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# Entrepreneurship, Sectoral Outputs and Environmental Improvement : International Evidence

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## **Abstract**

The relationship between entrepreneurship, output and environmental quality receives considerable attention from academics and policymakers, as society searches for solutions leading to environmental sustainability. Given this context, the current study contributes to this discussion by explaining how entrepreneurship and different sectoral outputs can help resolve the environmental problems of global socio-economic systems. So, we used data for 69 countries split across four homogeneous income-based panels: high-income, upper-middle-income, lower-middle-income, and low-income economies. Long-run elasticities suggest that (i) the rate of environmental damage due to the growth of sectoral outputs is much higher in the high-income sample; (ii) compared to output from other sectors, services makes the highest contribution to environmental degradation in high-income countries but its contribution in the other country samples is negative; indicating that a move to services economy would be beneficial for these countries; (iii) with the exception of the high-income sample, there is an inverted U-shaped relationship between output growth and environmental degradation across country samples and sectors; (iv) the contribution of entrepreneurial activity to environmental degradation is lower in high-income countries compared to other country samples; and (v) entrepreneurship activity in high-income countries initially degrades the environment but then improves environmental quality after a certain level, that is, an inverted U-shaped relationship between entrepreneurship and environmental pollution. The findings are sensitive to different income groups and sectoral analyzes. In particular, these empirical findings aid sound economic policymaking for improving environmental quality and sustainable economic development.

JEL Codes : Q5, O4, D2, C5.

Keywords: Entrepreneurship; Sectoral outputs; Environment; Economic stages of development.

## 1. Introduction

Since the mid 1980s, environmental concerns have been considered in the design of economic policy. Natural capital is considered to be an indispensable production input, and also a determinant of societal wellbeing (Costantini and Monni, 2008). The incorporation of environmental topics in economic growth theories and empirics is beginning to receive extensive consideration in the literature, and the question of whether output growth leads to more environmental degradation has become central in discussions among both economists and environmentalists.<sup>1</sup>

Moreover, concern about whether the social–ecological processes which allow human wellbeing to be sustained suggests that sustainable development should be a broad social goal. The role of entrepreneurship in achieving such goal is emerging as a subject of some debate. It is considered as the most important channel toward production of sustainable products and services, and implementation of new projects to address many environmental and social concerns. Several studies, such as Schumpeter (1934: 1942), Drucker (1985), and Matos and Hall (2007), among others, examine the link between resolution of global problems and entrepreneurship. For example, Cohen and Winn (2007) show that four types of market imperfection contribute to environmental pollution; they are considered as sources of significant entrepreneurial opportunity to establish the foundations for an emerging model of sustainable entrepreneurship by slowing the degradation and even gradually improving ecosystems. Similarly, York and Venkataraman (2010) propose entrepreneurship as a solution to, rather than a cause of, environmental degradation. These authors form a model that embraces the potential of entrepreneurship to supplement regulation, corporate social

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<sup>1</sup>Empirical debate over output growth and environmental quality began with the study by Grossman and Krueger (1991). The empirical association between them is described as the environmental Kuznets curve (EKC)<sup>1</sup>. The EKC describes a relationship where in the early stage of economic development environmental degradation increases with per capita income, and after a certain level of per capita income, environmental quality increases with a rise in per capita income (see Fig.1).

responsibility, and activism in resolving environmental problems. Shepherd and Pratzelt (2011) suggest that entrepreneurship can protect the ecosystem, improve environmental quality, reduce deforestation, and improve agricultural practices and freshwater supply. Since then, entrepreneurship could be a solution to numerous environmental and social problems (Wheeler et al., 2005; Senge et al., 2007; Hall et al., 2010)<sup>2</sup>. Starting from these considerations, we propose an EKC model which includes entrepreneurship as an aspect of sustainability.

