



Will be there New CO₂ Emitters in the Future? Evidence of Long-run Panel Co-integration for N-11 Countries

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ABSTRACT

This article tries to explore the long-run nexus between oil consumption, gross domestic product (GDP) and carbon dioxide (CO₂) emissions in the next eleven (N-11) countries over the period 1980-2013, by using the panel cointegration, the panel dynamic ordinary least squares (DOLS) and the panel fully modified ordinary least squares (FMOLS) approaches. The empirical findings indicate that there is a bidirectional long-run linkage between oil consumption - GDP per capita and oil consumption - CO₂ emissions. Moreover the inverted U-shaped linkage between the square of GDP per capita and CO₂ emissions, supports the existence of environmental kuznets curve hypothesis. With estimations through the panel DOLS and FMOLS, the long-run elasticity of oil consumption per capita to CO₂ emissions per capita is calculated about 0.96% and positive which is in contrast to the coefficient sign of its elasticity to GDP per capita (-0.48%). Moreover, the elasticity of GDP per capita and CO₂ emissions per capita to oil consumption per capita are -0.32% and 0.94%, respectively. These findings prove the negative contribution of non-renewable energy (oil) consumption per capita to GDP per capita in the N-11 group. Furthermore, due to the bidirectional long-run relationships between oil consumption and CO₂ emissions, these 11 countries should find the efficient energy policies which are in line with CO₂ mitigation and reaching a higher GDP per capita growth.

Keywords: Oil Consumption Per Capita, Gross Domestic Product Per Capita, Carbon Dioxide, Emissions Per Capita

JEL Classifications: E21, Q54, Q56

1. INTRODUCTION

Over the past few decades, based on the increasing threat of global warming and climate change, the carbon dioxide (CO₂) emissions-related issues have been attracting the scholars in the world. The findings of a high number of these studies such as Saboori and Sulaiman (2013), Saboori et al. (2014), Al-Mulali and Binti Che Sab (2012), Al-Mulali (2011), Ozturk and Acaravci (2010), Kasman and Duman (2015), Apergis and Payne (2010), Yildirim and Aslan (2012), Hannan (2015a) and Hannan (2015b) have reported a strong connection between economic growth and CO₂ emissions. In fact, many countries that experience high economic growth, account for a considerable contribution of global CO₂ emissions. But an interesting question that may be raised is what is the relationship between CO₂ emissions, economic growth and energy consumption in countries which have the potential to experience a high economic growth in the future. Since it is hard to predict which nation can reach to a high economic growth,

a proper group countries is next eleven (N-11) (Bangladesh, Egypt, Indonesia, Iran, South Korea, Mexico, Nigeria, Pakistan, the Philippines, Turkey and Vietnam) which was introduced by Goldman Sachs in 2005 and have a major potential to become next emerging countries.

It should be noted that the authors did not find any study incorporating all of the eleven nations in this group together (However, we have found some studies which considered one or some of these nations, i.e., (e.g. Alam et al. (2011) for India, Cheng (1997) for Mexico, Glasure and Lee (2002) for South Korea, Lotfalipour et al. (2010) for Iran, Shahbaz et al. (2013b) for Pakistan, Wolde-Rufael (2009) for Egypt and Yildirim and Aslan (2012) for Turkey)). Hence this research is different from the earlier literature and would be considered as the first attempt applying the panel approach for investigating the relationship between oil consumption per capita, CO₂ emissions per capita and gross domestic product (GDP) per capita in the N-11 group.

Furthermore, the findings of this study can shape an interesting picture of the problem of CO₂ emissions in the future. Policy makers will find out whether a group of countries like N-11 can become a new CO₂ emitters group in the future.

