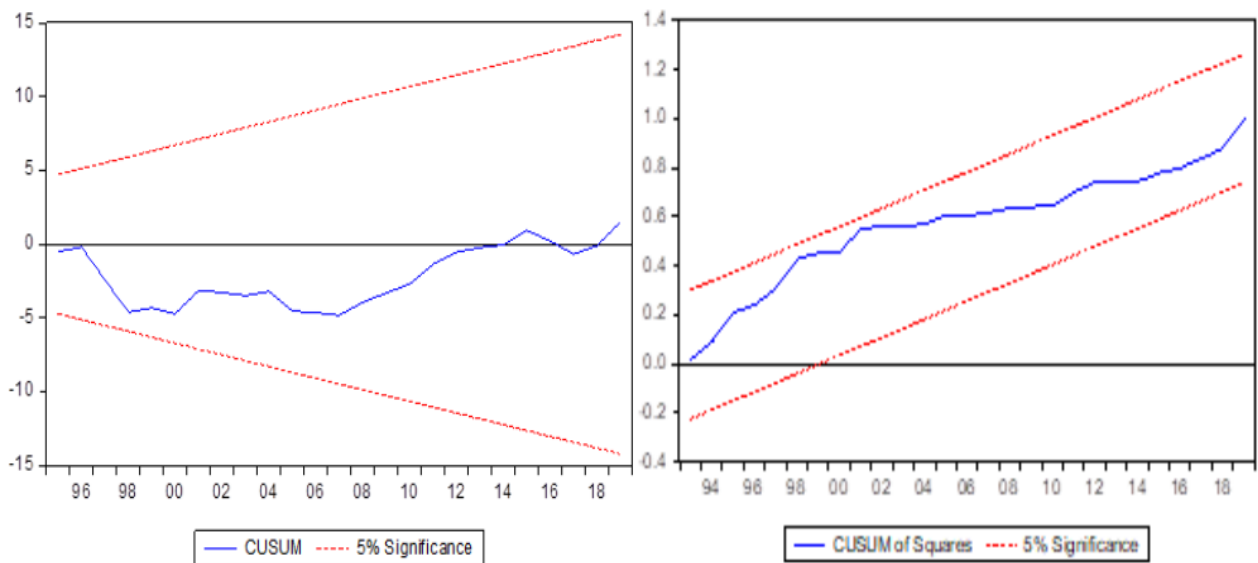


Figure-2. Stability test of CUSUM & CUSUM of Squares for CO2 emission model.

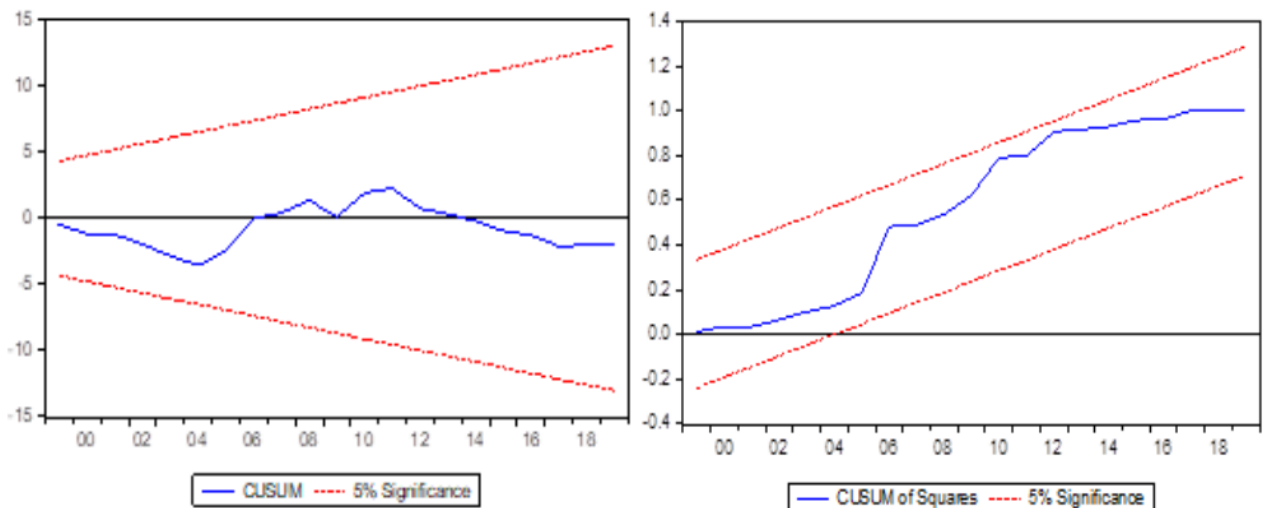


Figure-3. Stability test of CUSUM & CUSUM of Squares for Ecological footprint model.