This article makes two main contributions to the existing literature. First, we integrate entrepreneurship in the standard environmental Kuznets (EKC) model as an aspect of sustainability in order to examine the role of entrepreneurship activity on the environmental improvement. Specifically, we demonstrate that at early stages of economic development, entrepreneurial activity increases real incomes but damages the environment because at this stage, environmental quality is considered a luxury good. However, as countries achieve a certain level of economic development, the increased income from entrepreneurial activity contributes to the environmental improvement. Second, different sectoral outputs have been integrated in this model to identify the contribution of each sector on environmental quality, and to demonstrate that this contribution depends on the stages of economic development.

The rest of the article is organized as follows: section 2 provides a brief literature review; section 3 describes the empirical strategy; section 4 reports and discusses the empirical results; and section 5 concludes with some policy implications.

## **2. Theoretical framework and Hypotheses**

### *2.1. Entrepreneurship and Environment*

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<sup>2</sup> Several prestigious journals such as *Harvard Business Review*, *Journal of Business Venturing*, and *Entrepreneurship: Theory and Practice* published special issues covering this topic.

Currently, small businesses and entrepreneurship are economic fundamentals, and are responsible for breakthrough innovations which influence the growth of a free market economy and its general performance (Iyigun and Keskin, 2015). Originally, entrepreneurship was defined as establishing a business using individual capital and entrepreneurs and entrepreneurial activity have existed for a long time. However, Schumpeter introduced a new notion of entrepreneurship and of entrepreneurs as “innovators, who use a process of shattering the status quo of the existing products and services to set up new products, new services” (Sahin and Asunakutlu, 2014). In this perspective, entrepreneurship can be defined as the creation of new enterprising activities such as new ventures, strategic renewal, and innovation leading to better social and economic performance from companies (Habbershon et al., 2010).

Several researchers and practitioners view entrepreneurship as a channel for sustainable development, and expect the innovative power of entrepreneurship to produce the next industrial revolution and a more sustainable future. In this view, entrepreneurship is seen more and more as a significant tool for promoting the change to sustainable products and processes (Hall et al., 2010). Cohen and Winn (2007) provide evidence that four categories of market imperfections<sup>3</sup> contribute to environmental pollution, and see this as providing opportunities for significant entrepreneurial activity, and a model of sustainable entrepreneurship based on slowing environmental degradation and progressively enhancing the earth’s ecosystems. In addition, several environmentalists perceive the interconnection between business and the natural environment as a zero-sum game in which nature loses every time (Carson et al., 2003; Flannery, 2005). Similarly, Riti et al. (2015) investigate the causal relationship between entrepreneurship and the environment using a FMOLS approach

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<sup>3</sup> Inefficient firms, externalities, flawed pricing mechanisms, and information asymmetries.

for Nigeria in 2000-2012. They find that entrepreneurship has a negative impact on the environment which makes sustainable development unattainable.

However, other studies such as York and Venkataraman (2010) see entrepreneurship as a solution to rather than a cause of environmental degradation. Their model includes the potential for entrepreneurship to complement regulation, corporate social responsibility, and activism in relation to resolving environmental problems. Furthermore, according to Shepherd and Pratzelt (2011) entrepreneurial activity can preserve the ecosystem, counteract climate change, reduce environmental degradation and deforestation, improve agricultural practices and freshwater supply, and maintain biodiversity. In this context, the experience of developed countries shows that when countries reach a high level of economic development, the relationship between entrepreneurship and environmental damage becomes negative and takes an inverted U-shape form. So, increased entrepreneurial activity does not always increase environmental degradation. In addition, we can see that several works analyze the impact of entrepreneurial activity on environment but tend to overlook how this impact changes at different stages of development. For that reason, Acs et al. (1994) indicate that the level of entrepreneurship across country and time-specific contexts is explained mostly by the stage of economic development. Accordingly, we formulate the following hypothesis:

*Hypothesis 1.* The impact of entrepreneurship on environmental quality differs across stages of economic development.