Generally, Sachs in 2005 introduced these 11 nations as the future economies like BRICS (Sachs, 2007). Figure 1 illustrates the trends of the average of this variable for the N-11 group in comparison with the world trend during 1980-2013. It can be seen from the figure that the average GDP per capita of all countries in the world is around 2514 and 10610 U.S dollars in 1980 and 2013. While, the average GDP per capita in the N-11 group is nearly 1074 U.S. dollars in 1980 and 6383 U.S. dollars in 2013. It can be noted that the growth rate of this variable in the N-11 group which is 4.9% is higher than the growth rate in the world (about 3.2%) during 1980-2013.

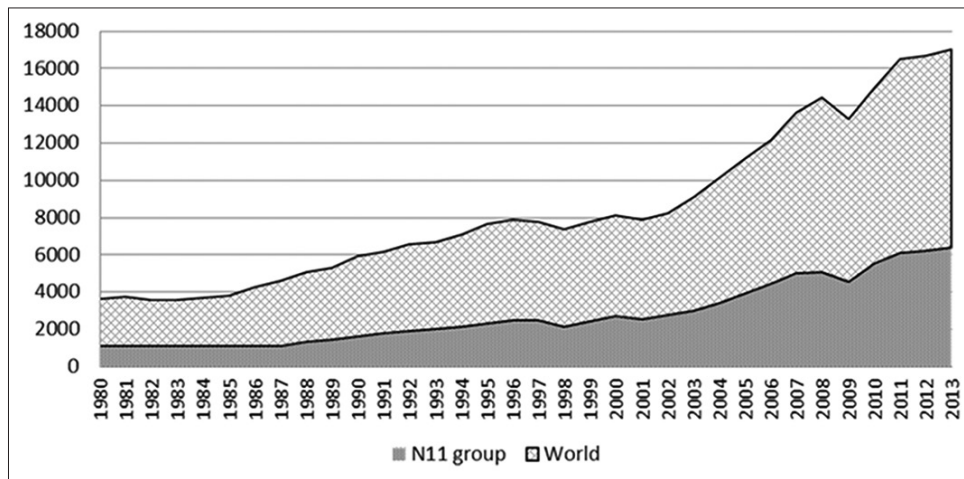
The first variable in the N-11 group that we consider here is GDP per capita. Figure 1 illustrates the trends of the average of this variable for the N-11 group in comparison with the world trend during 1980-2013. It can be seen from the figure that the average

GDP per capita of all countries in the world is around 2514 and 10610 U.S dollars in 1980 and 2013. While, the average GDP per capita in the N-11 group is nearly 1074 U.S. dollars in 1980 and 6383 U.S. dollars in 2013. It can be noted that the growth rate of this variable in the N-11 group which is 4.9% is higher than the growth rate in the world (about 3.2%) during 1980-2013.

The next variable is CO₂ emissions from consumption of petroleum in million metric tonnes from 1980 to 2013 (Figure 2). The N-11 countries were responsible for nearly 6.3% in 1980% and 12.1% in 2013 of global CO₂ emissions from consumption of petroleum. The high CO₂ emissions can be explained by a high density of population, dependence of national economies on manufacturing industries, a large share of fossil fuels on electricity generations of these nations.

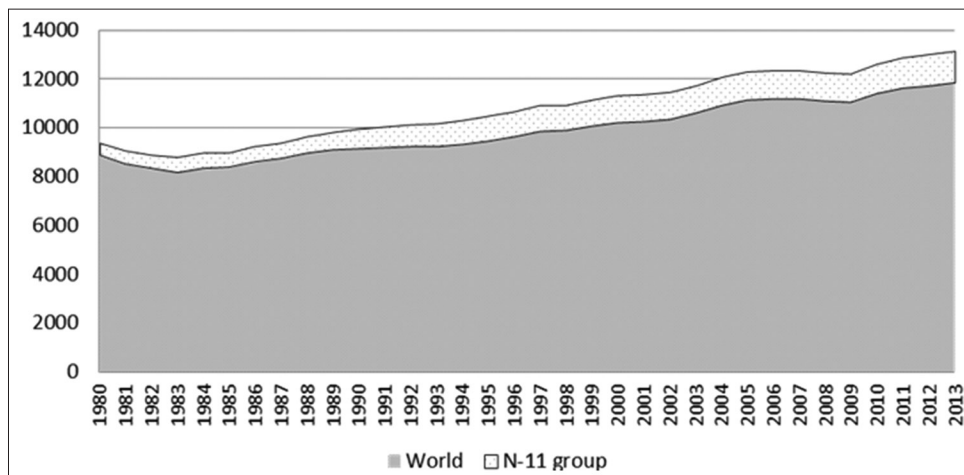
In addition, Figure 3 indicates oil consumption for these eleven countries and the entire world. It can be seen that the world oil consumption has increased over the period 1980-2013 from nearly 61233 thousand b/d to about 91243 thousand b/d. This increase has experienced a growth rate of 49%. In the case of N-11 group, the

