



INTERACTION BETWEEN ENERGY CONSUMPTION AND ECONOMIC GROWTH IN PAKISTAN: A MORE COMPREHENSIVE ANALYSIS USING ARDL APPROACH



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ABSTRACT

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Energy arguably plays a substantial part in the economic growth process. In this paper, we examine the intertemporal causal relationship between energy consumption and economic growth in Pakistan during the period of 1985–2017. Unlike the majority of the previous studies, we employ the newly developed autoregressive distributed lag (ARDL)-bounds testing approach by Pesaran *et al.* (2001) to examine this association. It is an attempt to explore the long run ties for energy consumption and energy intensity with economic growth, urbanization, trade openness, and financial development. Results postulate that the trade openness has a positive impact on energy consumption while urbanization and financial development have a negative influence. As far as sectoral analysis is concerned, agriculture and manufacturing share has a positive imprint on energy while the services sector has a negative effect. Overall, the study finds that energy consumption spurs economic growth in Pakistan. The findings have practical policy implications for decision makers in the area of macroeconomic planning.

Contribution/ Originality: This study is one of very few studies which have investigated the dynamics of the sector-wise impact of the economy on energy consumption. To make the paper more inclusive, effects of trade openness is checked with the help of different dimensions of trade openness; scale effect, technique effect, composition effect, and comparative advantage effect.

1. INTRODUCTION

Energy is recognized as fuel for industrial development and economic growth (Mirza *et al.*, 2019). The energy industry, along with its vital products, serves as an imperative factor in the production process of good and service and the main contributor to sustainable economic growth. Since the start of industrialization, the swift pace of economic growth is accompanied by hefty energy consumption. By increasing wages and boosting urbanization, industrialization creates a further increase to energy demand. For example, energy consumption augmented by more than 150% during last decade in China and documented as the world's biggest energy user in 2017. However, the use of energy, especially that of fossil fuel, has many hostile environmental impacts.

The energy consumption in terms of renewables is a noteworthy supplier to static greenhouse gas emissions. They are indispensable to keep the temperature of the earth warm. On the other side, the use of greenhouse gases caused by man-made actions, captivate more heat and lead to global warming. It causes climate change which documented as an extreme challenge for policymakers. The global climate change intimidates the wellbeing of

society, decreases economic development and alters the natural environment. So it becomes a key concern of policymaking of current century.

The potential for renewable energy technologies to fill the hole between supply and demand of energy in Pakistan is dynamic. Furthermore, decentralized renewable energy plans has the incentive to deliver electricity to rural and remote zones, in that way assisting to ease poverty and decreasing the prerequisite to collect and burn biomass fuel for energy scarcities. With a shortage over 5000 megawatt (MW) and continuously snowballing energy prices due to high fuel prices, the demand of sustainable, cheap, and clean energy is important for decreasing dependency on imported energy means. Like other developing economies, primary energy consumption has elevated 80 percent in the previous two decades in Pakistan.

There has been a plethora of research conducted by the scholars for exploring the connection between economic growth and energy consumption but there is no serious attempt has yet been made with the perspective of sectoral analysis in Pakistan. Growth has different sectors i.e. agricultural, manufacturing and services sector and each of these sectors contribute differently to energy use. Similarly, trade openness in Pakistan has different dimensions, for instance, scale effect, technique effect, and comparative advantage effect and they contribute differently to energy use as well. These are the key areas of interest in this study.

The prime objective of this study is to discover the dynamics of the sector-wise impact of the economy on energy consumption. The three sectors taken are the agricultural, manufacturing and services sector. Technology in the form of comparative advantage is also taken as a factor to impact energy use. Financial development, urbanization and environmental quality these all combine to affect energy use and so it can portray a larger image to conduct the study for the sake of Pakistan. To make the study more inclusive, the effect of trade openness is checked with the help of three different dimensions of trade openness i.e. scale effect, technique effect, composition effect, and comparative advantage effect. The aim is to check which sectors among the growth sectors would lead to more energy consumption and which strategy might be adopted in trade to diminish environmental pollution and hence, energy consumption can be handled in this way.

Remaining paper is structured as follow: Section two deals with the concerned literature review. Section three describes the research method. Section four shows results and empirical analysis. Chapter five presents conclusions.