## *2.2. Output and Environment*

Ecological modernization theory tries to clarify “how various institutions and social actors attempt to integrate environmental concerns into their everyday functioning, development, and relationships with others, including their relation with the natural world” (Mol et al., 2009). The theory builds upon a longstanding approach in environmental

economics which recognizes that income growth contributes to environmental damage, but argues that further income growth can lead to a reduction in such problems (Grossman and Krueger, 1995). The environment is perceived as a luxury good, subject to public demand through the workings of an advanced market. During earlier stages or periods of economic development, environmental harms increase, but as development and affluence reach a certain point, the value the public places on the natural environment increases.

As already mentioned, the empirical association between growth and environmental degradation is described as EKC. Several studies such as Grossman and Krueger (1993), Ozturk and Acaravci (2010), Lau et al. (2014), and Omri et al. (2015) test the validity of the EKC hypothesis but provide mixed results. Some find an inverted U-shaped relationship between economic growth and environmental degradation (e.g., Lindmark, 2002; Ang, 2007), others find a linear relationship (e.g. Azomahou et al., 2006) or no relationship (e.g. Ang, 2008; Chebbi, 2009) between these elements. This literature suffers from an omitted variables bias problem due to use of a bivariate model (Farhani et al., 2014). Other studies include other determinants of environmental degradation such as human development (Costantini and Monni, 2008 and Gurluk, 2009), financial development (Shahbaz et al., 2013, Omri et al., 2015), and trade liberalization (Tiba and Omri, 2015). However, these multivariate analyses also provide contrasting conclusions on the validity of the EKC hypothesis. While Hacıoğlu (2009) for Turkey, and Mensah (2014) for six African countries confirm the existence of an inverted U-shaped relationship between output growth and environmental pollution, others (Giovanis, 2013 for United Kingdom; Wang et al., 2013 for 150 nations) find no such evidence.

From the above, it is clear that most of the existing works focus on the impact of aggregate output on the environment but little attention is paid to the sectoral level of outputs at different stages of economic development. For the ecological modernization theory, the

impact of output on environmental degradation may increase for low- to middle-income countries but eventually declines for high-income countries. As high-income countries shift toward low carbon fuels, the output elasticity of emissions is likely to decline. The theory shows also that the output elasticity of emissions is affected by the level of technology efficiency. High levels of technology efficiency in high-income countries can help to reduce emissions. In this context, only few works such as Li and Lin (2015), Poumanyvong and Kaneko (2015) introduce industry sector in their analyses. These authors show that the impact of industrialization on environmental pollution is assumed to be positive but it is well known that at different stages of development, energy consumption takes different forms and involves different processes, causing the effects of industrialization on environmental degradation to vary. However, experience in developed countries shows that industrialization affects environmental degradation in different ways across different stages of development stages. In generally, in the middle phase of industrialization (pre-industrial and industrial economies), energy-intensive industries grow rapidly, and the effects of industrialization on environmental degradation are large and positive; however, in the later stages of industrialization (post-industrial economies), the effects become negative due to better energy efficiency and wide use of carbon-free energy types. We thus propose the following hypothesis:

*Hypothesis 2.* The impact of output on environmental quality differs across economic sectors and stages of economic development.

### **3. Empirical strategy**

#### *3.1. Data and Models*

The article estimates the relationships between entrepreneurship, GDP, and different sectoral outputs, and environmental quality by controlling for per capita energy use, per capita trade openness, per capita financial development, and human development. We measure