#### 4.2. Results of FMOLS and DOLS Estimations

For confirming the long-run effects of the variables on CO2 and EFP as reported in Tables 7 and 8, the current study applied two robustness checks (DOLS and FMOLS). The application of dynamic OLS results which were different from those obtained by fully modified OLS. Table 7 shows the result for the dependent variable CO2 and



Table 8 shows the result for the dependent variable LEFP. Table 7 the export is turned out to be a major factor causing CO<sub>2</sub> emission means a 1% increase in export will cause 0.0409% in FMOLS and 0.0133% in DOLS estimations. The coefficients of LRGDP and LRGDP<sup>2</sup> for both FMOLS and DOLS results in it can be concluded that the EKC hypothesis is valid for model 1a, a similar result was found in the previous result. The energy consumption (LEU) was surprisingly found negative and not significant in both FMOLS and DOLS with values of .1646% and .3147%. The coefficients of financial development (LFD) are showing positive and significant at a 1% level with values of 0.4786% and 0.574% in FMOLS and DOLS respectively. This means with the improvement of FD in the country will reduce the environmental quality. This result is consistent with recent research by Al-mulali and Sheau-Ting (2014), which demonstrated that financial development decreases environmental quality in European nations. When a result, as more carbon emissions are produced, financial progress is linked to severe environmental effects.

**Table-7. Estimation results by FMOLS & DOLS (Dependent variable CO<sub>2</sub>).**

| Variables         | FMOLS                | DOLS                 |
|-------------------|----------------------|----------------------|
| LEX               | 0.0409**<br>(0.0138) | 0.0133**<br>(0.4437) |
| LGDP              | 4.0141<br>(0.0492)   | 7.8593<br>(0.0005)   |
| LGDP <sup>2</sup> | -0.2474<br>(0.0858)  | -0.5363<br>(0.0019)  |
| LEU               | -0.1648<br>(0.5409)  | -0.3147<br>0.1791    |
| LFD               | 0.4786**<br>(0.0019) | 0.5749**<br>(0.0001) |

Note: \*\*\*, \*\*, \* denote Significant at the 1%, 5% and 10% levels respectively, Figures in parentheses.

Table 8 shows the results for the ecological model. In terms of LRGDP and LRGDP<sup>2</sup> results for model 2, FMOLS is not significant whereas DOLS is significant to mean increasing economic growth will help to reduce the ecological footprint. Whereas, in DOLS model shows that economic growth will reduce the environmental quality of Bangladesh.

**Table-8. Estimation results by FMOLS & DOLS (Dependent variable EFP).**

| Variables         | FMOLS                  | DOLS                  |
|-------------------|------------------------|-----------------------|
| LEX               | -0.0745***<br>(0.0001) | 0.2018<br>(0.1142)    |
| LGDP              | 0.7857<br>(0.6930)     | 8.6794**<br>(0.0730)  |
| LGDP <sup>2</sup> | -0.0435<br>(0.7584)    | -7.7172**<br>(0.0743) |
| LEU               | 0.4419<br>(0.1101)     | 0.2588<br>(0.7438)    |
| LFD               | 0.0926<br>(0.5173)     | -5.9594**<br>(0.0786) |

Note: \*\*\*, \*\*, \* denote Significant at the 1%, 5% and 10% levels respectively, Figures in parentheses.

However, the DOLS result is consistent with previous studies such as Ozokcu and Ozdemir (2017); Pal and Mitra (2017) reported that there is no significant evidence of EKC. This implied that the continued economic growth in Bangladesh shall be expected to harmful for the environmental improvements. The energy consumption (LEU) shows a positive coefficient in FMOLS and DOLS respectively. As the export performance and economic performance increase it the consumption of energy will also rise. This is the situation in Bangladesh. In order to abate the devastating impact of coal consumption results in a higher ecological impact. Charfeddine and Mrabet (2017) have examined energy consumption degrades environmental quality by increasing ecological footprint by

applying FMOLS and DOLS. Interesting and conflicting results have been achieved with the financial development (LFD) of the capital market. Contrasting results were found for the increase in CO<sub>2</sub> emissions on the capital market. The FMOLS estimates in particular have a good impact but the DOLS estimates have shown a significant negative impact of FD on the degradation of the environment.