2. LITERATURE REVIEW

In the last one or two decades, plenty of studies conducted by the researchers that found the causal nexus between energy consumption and economic growth mostly, the proxies used for these two are income and employment respectively. The findings have been ambiguous and conflicting (Asafu-Adjaye *et al.*, 2016). The first of the groundbreaking study done was by Kraft and Kraft (1978) which inferred that there is a causality from GNP to energy consumption in US. In the same way, Akarca and Long (1979) take monthly data of US find a unidirectional Granger Causality from energy consumption to the employment, having no feedback. These findings have been challenged by many researchers by then. Empirical evidences provided by Yu and Choi (1985); Erol and Yu (1987) find no causal nexus between energy consumption and GNP (a proxy for income).

Another strand of literature analyze this issue from another perspective as Kalimeris *et al.* (2014) review the energy to GDP causality using a meta-analysis approach which is quite different 158 studies have been taken for a period of 1978-2011. Multinomial logistic regression method results do not indicate the presence of direction of causality. It rejects the neutrality hypothesis. For the sake of Pakistan, Aqeel and Butt (2001) investigate the association of energy consumption to both the economic growth and employment in Pakistan. The methodology used is co-integration and Hsiao's Granger causality. Results indicate that total energy consumption as well as that of petroleum is caused by economic growth. The reason for these conflicting empirical findings lies in the choice of approaches and methodologies used for this study. In order to proceed with the advancement in time series data, in the last decade, bivariate causality tests have been used but these also have conflicting results.

The connection between economic growth and financial development is quite complex. Sadorsky (2011) studies the impact of financial development on energy consumption for nine European nations. Results confirm the statistically significant and positive relationship between energy consumption and financial development. Whereas, Çoban and Topcu (2013) study the effect of financial development on consumption of energy in the Europe. GMM based results do not contain any significant nexus but there is a strong proof of the effect of financial development on the energy consumption in the members that are old, irrespective of stock market or banking sector. For the new members, the same impact is dependent on the way the financial development is measured. Similarly, Furuoka (2015) takes the nexus between energy use and financial development for the period of 1980-2012. Heterogeneous panel causality test describes a long run equilibrium relationship between energy use and finance. The heterogeneous panel causality test further shows causality that is unidirectional and that runs from energy consumption to financial development.

Further extension in analysis made by Farhani and Solarin (2017) by examining the time series data of United States. The results suggest co-integration among them. Also, financial development lessens demand of energy in the long run but also stimulates in the short run. Nasreen *et al.* (2017) aim to study the nexus between financial stability, carbon dioxide emissions, energy consumption and economic growth for South Asian countries. Granger causality and bounds tests for co-integration result expresses that the environmental quality is improved by financial stability. As far as energy intensity is concerned, Voigt *et al.* (2014) studies the trends in energy intensity in 40 foremost economies. At the country level, the improvements in energy intensity are largely caused by the technological change. While at a global level, there is a shift of global economy to more energy intensive countries but still, aggregate energy efficiency is followed and improved by technological change. Likewise, Mirza *et al.* (2019) attempt to find out how to decompose the energy consumption and energy intensity into activity and efficiency changes. Fischer ideal index decomposition method suggest that energy intensity has been increasing to 53 percent during 1972-2011. Around 72 percent of this increase is due to the inefficient use of energy.

On the other hand, Tugcu and Topcu (2018) studies the nonlinear relationship between energy consumption and trade. Heterogeneity is involved to employ a panel framework and cross sectional dependence is checked. The sample used is of OECD countries from 1990-2015. Outcomes display that the effect of trade on energy consumption reveals an inverted U-shaped pattern and the nonlinear relationship is robust to estimation methods. Moreover, Fan *et al.* (2017) extends the analysis and empirically investigate the impact of urbanization on energy consumption taking into account the provincial differences. The results say that urbanization increases CO₂ emissions but it is not the case always. Urbanization strongly affects the regional CO₂ emissions in Northern China where there is a coal and heavy industry base.

In a nutshell, after keen evaluation of plethora of literature on economic growth and energy consumption, we divided the current study analysis into four different models with different explanatory variables taking into account. Conceptual discussion is provided in the next section.