environmental pollution using CO<sub>2</sub> emissions. Real agriculture value added per capita ( $Y_A$ ), real industry value added per capita ( $Y_I$ ), and real services value added per capita ( $Y_S$ ) respectively measure sectoral outputs from the agriculture, industry and services sectors. The indicator of environmental degradation (E) is measured in metric tons per capita. The indicator of entrepreneurship activity (EP) is defined as the total number of new registered businesses as a percentage of the working-age population (Thai and Turkina, 2013; Dau and Cazurra, 2014). The indicator of foreign trade (T) is defined as export plus import divided by population, i.e. total per capita trade volume. The indicator of financial development (FD) is defined as private sector credit plus domestic credit provided by the banking sector divided by the population. Energy consumption (EC) in kg of oil equivalent per capita is used to measure energy consumption. The indicator of human development is measured by the modified human development index (Gürlük, 2009). Based on data availability, 69 countries were selected for the empirical estimation over the period 2001-2011<sup>4</sup>. Table 1 presents a detailed description of the variables used.

Using the World Bank classification<sup>5</sup>, we can split our sample of 69 countries into four homogeneous groups: high-income countries (22 countries), upper-middle-income countries (14 countries), lower-middle-income countries (23 countries), and low-income countries (10 countries)<sup>6</sup>. In our analyses, we used the following samples (table 2): sample 1 includes only high-income countries, sample 2 includes both high and upper-middle-income countries, sample 3 includes both upper-middle-income and lower-middle-income countries, and sample 4 includes both lower-middle-income and low-income countries.

**Table 1**  
Definition of the variables used in the analysis.

Variable	Definition	Data Source
Per capita CO <sub>2</sub> emissions (E)	CO <sub>2</sub> emissions are the release of carbon into the atmosphere. This indicator is used as a measure of	World Development Indicators

<sup>4</sup> Country selection and the period of study were based on the availability of data.

<sup>5</sup> <http://data.worldbank.org/about/country-classifications>.

<sup>6</sup> Lists of countries included in each panel are provided in Appendix.

	environmental degradation. Data is in metric tons per capita.	
Entrepreneurship (EP)	Measured as the total number of new registered businesses as a percentage of the working-age population.	Global Entrepreneurship Monitor (GEM)
GDP (Y)	Measured by per capita US\$ (2005).	World Development Indicators
Agricultural output (Y <sub>A</sub> )	Measured by per capita agricultural value added.	Calculated using data from World Development Indicators
Industrial output (Y <sub>I</sub> )	Measured by per capita industry value added.	Calculated using data from World Bank
Services output (Y <sub>S</sub> )	Measured by per capita services value added.	Calculated using data from World Development Indicators
Foreign trade (T)	Defined as export plus import divided by population i.e. total trade volume per capita.	Calculated using data from World Development Indicators
Financial development (FD)	Defined as private sector credit plus domestic credit provided by banking sector divided by population i.e. financial development per capita.	Calculated using data from World Bank
Energy consumption (EC)	Measured as kg of oil equivalent per capita.	World Bank
Human development (MHDI)	Measured using the modified human development index (MHDI) which measures the average achievements in a country in two basic dimensions of human development (education and life expectancy).	Calculated using data from World Bank

**Table 2**  
Samples presentation.

Countries	Sample 1	Sample 2	Sample 3	Sample 4
High-Income	HI	HI		
Upper-Middle-Income		UMI	UMI	
Lower-Middle-Income			LMI	LMI
Low-Income				LI
Total	22 countries	36 countries	37 countries	33 countries

In line with the literature, we formulate the following model:

$$E_{it} = \alpha_0 + \alpha_1 Y_{it} + \alpha_2 EP_{it} + \alpha_3 T_{it} + \alpha_4 FD_{it} + \alpha_5 HDI_{it} + \alpha_6 EC_{it} + \varepsilon_{it} \quad (1)$$