Figure 1: Gross domestic product per capita in the next eleven group and World, %, 1980-2013

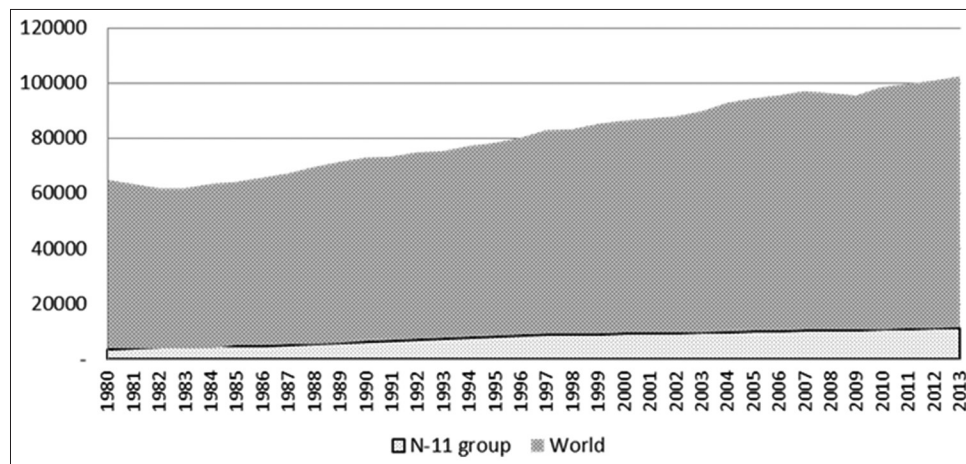


Source: Authors' compilation of the World Bank Database

Figure 2: Carbon dioxide emissions the next eleven group and World, in million metric tonnes, 1980-2013



Source: Authors' compilation of the international energy statistics

Figure 3: Oil consumption in the next eleven group and World, Thousands b/d, 1980-2013

Source: Authors' compilation of the BP statistical review of world energy

related oil consumption has boosted up through a 206% growth rate from 3631 thousand b/d in 1980 to nearly 11114 thousand b/d in 2013. It is clear that a higher oil consumption growth has been experienced by the N-11 group rather than the entire world during 1980-2013.

The rest of the article is outlined as follows: Section 2 considers data description and research methodology. The next explains results and the last section concludes the paper.

2. MATERIALS AND METHODS

2.1. Description of the Dataset

The six variables used in this study include per capita CO₂ emissions from the consumption of petroleum in metric tonnes, GDP per capita and square of GDP per capita in current US dollars, crude oil consumption per capita (as a proxy of non-renewable energy consumption) in barrels per day, trade openness and urbanization growth in percent as control variables (to overcome the omitted variable bias problem). All of the variables are used in the natural logarithmic form to reach a better result. Based on Wooldridge (2013), this form has many advantages such as satisfying the classical linear model assumptions than a form using the level of variables. The symbols, definitions and units of the research variables are represented in Table 1.