3. THEORETICAL FRAMEWORK AND METHODOLOGY

3.1. Theoretical Framework

Energy demand and its consumption has crucial role for a country. It is not confined to country but has global impacts and consequences as well. This study investigates the relationship between economic growth and energy consumption for Pakistan. It also incorporates the consequences that environment of Pakistan faces. When energy burns, it releases dangerous chemicals which harms the entire atmosphere and specifies living and breathing under that environment. We estimates four different models, first of which examines effects of financial development, income, urbanization and trade openness on energy demand. Since liberalization of financial markets tend to promote growth, hence following Bekaert and Harvey (2000) we have the following model to estimate impacts of financial development and income on energy demand.

$$ED = f(FD, GDP)$$

Where ED stands for energy demand, FD stands for financial development and GDP indicates gross domestic product. Similarly, urbanization has been witnessed to increase the energy consumption ie, the more the urbanization, the higher is supposed to be the energy consumption. Hence forth, we would be taking urbanization as control variable and augment our model as below:

$$ED = f(FD, GDP, UR)$$

Where UR indicates urbanization (Sbia *et al.*, 2014) points out that another control variable which is supposed to have an impact on energy consumption is trade openness. Trade openness can have positive as well as negative impacts on energy consumption. Its impact can be negative if increasing trade flows result in bringing innovative technologies while positive when it increases the scale of production. Thus we are augmenting our model as follow in Equation 1:

$$ED = f(FD, GDP, GDP^2, UR, TR) \quad (1)$$

Where TR indicates trade openness. Similarly, we also add square of the GDP to account for Kuznets Curve for energy consumption. We further want to explore the sector wise impact of income on energy use, following Ling *et al.* (2015) we estimate another model by including the share of agriculture, manufacturing and services sector. For this purpose, we estimate the following Equation 2 model:

$$ED = f(FD, MS, AS, SS, UR, TR) \quad (2)$$

Where, FD is financial development, MS, AS and SS are manufacturing shares, agriculture shares and services shares respectively. To look further into determinants of energy demand, we take into account more of the research work. Literature further recommends that trade openness encourages mass awareness to demand for clean environment, energy-efficient technology transfer and government policy course toward ecological welcoming programs. The environmental significance of trade via energy consumption is varied by income effect, technique effect, and composition effect Jena and Grote (2008):

$$ED = f(GDP, GDP^2, K, TR, K.TR) \quad (3)$$

In Equation 3, GDP, GDP² are gross domestic product and its squared and they show scale effect and technique effect respectively. K is capital-labor ratio represents composite effect, TR is trade openness depicts trade effect while K.TR is comparative advantage effect:

$$EI_t = f(GDP, Krate, K/L) \quad (4)$$

Where in Equation 4, EI_t is energy intensity, it is ratio of energy use to GDP, while K denotes capital growth rate and K/L is ratio of capital and labor. We have taken energy intensity as dependent variable to check its determinants. However, we used GDP and capital growth rate and capital-labor ratio as explanatory variables. Variable of GDP is included to show the level of economic development. There is general belief that as economy develops energy efficiency also improves, so accordingly we expect GDP sign for model (4) to be negative. Following Metcalf (2008) capital-labor ratio is used as a proxy for level of technology. The intuition is that technology, energy and capital can be substituted. However, we expect capital-labor ratio to have a negative sign, since energy intensity may lower energy use because of improvements in the technologies. We also introduce the growth of capital stock in the model which is used to account for the speed by which old machines are replaced by new ones.

3.2. Econometric Methodology

Our main emphasis is to estimate dynamics of energy consumption for the country Pakistan, and since we have to deal with time series data, it has its own problems and properties. One of the most important properties of the time series is data stationarity, it must be checked otherwise simple ordinary least squares (OLS) will provide spurious coefficients. Fortunately, researchers have found the way to deal with this type of problem, if variable are non-stationary or there exists unit root in the series, they prefer to estimate co-integration techniques to estimate any relationships given variables and models.