To test the validity of the EKC hypothesis, we specify and estimate the following multiple regression equations:

$$E_{it} = \alpha_0 + \alpha_1 Y_{it} + \alpha_2 Y_{it}^2 + \alpha_3 EP_{it} + \alpha_4 T_{it} + \alpha_5 FD_{it} + \alpha_6 HDI_{it} + \alpha_7 EC_{it} + \varepsilon_{it} \quad (2)$$

$$E_{it} = \alpha_0 + \alpha_1 EP_{it} + \alpha_2 EP_{it}^2 + \alpha_3 Y_{it} + \alpha_4 T_{it} + \alpha_5 FD_{it} + \alpha_6 HDI_{it} + \alpha_7 EC_{it} + \varepsilon_{it} \quad (3)$$

where  $i$ ,  $t$ , and  $\varepsilon$  are the country, the time period, and the error term respectively. In Eq.2 (EKC), the parameters  $\alpha_1, \dots, \alpha_7$  are the respective CO<sub>2</sub> emissions long-run elasticities with respect to income, squared income, entrepreneurship, trade, financial development, human

development and energy consumption. Based on the EKC hypothesis, the expected signs of  $\partial Y / \partial E > 0$  and  $\partial Y^2 / \partial E < 0$  lead to an inverted U-shaped relationship between emissions and income growth. In this study, the EKC hypothesis is extended further by replacing real GDP per capita by sectoral output in order to validate it across sectors. The logic behind this relationship is that at early stages of economic development, sectoral output induces pollution ( $\partial Y_A / \partial E > 0, \partial Y_I / \partial E > 0, \partial Y_S / \partial E > 0$ ); however, as income rises the incidence of further environmental damage decreases ( $\partial Y_A^2 / \partial E < 0, \partial Y_I^2 / \partial E < 0, \partial Y_S^2 / \partial E < 0$ ), due to higher environmental consciousness and use of modern technology which generates less pollution.

In Eq.3 (MEKC), we replace the square of GDP ( $Y^2$ ) by the square of entrepreneurship ( $EP^2$ ) in order to examine the quadratic relationship between entrepreneurship and environmental degradation. The logic underlying this relationship is that at early stages of economic development, entrepreneurial activity increases real incomes but damages the environment because at this stage, environmental quality is considered a luxury good. However, as countries achieve a certain level of economic development, the increased income from entrepreneurial activity encourages a higher societal demand for a clean environment, and induces efforts to reduce environmental damage by increasing the number of environmentally friendly projects and introducing clean production to improve environmental quality.

## *2.2. Estimation procedures*

In estimating the final versions of Equations (2) and (3) related respectively to the EKC and MEKC models, we use recently developed panel econometric techniques. They improve the statistical reliability of our tests by integrating cross-country heterogeneity and cross-country dependence. For heterogeneous countries, assuming cross-sectional

independence across panels could as Banerjee et al. (2004) and others suggest, distort the results.

To estimate our two models as a panel cointegration model, we consider a three-step empirical methodology. First, we analyze the cross-sectional dependence and check the stationarity of the series. Second, we perform a cointegration test to examine the long-run dynamics of cross-sectional dependence across countries. Third, we estimate the long-run relationships among the variables using fully modified ordinary least square (FMOLS) techniques.

### *2.3.1. Cross-sectional dependence and panel unit root tests*

The sample data were examined first using the Pesaran (2004) test for cross-sectional dependence (CD) to determine the presence of (CD) or cross sectional independence. This is an important step before applying panel unit root tests. The conventional unit root tests can provide weak findings due to low power if they are applied to series with CD. Therefore, we applied the cross-sectionally augmented panel unit root test (CIPS), one of the unit root tests from the second-generation developed by Pesaran (2007), which assumes that a series is CD. This unit root test is applied to investigate the order of integration in the series. This is a prerequisite for panel cointegration models. If the variables considered are  $I(1)$ , then it can be concluded that the variables tested are stationary at their first difference, suggesting that this group of variables may be cointegrated in the long-run. The next subsection provides a detailed discussion of the panel cointegration test.