## 5. CONCLUSION & POLICY IMPLICATIONS

This paper has examined the long-run co-integration among financial development, export, energy consumption, GDP, and CO<sub>2</sub> emissions in the case of Bangladesh throughout 1983-2019 by employing FMOLS and DOLS methods. In addition, an inverted U-shaped link between per capita real income and per capita carbon emissions was discovered, confirming the validity of the EKC hypothesis in the Bangladesh economy over the longer term. This signifies that the per capita income is at high-level Carbon emissions rise in tandem with per capita real GDP at first. When it reaches a certain point, then a rise Carbon emissions are reduced when per capita real income rises. However, in the ecological footprint measure, there is no presence of EKC hypothesis. Since economic growth has a negative impact on the environment, it appears important to call for an increase in energy efficiency in manufacturing technologies if long-term economic development is desired. In this regard, the regulator should ensure that investment projects have minimal environmental effects by approving the most socially and ecologically responsible investments that employ cleaner industrial technology.

The financial development variables show a negative impact on the environment in both models whereas the energy consumption has a positive impact on the environment when used CO<sub>2</sub> emission measure whereas the result turns into opposite when the ecological footprint measure is taken. The findings of this study have policy implications for Bangladesh. To begin with, environmentally conscious essential industries can benefit from trade subsidies pollution-intensive industries can be taxed in the most efficient way possible. The financial sector may give a variety of lending facilities to the real sector that wishes to embrace cleaner and more environmentally friendly technology, therefore supporting these investments. Thus, it may be claimed that the Bangladeshi government should maintain sensitivity to both environmental and economic goals. As a result, Bangladesh's policymakers should think more about environmentally-friendly technology in order to emit less carbon and raise awareness about the issue. Nonetheless, policymakers in Bangladesh must carefully balance their connection and place a premium on emission reduction, since rapid expansion would obstruct long-term economic development while substantially increasing the cost of living. In the future, the expense of pollution management will be higher. Governments, in particular, could find it legitimate to mobilize greater financial resources for industrial upgrading, which might boost productivity and reduce energy use and, ultimately, economic growth as well as a reduction in carbon emissions in long run. As a result, the authorities may consider prioritizing the growth of the stock market, which performs better in terms of limiting carbon emissions than the expansion of financial institutions. Overall, by prohibiting the import of energy-inefficient devices, offering subsidies on low-carbon technology, and imposing or raising taxes on fossil fuel consumption, the government may improve the quality of life of households and meet the well-being goals. Future studies may use the findings and methods of this study to gain a better understanding of the finance–environmental nexus in other developing nations.

**Funding:** This study received no specific financial support.

**Competing Interests:** The author declares that there are no conflicts of interests regarding the publication of this paper.

## REFERENCES

- Al-Mulali, U., & Sheau-Ting, L. (2014). Econometric analysis of trade, exports, imports, energy consumption and CO<sub>2</sub> emission in six regions. *Renewable and Sustainable Energy Reviews*, 33, 484-498. Available at: <https://doi.org/10.1016/j.rser.2014.02.010>.