Co-integration is broader concept under which comes different techniques, few of them are widely used based on their popularity, which are; single equation approaches including residual based Engle-Granger single equation technique, Engle and Granger (1987) and ARDL technique, Pesaran *et al.* (2001) and multiple equation approaches which includes Johansen-Juselius (JJ) technique, Johansen and Juselius (1990). Since we are interested in finding our dynamic relationship among variables, this study will apply ARDL approach to co-integration.

Speaking of ARDL technique, it is superior to other mentioned integrated techniques. Firstly, ARDL is flexible as compared to other approaches, that is, when order of integration is not same i-e some are I (1) and some are I (0), it can also be employed. In contrast, ARDL should not be used if any of the variables is integrated of order two, symbolically, I (2). Its flexibility also includes introduction of lags of both dependent and independent variables in the model, when lags of dependent variable are incorporated it is called “autoregressive”, while inclusion of lags of independent variables makes it “distributed lag”, thus, allows past values to impact dependent variable. Secondly, when ARDL takes sufficient number of lags, it uses general to specific framework to deal with and to capture data generating process. Moreover, estimates using ARDL are consistent if there is a short span of data. To attain optimal lag length, ARDL estimates the expression of (p+1) K number of regression. In the mentioned expression, k denotes number of variables while p denotes maximum lags.

Thirdly, ARDL is relatively robust when sample size is finite or small. According to Pesaran and Shin (1998) ARDL is superior in case of small sample on Johansen co-integration technique which requires sample to be large enough to produce valid and reliable results. In addition to that, the techniques of Johansen and Juselius (1990) and Engle and Granger (1987) do not yield reliable results in small sample case. Briefly speaking, in situation involving endogeneity, small size of sample and varying order of integration among variables, ARDL approach given by Pesaran *et al.* (2001) is used to find out short and long run connections among various variables.

3.2.1. Econometric Models of the Study

Based on availability, data on respective variables are taken from 1985 to 2016 for Pakistan. Since Pakistan is facing energy shortage against achieving its desired energy needs, so it will be interesting to study case of Pakistan. Complete variables description and data sources are presented in the appendix section of this study. Econometrical models of the study are described below:

Model 1

$$\ln EU_t = \beta_0 + \beta_1 \ln FD_t + \beta_2 \ln GDP_t + \beta_3 \ln GDP_t^2 + \beta_4 \ln UR_t + \beta_5 TR_t + \epsilon_t$$

Model 2

$$\ln EU_t = \gamma_0 + \gamma_1 \ln FD_t + \gamma_2 \ln AS_t + \gamma_3 \ln MS_t + \gamma_4 \ln SS_t + \gamma_5 \ln UR_t + \gamma_5 \ln TR_t + \mu_t$$

Model 3

$$\ln EU = \alpha_0 + \alpha_1 \ln GDP + \alpha_2 \ln GDP_t^2 + \alpha_3 \ln K_L_t + \alpha_4 \ln TR_t + \alpha_5 \ln K.TR_t + \mu_t$$

Model 4

$$\ln EI_t = \beta_0 + \beta_1 \ln GDP + \beta_2 \ln Krate + \beta_3 \ln K_L_t + \mu_t$$

Where \ln denotes natural logarithm, $\alpha_0, \gamma_0, \beta_0$ are intercepts, while $\beta's, \gamma's$ and $\alpha's$ are coefficients of respective variables. $\ln FD$ is natural log of financial development, $\ln GDP$ is natural log gross domestic product, $\ln UR$ is natural log of urbanization, $\ln TR$ is natural log of trade openness. $\ln AS$ is natural log of agriculture share, $\ln MS$ is natural log of manufacturing share $\ln SS$ is natural log of services sector, $\ln K_L$ is natural log of capital-labour ratio while $\ln K.TR$ is comparative advantage and $\ln Krate$ is growth rate of capital. The general form for ARDL model is in Equation 5:

$$\begin{aligned} \Delta E_t = & \\ & \alpha_0 + \alpha_1 E_{t-1} + \alpha_2 GDP_{t-1} + \alpha_3 FD_{t-1} + \alpha_4 TR_{t-1} + \alpha_5 UR_{t-1} + \alpha_6 \sum_{i=1}^p \Delta E_{t-i} + \alpha_7 \sum_{i=0}^p \Delta GDP_{t-i} + \\ & \alpha_8 \sum_{i=0}^p \Delta FD_{t-i} + \alpha_9 \sum_{i=0}^p TR_{t-i} + \alpha_{10} \sum_{i=0}^p UR_{t-i} + \varepsilon_t \end{aligned} \quad (5)$$