### *2.2.2. Panel cointegration tests*

After confirming that the series is stationary by applying the Pesaran (2004) CD test and Pesaran (2007) CIPS unit root tests to the underlying models, we can perform panel cointegration analysis. The literature suggests a number of panel cointegration tests e.g. the Pedroni (1999, 2004) panel cointegration test, and the Kao (1999) panel cointegration test . In

our study we want also to check for a long-run equilibrium relationship among the variables, using the Pedroni (1999, 2004) panel cointegration test. Pedroni suggests seven different statistics to test for cointegration relationships in heterogeneous panels. These tests are corrected for bias introduced by potentially endogenous regressors, and are classified into within dimension and between dimensions statistics. The first sets are described as panel cointegration statistics, and the second are termed mean panel cointegration statistics.

### *2.2.3. Panel long-run estimation*

After all the variables are cointegrated, the next step is to estimate the associated long-run cointegration parameters. Fixed effects, random effects and general method of moment methods can lead to inconsistent and misleading coefficients when applied to cointegrated panel data. Among the existing panel data cointegration techniques, we use Pedroni (1999) Fully Modified Ordinary Least Squares (FMOLS) estimator which deals with possible heteroskedasticity and autocorrelation of the residuals, takes into account the presence of nuisance parameters, is asymptotically unbiased and, more importantly, deals with potential endogeneity of regressors. Tables 5 and 6 present the results of the long-run estimations using the FMOLS method.

## **4. Results and discussion**

The results in table 3 are for the Pesaran cross-sectional dependence test which is applied to all variables. The null of cross-sectional independence is rejected for each selected variable. Formal econometric modeling requires an understating of the integrating properties of the data. Thus, we apply Pesaran (2007) panel unit root test. Its results are reported in Table 3 and indicate that all series under consideration are non-stationary at their level form. However, at first difference level, the all series of variables are integrated. It implies that the selected series are integrated at  $I(1)$  in each panel.

Since at the first difference the variables are stationary for both panel EKC and panel MEKC, Pedroni's (1999, 2004) cointegration test is employed to examine the long-run equilibrium relationship between the variables. The results of the Pedroni (1999, 2004) panel cointegration tests are reported in table 4. Pedroni uses four within dimension (panel) test statistics, and three between dimension (group) statistics to check whether the selected panel data are cointegrated. Within dimension statistics contain the estimated values of the test statistics based on estimators pooling the autoregressive coefficient across different cross-sections for the unit root test on the estimated residuals. Between dimension statistics report the estimated values of the test statistics based on estimators that average individually estimated coefficients for each cross-section.

**Table 3**

Results of the panel unit root and cross-sectional dependence tests.

Variables	Pesaran CD test		CIPS test			
	CD-test	p-value	Level		$\Delta$	
			T-stat	p-value	T-stat	p-value
Sample 1 (HI): High-income countries						
LnE	10.124*	(0.000)	-1.197	(1.000)	-2.220*	(0.000)
LnY	12.085*	(0.000)	-1.205	(1.000)	-3.542*	(0.000)
LnY <sub>A</sub>	10.250*	(0.009)	-2.558	(0.998)	-2.119*	(0.003)
LnY <sub>I</sub>	11.529*	(0.000)	-2.184	(0.999)	-1.905*	(0.000)
LnY <sub>S</sub>	9.524*	(0.000)	1.893	(1.000)	-3.273*	(0.000)
Ln EP	10.552*	(0.000)	-2.052	(1.000)	-2.087**	(0.022)
LnT	14.921*	(0.000)	-1.013	(1.000)	-3.845*	(0.000)
LnFD	12.021*	(0.000)	1.464	(1.000)	-2.404*	(0.000)
MHDI	8.6103**	(0.017)	-2.231	(1.000)	-2.430**	(0.011)
LnEC	10.826*	(0.000)	2.118	(1.000)	-1.333*	(0.000)
Sample 2 (HI & UMI): High-income countries and Upper-middle-income						
LnE	5.122**	(0.023)	-0.997	(0.817)	-4.457*	(0.000)
LnY	10.129*	(0.000)	-2.655	(1.000)	-5.009*	(0.000)
LnY <sub>A</sub>	7.147*	(0.000)	-1.923	(0.998)	-1.957*	(0.000)
LnY <sub>I</sub>	4.509**	(0.014)	-0.804	(0.789)	-3.109*	(0.000)
LnY <sub>S</sub>	6.338*	(0.000)	-2.078	(1.000)	-7.114*	(0.000)
Ln EP	7.087*	(0.000)	-1.425	(0.998)	-5.727*	(0.000)
LnT	10.842*	(0.000)	-1.579	(1.000)	-2.910*	(0.000)
LnFD	15.530*	(0.000)	-1.930	(1.000)	-4.325*	(0.000)
MHDI	3.391**	(0.046)	-1.089	(0.999)	-3.185*	(0.000)
LnEC	10.826*	(0.008)	-3.108	(1.000)	-5.263*	(0.000)
Sample 3 (UMI & LMI): Upper-middle-income and Lower-middle-income						
LnE	11.392*	(0.000)	-2.533	(0.999)	-4.492*	(0.000)
LnY	9.711*	(0.000)	-2.203	(1.000)	-3.631*	(0.000)
LnY <sub>A</sub>	6.112*	(0.000)	-2.412	(1.000)	-3.221*	(0.000)