Countries in our sample which are known as N-11 contain South Korea, Indonesia, Iran, Mexico, the Philippines, Turkey, Bangladesh, Egypt, Nigeria, Pakistan, Vietnam. Data on the explained five variables for all these eleven nations are annually from 1980 to 2013. The main sources of the data are "World Bank," (2015), "International Energy Statistics," (2015) and "BP Statistical Review of World Energy 2015," (2015).

The summary descriptive statistics (mean, standard deviation, maximum and minimum) associated with GDP per capita, CO₂ emissions per capita, oil consumption per capita, trade openness and urbanization growth can be reported as follows: The mean GDP per capita ranges from 412.58 U.S. dollars in Bangladesh to about 11550.62 U.S. dollars in South Korea. Based on the data, as for the

Table 1: Variables definition

Variable	Definition
LGDPCC	Logarithm of GDP per capita in the selected countries
LGDPCC2	Logarithm of GDP per capita squared in the selected countries
LCO ₂ PC	Logarithm of CO ₂ emissions per capita in the selected countries
LOILCONPC	Logarithm of oil consumption per capita in the selected countries
LTRADE	Logarithm of trade openness in the selected countries
LURBAN	Logarithm of urbanization growth in the selected countries

Trade openness is the sum of exports and imports as a share of GDP, Source: Authors' compilation, GDP: Gross domestic product, CO₂: Carbon dioxide

crude oil consumption per capita, South Korea and Bangladesh have the highest and lowest mean of 3590.2 and 48.3 barrels per day from 1980 to 2013. With respect to CO₂ emissions per capita, which is measured in metric tonnes has the highest mean in South Korea, Iran and Mexico, respectively. In realizing the trade openness, Vietnam has the highest mean, whereas Bangladesh has the lowest trade openness degree. Finally, in terms of urbanization growth, Nigeria has the highest mean, followed by Bangladesh and Indonesia.

In sum, for all these 11 countries, the mean of GDP per capita, CO₂ emissions per capita, oil consumption per capita, trade openness and urbanization growth is nearly 2767.9 U.S. dollars, 1.27 metric tonnes, 976.6 barrels per day, 52.5% and 3.2%, respectively Table 2.

The following Table 3 illustrates the correlation matrix. It can be seen that the correlations between GDP per capita, square of GDP per capita and CO₂ emissions per capita are positive. Oil consumption per capita is positively related to CO₂ emissions per capita, GDP per capita and square of GDP per capita. The correlation shows a positive correlation between Trade openness and all the four variables CO₂ emissions, GDP, square of GDP and Oil consumption per capita. Finally, Urbanization growth is negatively correlated with all the other five variables.

2.2. Methodology

Following a large number of previous studies where the relationship between energy consumption, GDP and CO₂ emissions have been proved (e.g. Alam et al., 2011; Al-Iriani, 2006; Al-Mulali, 2011; Bhattacharaya and Bhattacharya, 2015; Bildirici and Bakirtas, 2014; Chang et al., 2009; Huang et al., 2008; Lee and Chang, 2008; Saidi and Hammami, 2015; Shahbaz et al., 2013; Soytas et al., 2007; Squalli, 2007; Zhang and Cheng, 2009), our research model under the environmental kuznets curve (EKC) is proposed as:

$$CO_2 \text{ emissions per capita} = f(\text{GDP per capita, Square of GDP per capita, Energy consumption per capita, Urbanization growth, trade openness}) \quad (1)$$

Or it can be considered as:

$$CO_2PC = \beta_0 GDPPC + \beta_1 GDPPC^2 + \beta_2 OILCONPC + \beta_3 URBAN + \beta_4 OPEN + \beta_5 \quad (2)$$

The above equation shows that CO₂ emissions per capita can be a function of per capita GDP and square of GDP per capita, square of GDP per capita, oil consumption per capita, urbanization growth and trade openness. To write the equation (2) in a form of econometric, particularly, a panel data, the following equation in the logarithmic form can be arranged as follows:

$$ICO_2PC_{i,t} = \beta_0 + \beta_1 IGDPPC_{i,t} + \beta_2 IGDPPC_{i,t}^2 + \beta_3 Iloilconpc_{i,t} + \beta_4 Iurban_{i,t} + \beta_5 Iopen_{i,t} + \varepsilon_{i,t} \quad (3)$$

Where i indicates 11 countries (i.e., South Korea, Indonesia, Iran, Mexico, the Philippines, Turkey, Bangladesh, Egypt, Nigeria, Pakistan, Vietnam), t is a time period that in this research covers 33 years from 1980 to 2013. Other symbols were defined in Table 1.