Where α_0 is intercept parameter while α_1 to α_{10} on right hand side are long run parameters indicating long run relationship. P shows number of lags, ε_t is error term which is white noise in the model. The terms along with delta sign and summation shows error correction estimates for short run. There are two steps in ARDL approach for calculating F-statistics for co-integration. First is the selection of lag length of the ARDL model, thus optimal number of lags must be selected before estimating ARDL model. There are different criterions for selection of optimal number of lags such as Akaike Information Criterion (AIC), Schwarz Information Criterion (SC), Log Likelihood Ratios (LR) and Log Likelihood test (LL). These all criterions have same null hypothesis that is, selected order of lag is optimal.

Once number of optimal lags are selected, we will go for second step of ARDL approach, which is to find out long run relationship of selected ARDL model. Prior to this, we will make use of Wald or F-test (Pesaran, 1997). Wald test is applied when we need to test for the significance of lagged levels of the variable. The variables which are incorporated in the unrestricted equilibrium error correction model. Speaking of statistical hypotheses, Wald test has null hypothesis of "no co-integration exists among the variables", while alternative hypotheses are:

$$H_0: \alpha_i = 0, \quad H_1: \alpha_i \neq 0$$

Pesaran *et al.* (2001) suggested critical values of F-statistics, which are used to make decisions, these values have $I(0)$ and $I(1)$ data generating process. Thumb rule for decision making for this test is following; If calculated value of F-statistics is greater than tabulated/critical values of $I(1)$, i-e upper bound, we reject null hypothesis meaning that there exists long run relationship. While if calculated value for F-statistics is less than that of critical/tabulated values of $I(0)$, i-e lower bound, we accept null hypothesis meaning that there exist log run relationship among variables. Moreover, result may be inconclusive if calculated values lies in between upper bound $I(1)$ and lower bound $I(0)$. This is the reason for ARDL as not valid technique for $I(2)$, because it has only two bounds. Once we have successfully applied Wald test, and found that there exists long run association among variables, we will move to our next step which is to estimate long run coefficients using ARDL model equation 5.

When we attain long run coefficients of the ARDL model for our variables, we may estimate short run coefficients as well. For short run analysis, it is necessary to retrieve error correction model from ARDL through linear transformation. The interesting fact regarding error correction model is that it integrates short run adjustments with long run, and luckily does not lose information. The main purpose of ECM is give information about speed of adjustment or say convergence of dependent variable after short run disturbances in independent variables towards long run equilibrium. Lower the value of coefficient of error correction term slower the speed of adjustment and vice versa. Another fact regarding error correction term is that it must be negative and significant

at high level of significance, which indicates that long run relationship is achievable among variables. ECM along with short run coefficient takes the form of Equation 6:

$$\Delta E_t = \alpha_0 + \alpha_1 ECM_{t-1} + \alpha_2 \sum_{i=1}^p \Delta E_{t-i} + \alpha_3 \sum_{i=0}^p \Delta GDP_{t-i} + \alpha_4 \sum_{i=0}^p \Delta FD_{t-i} + \alpha_5 \sum_{i=0}^p \Delta TR_{t-i} + \alpha_6 \sum_{i=0}^p \Delta UR_{t-i} + \varepsilon_t \tag{6}$$

Lastly but most importantly, diagnostic tests have vital importance since they diagnose problem regarding model specification and data used. Therefore, we have applied different diagnostic tests such as test for serial correlation, functional form, heteroskedascity and normality of residuals. These diagnostic tests include Ramsey RESET test, which tells whether functional form of model we have estimated is correct. Breusch Godfrey serial correlation LM test, which is very useful and widely used for checking serial correlation. For normality of residuals we have used Jarque-Berra test. In last, presence of heteroskedascity is checked via applying ARCH test. To check whether our model is structurally stable, Pesaran (1997) recommend use of CUSUM and CUSUMSQ tests proposed by Brown *et al.* (1975) which are widely used to check stability of model. Rule of thumb here is that, if these plots lie within the critical bounds at 5% level of significance, we cannot reject null hypothesis rather we accept it, and conclude that our model is stable. Null hypothesis is “all the coefficients in given regression are stable”.