- Alam, M. J., Begum, I. A., Buysse, J., & Van Huylenbroeck, G. (2012). Energy consumption, carbon emissions and economic growth nexus in Bangladesh: Cointegration and dynamic causality analysis. *Energy Policy*, 45, 217-225. Available at: <https://doi.org/10.1016/j.enpol.2012.02.022>.
- Ali, H. S., Law, S. H., Lin, W. L., Yusop, Z., Chin, L., & Bare, U. A. A. (2019). Financial development and carbon dioxide emissions in Nigeria: Evidence from the ARDL bounds approach. *GeoJournal*, 84(3), 641-655.
- Alom, K., Uddin, A. N. M., & Islam, N. (2017). Energy consumption, CO<sub>2</sub> emissions, urbanization and financial development in Bangladesh: vector error correction model. *Journal of Global Economics, Management and Business Research*, 9(4), 178-189.
- Altıntaş, H., & Kassouri, Y. (2020). Is the environmental Kuznets Curve in Europe related to the per-capita ecological footprint or CO<sub>2</sub> emissions? *Ecological Indicators*, 113, 106187. Available at: <https://doi.org/10.1016/j.ecolind.2020.106187>.
- Aufderheide, P., & Rich, B. (1988). Environmental reform and the multilateral banks. *World Policy Journal*, 5(2), 301-321.
- Bekhet, H. A., Matar, A., & Yasmin, T. (2017). CO<sub>2</sub> emissions, energy consumption, economic growth, and financial development in GCC countries: Dynamic simultaneous equation models. *Renewable and Sustainable Energy Reviews*, 70, 117-132. Available at: <https://doi.org/10.1016/j.rser.2016.11.089>.
- Brown, R. L., Durbin, J., & Evans, J. M. (1975). Techniques for testing the constancy of regression relationships over time. *Journal of the Royal Statistical Society: Series B (Methodological)*, 37(2), 149-163.
- Charfeddine, L., & Mrabet, Z. (2017). The impact of economic development and social-political factors on ecological footprint: A panel data analysis for 15 MENA countries. *Renewable and Sustainable Energy Reviews*, 76, 138-154. Available at: <https://doi.org/10.1016/j.rser.2017.03.031>.
- Coban, S., & Topcu, M. (2013). The nexus between financial development and energy consumption in the EU: A dynamic panel data analysis. *Energy Economics*, 39, 81-88. Available at: <https://doi.org/10.1016/j.eneco.2013.04.001>.
- Costantini, V., & Monni, S. (2008). Environment, human development and economic growth. *Ecological Economics*, 64(4), 867-880.
- Destek, M. A., & Sarkodie, S. A. (2019). Investigation of environmental Kuznets curve for ecological footprint: The role of energy and financial development. *Science of the Total Environment*, 650, 2483-2489. Available at: <https://doi.org/10.1016/j.scitotenv.2018.10.017>.
- Farhani, S., & Ozturk, I. (2015). Causal relationship between CO<sub>2</sub> emissions, real GDP, energy consumption, financial development, trade openness, and urbanization in Tunisia. *Environmental Science and Pollution Research*, 22(20), 15663-15676.
- Ghosh, B. C., Alam, K. J., & Osmani, M. A. G. (2014). Economic growth, CO<sub>2</sub> emissions and energy consumption: The case of Bangladesh. *International Journal of Business and Economics Research*, 3(6), 220-227.
- Grossman, G. M., & Krueger, A. B. (1995). Economic growth and the environment. *The Quarterly Journal of Economics*, 110(2), 353-377.
- Inglesii-Lotz, R., & Dogan, E. (2018). The role of renewable versus non-renewable energy to the level of CO<sub>2</sub> emissions a panel analysis of sub-Saharan Africa's Big 10 electricity generators. *Renewable Energy*, 123, 36-43. Available at: <https://doi.org/10.1016/j.renene.2018.02.041>.
- Islam, F., Shahbaz, M., Ahmed, A. U., & Alam, M. M. (2013). Financial development and energy consumption nexus in Malaysia: A multivariate time series analysis. *Economic Modelling*, 30, 435-441. Available at: <https://doi.org/10.1016/j.econmod.2012.09.033>.
- Jalil, A., & Feridun, M. (2011). The impact of growth, energy and financial development on the environment in China: A cointegration analysis. *Energy Economics*, 33(2), 284-291.
- Katircioglu, S., & Katircioglu, S. (2018). Testing the role of fiscal policy in the environmental degradation: The case of Turkey. *Environmental Science and Pollution Research*, 25(6), 5616-5630.
- Majeed, M. T., & Mazhar, M. (2019). Financial development and ecological footprint: A global panel data analysis. *Pakistan Journal of Commerce and Social Sciences (PJCSS)*, 13(2), 487-514.