Before implication of the cointegration test, the panel unit root tests should be performed to find out whether all the series can be integrated of the same order. Actually, it is widely believed that the panel unit root tests are better than the unit root tests for the individual time series (Al-Mulali and Binti Che Sab, 2012). In this study, three

Table 2: Summary statistics for the variables, 1980-2013

Countries	GDPPC	CO ₂ PC	OILCONPC	Trade	Urban
Total N-11					
Mean	2767.94	1.27	976.63	52.54	3.26
Stdev	4193.77	1.31	1153.48	25.76	1.41
Max	25997.88	6.06	5205.25	165.09	10.82
Min	97.15	0.05	31.75	13.77	0.56

Source: Authors' compilation, CO₂: Carbon dioxide, N-11: Next eleven

Table 3: Correlation matrix for the variables

Variables	CO ₂ PC	GDPPC	GDPPC2	OILCONPC	TRADE	URBAN
CO ₂ PC	1.00	-	-	-	-	-
GDPPC	0.77	1.00	-	-	-	-
GDPPC2	0.61	0.92	1.00	-	-	-
OILCONPC	0.98	0.83	0.71	1.00	-	-
TRADE	0.07	0.21	0.25	0.11	1.00	-
URBAN	-0.49	-0.52	-0.41	-0.50	-0.31	1.00

Source: Authors' compilation, CO₂: Carbon dioxide

types of the panel unit root tests are computed which are LLC (Levin et al. 2002), augmented Dickey-Fuller (ADF) and PP- Fisher statistics (Maddala and Wu 1999) and (Choi, 2001). These three panel unit root tests consider a common LLC or individual (Fisher type test using ADF and PP test) unit root across the countries (cross sections). The hypotheses of these three panel unit root tests are as follows:

$$\begin{cases} H_0 : \text{Panel data has unit root} \\ H_1 : \text{Panel data has not unit root} \end{cases}$$

If the panel unit root tests prove that the variables are integrated of the same order, then we would perform the panel cointegration test to explore whether there is a long-run relationship between the variables of the model. In this research, to analyze the long-run relationship between variables, the Pedroni panel cointegration test (Pedroni, 1999; Pedroni, 2004) is implied for the residuals from the following equations. It can be noted that Pedroni heterogeneous cointegration test extends the Engle-Granger approach to panel data models (Liddle, 2012).

$$ICO_{2pcit} = \alpha_i + \delta_i t + \beta_{1i} IGDPPC_{it} + \beta_{2i} IGDPPC_{it}^2 + \beta_{3i} Iloilconpc_{it} + \beta_{4i} Iopen_i + \beta_{5i} Iurban_{it} + \epsilon_{it} \quad (4)$$

$$\hat{\epsilon}_{it} = \rho_i \hat{\epsilon}_{it-1} + \epsilon_{it} \quad (5)$$

Where i represents the number of countries in the panel, t indicates the number of observations over 1980-2013 in the panel and ε shows residuals. To estimate the equation (4) and (5), Pedroni has introduced seven various statistics which contain 4 within dimension statistics and 3 between dimension statistics (These statistics allow for heterogeneity of the variables in cross sections). The null hypothesis of all these statistics are “no cointegration” or “ρ_i=1 for all i.” In this study, the majority results of these 7 statistics are considered as the final decision.