LnY <sub>I</sub>	8.298*	(0.000)	-1.883	(0.958)	-2.705*	(0.000)
LnY <sub>S</sub>	4.651*	(0.002)	-2.118	(1.000)	-2.150*	(0.000)
Ln EP	4.475*	(0.003)	-1.699	(1.000)	-2.625*	(0.000)
LnT	10.179*	(0.000)	-2.560	(1.000)	-3.002*	(0.000)
LnFD	12.615*	(0.000)	-1.704	(1.000)	-3.529*	(0.000)
MHDI	6.297*	(0.000)	-2.593	(1.000)	-2.057*	(0.000)
LnEC	2.574*	(0.000)	-1.856	(0.889)	-2.119*	(0.000)
Sample 4 (LMI & LI): Lower-middle-income and Low-income						
LnE	10.390*	(0.000)	-2.233	(0.765)	-3.397*	(0.000)
LnY	6.711*	(0.000)	-1.703	(1.000)	-2.930*	(0.000)
LnY <sub>A</sub>	6.252*	(0.000)	-1.712	(1.000)	-2.920*	(0.000)
LnY <sub>I</sub>	9.348*	(0.000)	-2.083	(0.958)	-3.005*	(0.000)
LnY <sub>S</sub>	5.601*	(0.000)	-2.108	(0.941)	-3.050*	(0.000)
Ln EP	7.475*	(0.000)	-1.719	(1.000)	-2.921*	(0.000)
LnT	7.179*	(0.000)	-1.720	(1.000)	-2.902*	(0.000)
LnFD	11.61*	(0.000)	-1.754	(1.000)	-3.322*	(0.000)
MHDI	7.181*	(0.000)	-2.192	(0.880)	3.365*	(0.000)
LnEC	6.895*	(0.000)	-2.179	(0.900)	-2.877*	(0.000)

Notes: All panel unit root tests were performed with restricted intercept and trend for all the variables. \* and \*\* are statistical significance at the 1% and 5% levels, respectively.

The results of the within dimension tests and the between dimension tests provide strong evidence that the null hypothesis of no cointegration in each panel should be rejected. Having confirmed the cointegration between these variables, in the next step we estimate the long-run coefficients. The test-statistics for all seven tests show that the null hypothesis of no cointegration can be rejected. Therefore, in our sample period, all the variables we consider have long-run associations. This leads to the conclusion that the variables considered are cointegrated for the four samples, and share a two long run equilibrium relationship with all the variables in eqs. (2) and (3). After confirming the cointegration among variables is confirmed, the long-run coefficients are estimated.