- Majeed, M. T., Samreen, I., Tauqir, A., & Mazhar, M. (2020). The asymmetric relationship between financial development and CO<sub>2</sub> emissions: The case of Pakistan. *SN Applied Sciences*, 2(5), 1-11.
- Maji, I. K., Habibullah, M. S., & Saari, M. Y. (2017). Financial development and sectoral CO<sub>2</sub> emissions in Malaysia. *Environmental Science and Pollution Research*, 24(8), 7160-7176.
- Mesagan, E. P., & Nwachukwu, M. I. (2018). Determinants of environmental quality in Nigeria: Assessing the role of financial development. *Econometric Research in Finance*, 3(1), 55-78. Available at: <https://doi.org/10.33119/erfin.2018.3.1.3>.
- Moghadam, H. E., & Dehbashi, V. (2018). The impact of financial development and trade on environmental quality in Iran. *Empirical Economics*, 54(4), 1777-1799.
- Narayan, P. K. (2005). The saving and investment nexus for China: Evidence from cointegration tests. *Applied Economics*, 37(17), 1979-1990.
- Nathaniel, S. P. (2020). Ecological footprint, energy use, trade, and urbanization linkage in Indonesia. *GeoJournal*, 1-14. Available at: <https://doi.org/10.1007/s10708-020-10175-7>.
- Ozokcu, S., & Ozdemir, Ö. (2017). Economic growth, energy, and environmental Kuznets curve. *Renewable and Sustainable Energy Reviews*, 72, 639-647. Available at: <https://doi.org/10.1016/j.rser.2017.01.059>.
- Pal, D., & Mitra, S. K. (2017). The environmental Kuznets curve for carbon dioxide in India and China: Growth and pollution at crossroad. *Journal of Policy Modeling*, 39(2), 371-385.
- Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level relationships. *Journal of Applied Econometrics*, 16(3), 289-326.
- Raza, S., & Shah, N. (2018). *Impact of financial development, economic growth and energy consumption on environmental degradation: Evidence from Pakistan*. Germany: University Library of Munich.
- Salahuddin, M., Alam, K., Ozturk, I., & Sohag, K. (2018). The effects of electricity consumption, economic growth, financial development and foreign direct investment on CO<sub>2</sub> emissions in Kuwait. *Renewable and Sustainable Energy Reviews*, 81, 2002-2010. Available at: <https://doi.org/10.1016/j.rser.2017.06.009>.
- Sarkodie, S. A., & Strezov, V. (2019). Effect of foreign direct investments, economic development and energy consumption on greenhouse gas emissions in developing countries. *Science of the Total Environment*, 646, 862-871. Available at: <https://doi.org/10.1016/j.scitotenv.2018.07.365>.
- Sehrawat, M., Giri, A. K., & Mohapatra, G. (2015). The impact of financial development, economic growth and energy consumption on environmental degradation. *Management of Environmental Quality: An International Journal*, 26(5), 666-682. Available at: <https://doi.org/10.1108/meq-05-2014-0063>.
- Shahbaz, M., Khan, S., & Tahir, M. I. (2013). The dynamic links between energy consumption, economic growth, financial development and trade in China: Fresh evidence from multivariate framework analysis. *Energy economics*, 40, 8-21. Available at: <https://doi.org/10.1016/j.eneco.2013.06.006>.
- Tamazian, A., & Rao, B. B. (2010). Do economic, financial and institutional developments matter for environmental degradation? Evidence from transitional economies. *Energy Economics*, 32(1), 137-145.
- Tang, C. F., & Tan, B. W. (2014). The linkages among energy consumption, economic growth, relative price, foreign direct investment, and financial development in Malaysia. *Quality & Quantity*, 48(2), 781-797.
- Uddin, M. M. M., & Wadud, M. A. (2014). Carbon emission and economic growth of SAARC countries: A vector autoregressive (VAR) analysis. *International Journal of Business and Management Review*, 2(2), 7-26.
- Uddin, G. A., Salahuddin, M., Alam, K., & Gow, J. (2017). Ecological footprint and real income: Panel data evidence from the 27 highest emitting countries. *Ecological Indicators*, 77, 166-175. Available at: <https://doi.org/10.1016/j.ecolind.2017.01.003>.
- WDI, T. (2020). World development indicators (DataBank). Retrieved from: <http://data.worldbank.org/>.
- Zakaria, M., & Bibi, S. (2019). Financial development and environment in South Asia: The role of institutional quality. *Environmental Science and Pollution Research*, 26(8), 7926-7937.

- Zhang, Y. J. (2011). The impact of financial development on carbon emissions: An empirical analysis in China. *Energy Policy*, 39(4), 2197-2203.
- Zhang, Y. J., Peng, Y. L., Ma, C. Q., & Shen, B. (2017). Can environmental innovation facilitate carbon emissions reduction? Evidence from China. *Energy Policy*, 100, 18-28. Available at: <https://doi.org/10.1016/j.enpol.2016.10.005>.
- Zhang, Y., & Zhang, S. (2018). The impacts of GDP, trade structure, exchange rate and FDI inflows on China's carbon emissions. *Energy Policy*, 120, 347-353. Available at: <https://doi.org/10.1016/j.enpol.2018.05.056>.

*Views and opinions expressed in this article are the views and opinions of the author(s), Energy Economics Letters shall not be responsible or answerable for any loss, damage or liability etc. caused in relation to/arising out of the use of the content.*