Furthermore, besides the Pedroni test, the Kao (1999) panel cointegration test is applied which has a similar null hypothesis to the Pedroni test. Based on Kasman and Duman (2015), the main point of this cointegration test is consideration of intercepts in cross section and homogenous coefficients on the first stage regression.

After finding a long-run relationship through the Pedroni and Kao panel cointegration tests, the panel fully modified ordinary least squares (FMOLS) and the panel dynamic ordinary least squares (DOLS) approaches are applied to estimate the long-run cointegration vector.

Table 4: Panel unit root test results

Variable	Levin, Lin and Chu t	ADF-Fisher Chi-square	Phillips-Perron - Fisher Chi-square	H0 (majority)	Stationary
LGDPPC	-0.72 [0.23]	21.66 [0.48]	15.81 [0.82]	Accept	No
D (LGDPPC)	-12.40 [0.00]	158.93 [0.00]	354.65 [0.00]	Reject	Yes
LGDPPC2	-0.30 [0.38]	18.60 [0.66]	12.16 [0.95]	Accept	No
D (LGDPPC2)	-12.32 [0.00]	158.96 [0.00]	399.70 [0.00]	Reject	Yes
LCO ₂ PC	-0.76 [0.22]	38.64 [0.01]	3.15 [0.09]	Accept	No
D (LCO ₂ PC)	-11.05 [0.00]	202.37 [0.00]	256.74 [0.00]	Reject	Yes
LOILCONPC	-1.07 [0.14]	30.56 [0.10]	2.39 [0.43]	Accept	No
D (LOILCONPC)	-12.24 [0.00]	167.56 [0.00]	678.88 [0.00]	Reject	Yes
LTRADE	2.26 [0.98]	45.78 [0.00]	34.25 [0.06]	Accept	No
D (LTRADE)	-7.74 [0.00]	193.33 [0.00]	230.79 [0.00]	Reject	Yes
LURBAN	-0.05 [0.47]	38.54 [0.01]	31.00 [0.09]	Accept	No
D (LURBAN)	-8.39 [0.00]	136.95 [0.00]	348.10 [0.00]	Reject	Yes

Numbers in brackets indicate P values at the 5% level, Source: Authors' compilation, ADF: Augmented Dickey-Fuller

Table 5: Pedroni panel cointegration test results

Perdroni statistics	Statistic	P	Weighted statistic	P
Panel v-statistic	2.05*	0.02	-0.78	0.78
Panel rho-statistic	-1.73*	0.04	-0.34	0.36
Panel PP-statistic	-6.76*	0.00	-4.20*	0.00
Panel ADF-statistic	-7.38*	0.00	-4.81*	0.00
Group rho-statistic	0.69	0.75	-	-
Group PP-statistic	-6.23*	0.00	-	-
Group ADF-statistic	-5.98*	0.00	-	-

(*) Shows statistical significance at the 5% level, Source: Authors' compilation, ADF: Augmented Dickey-Fuller

Table 6: Kao panel cointegration test results

Dependent variables	T-satistic	P-value	Result
LCO ₂ PC	-9.10	0.00	Cointegrated

Source: Authors' compilation

3. RESULTS

3.1. Unit Root Tests

In order to determine the stationarity of all the underlying time series data, we carry out tree panel unit root tests for the variables at levels and first differences including individual intercept and trend. The results for LLC, the ADF - and the Phillips-Perron- Fisher type tests are reported in Table 4. It should be noted that the optimal lag length was selected automatically using the SIC (Shwarz Information Criteria) and the Newey-West method.

According to the reported p-values in the above Table 4, all the series are non-stationary at levels (means accepting the null hypothesis representing that the series contain a panel unit root) and stationary (rejecting the null hypothesis) at their first difference which stands for the integration at I(1).

3.2. Panel Cointegration Test

Since all the variables are cointegrated at I(1), the Pedroni and Kao panel cointegration tests can be applied to find out whether there is any long-run equilibrium relationship between the series. The achieved results are presented in the following Tables 5 and 6. From the results, by considering the Pedroni test and all the panel, group and weighted statistics, it indicates

that the p-values of eight statistics are <0.05 and hence, the majority of the all statistics tests can significantly reject the H₀ of no cointegration at the 5% significance level. Furthermore, the Kao panel cointegration test result depicts that all series in our model are cointegrated.

In sum, it can be concluded that there is an evidence of a long-run relationship between variables in the N-11 countries. These findings are in line with some previous researches such as Abid (2015) in Tunisia; Al-Mulali and Binti Che Sab (2012) in the Sub Saharan African Countries; Ang (2008) in Malaysia; Heidari et al. (2015) in Pakistan; Menyah and Wolde-Rufael (2010) in South Africa; Ozturk and Acaravci (2013) in Turkey; Saboori et al. (2014) in OECD; Salahuddin et al. (2015) in GCC countries; Vidyarthi (2013) in India; Wang (2013) in 138 countries and Yang and Zhao (2014) in India.

3.3. Panel Cointegration Estimations

Since the pedroni cointegration test depicts the long-run relationship between variables, the cointegrating coefficients of the series can be estimated by using the panel DOLS and FMOLS approaches. The following table summarizes the results of these estimations:

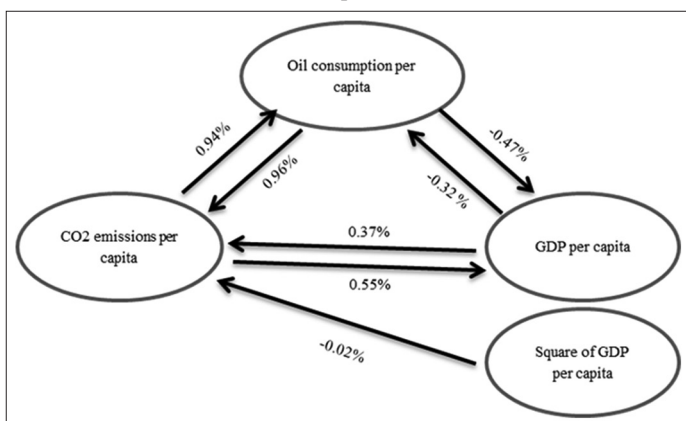
Based on the two applied estimation approaches and considering CO₂ emissions per capita as dependent variable, the coefficients of GDP per capita are statistically significant and positive, while the square of GDP per capita has negative coefficients of about -0.02. According to EKC hypothesis, there is a non-linear linkage between GDP per capita and CO₂ emissions per capita which can be interpreted as an inverted U-shaped (It proves the EKC hypothesis). It means that CO₂ emissions per capita begins to boost up until a specific level of income and then it goes down. This result is in line with some earlier studies such as Farhani and Shahbaz (2014), Kasman and Duman (2015) and Wang et al. (2011). Moreover, the results in Table 7 report a positive relationship between oil consumption per capita and CO₂ emissions per capita. In other words, a 1% increase in the N-11 countries' oil consumption per capita increases CO₂ emissions per capita by 0.96%. In the case of considering GDP per capita as dependent variable, the results support statistically positive significant long-run CO₂ emissions per capita - GDP per capita nexus. In contrast, oil consumption

Table 7: Panel FMOLS and DOLS estimations

Estimation approach	Dependent variables	Independent variables	Coefficient	t-statistic	P
FMOLS	LCO ₂ PC	LGDPPC	0.36	11.79	0.00
		LGDPPC2	-0.02	-11.86	0.00
		LOILCONPC	0.95	86.14	0.00
DOLS	LCO ₂ PC	LGDPPC	0.37	3.53	0.00
		LGDPPC2	-0.02	-3.69	0.00
		LOILCONPC	0.96	28.82	0.00
FMOLS	LGDPPC	LCO ₂ PC	0.56	6.79	0.00
		LOILCONPC	-0.51	-6.01	0.00
DOLS	LGDPPC	LCO ₂ PC	0.53	5.24	0.00
		LOILCONPC	-0.44	-4.17	0.00
FMOLS	LOILCONPC	LGDPPC	-0.43	-5.87	0.00
		LCO ₂ PC	0.97	38.18	0.00
DOLS	LOILCONPC	LGDPPC	-0.21	-2.30	0.02
		LCO ₂ PC	0.92	34.64	0.00

Source: Authors' compilation, DOLS: Dynamic ordinary least squares, FMOLS: Fully modified ordinary least squares

Figure 4: Long-run relationships between oil consumption per capita, carbon dioxide emissions per capita and gross domestic product per capita



Source: Authors' compilation

per capita has a negative coefficient of nearly -0.48 which means a 1% increase in the N-11 countries' oil consumption per capita leads to decreasing GDP per capita by 0.48%. In the last case, GDP per capita has a tendency to negatively affect oil consumption per capita, while there is a positive long-run linkage between CO₂ emissions per capita and oil consumption per capita (The diagram of long run linkage between the variables of our model is shown in Figure 4).

In sum, the long-run estimations prove the long-run positive effects of non-renewable energy on CO₂ emissions per capita and also support the long-run linkage between GDP per capita and CO₂ emissions per capita. These results are in line with some earlier studies (e.g. Begum et al., 2015; Ben Jebli and Ben Youssef, 2015; Bloch et al., 2012; Chandran Govindaraju and Tang, 2013; Farhani and Shahbaz, 2014; Long et al., 2015; Shafiei and Salim, 2014; Shahbaz et al., 2013; Tang and Tan, 2015 and Yildirim, 2014) who find a positive effect of non-renewable energy consumption on CO₂ emissions (deteriorate environment).

The diagram of long-run linkage between the variables of our model is as follows:

4. CONCLUDING REMARKS AND FUTURE DIRECTIONS

In this study, we have empirically tried to explore the dynamic long-run linkage between CO₂ emissions per capita, oil consumption per capita and GDP per capita for N-11 countries, i.e., South Korea, Indonesia, Iran, Mexico, the Philippines, Turkey, Bangladesh, Egypt, Nigeria, Pakistan, Vietnam using panel cointegration, the Fully Modified and the DOLS estimations for a wide range of a set of data from 1980 to 2013. In doing so, we implied various panel unit root tests to seek the variables' order of integration. The long-run relationships among variables CO₂ emissions per capita, oil consumption per capita and GDP per capita were analyzed by using the Pedroni panel cointegration test. The long-run coefficients were investigated by applying the FMOLS and DOLS. The empirical results indicated that in these eleven countries over the 33 years (1980-2013), there is long-run relationships between these three variables. Following the standard EKC hypothesis, the finding of this research proves an inverted U-shaped relationship between CO₂ emissions, GDP per capita and square of GDP per capita. However, according to our estimations, a 1% increase in GDP per capita leads to increase of CO₂ emissions per capita by nearly 0.37% in the 11 next countries.

The long-run elasticity of oil consumption per capita to GDP per capita in both the panel FMOLS and the DOLS estimations, is estimated to be around -0.48%. This amount of elasticity depicts the negative linkage between non-renewable energy consumption per capita and GDP per capita in the N-11 countries in the long-run. Furthermore, the long-run elasticity of oil consumption per capita to CO₂ emissions per capita is calculated to be about 0.96%, which is more than the amount of its elasticity to GDP per capita. It can be concluded that the contribution of oil consumption per capita to GDP per capita is in contrast to the contribution to CO₂ emissions per capita in all 11 countries under study.

The findings of this research indicate that oil consumption per capita affect GDP per capita and CO₂ emissions per capita. Further research should try to explore the best policies to reduce CO₂ emissions and increase GDP through some qualitative decision

making methods (such as ANP or DNP) or combined qualitative-quantitative methods (such as ANP-VAR model). Furthermore, the various energy sources - CO₂ emissions nexus can be further investigated in the N-11 group.

5. ACKNOWLEDGEMENT

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