4 RESULTS AND EMPIRICAL ANALYSIS

First part of this section presents graphical representations of dependent variable i-e energy consumption against all other explanatory variables to discover patterns and/or trends of variables. Figure 1 exhibit relationship between energy use and agriculture sector, trend is positively sloped indicating positive relationship. Figure 2 shows relationship between energy use and labor force, likewise, there is positive pattern shown by graph. Figure 3 depicts financial development against energy use, shows negative trend between these two. Figure 4 shows relationship between energy use and GDP which is also positively sloped, similarly, Figure 5 and Figure 6 depict energy use against manufacturing sector and capital respectively. Both tend to show positive pattern. Figure 7 shows positive relationship between energy use with services sector while Figure 8 shows negative trend between energy use and trade. Figure 9 exhibits positive pattern for urbanization against energy use.

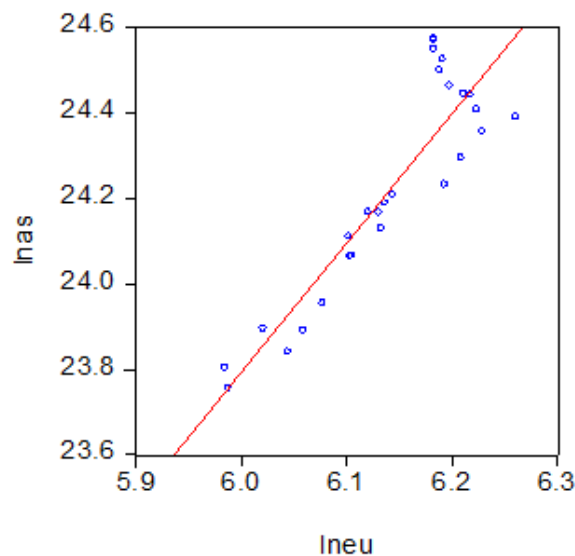


Figure-1. Energy use vs agri. sector.

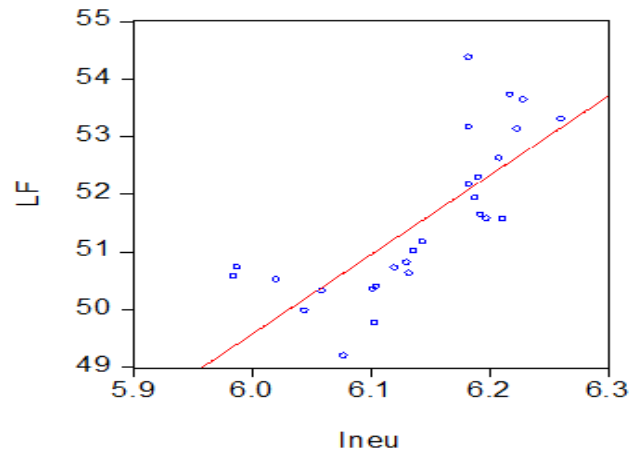


Figure-2. Energy use vs labor force.

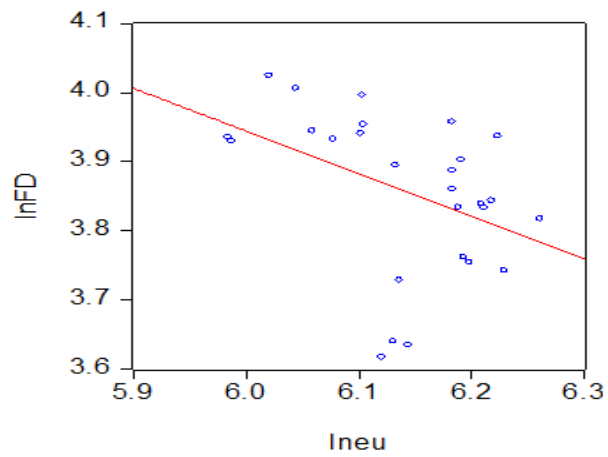


Figure-3. Energy use vs fin. dev.

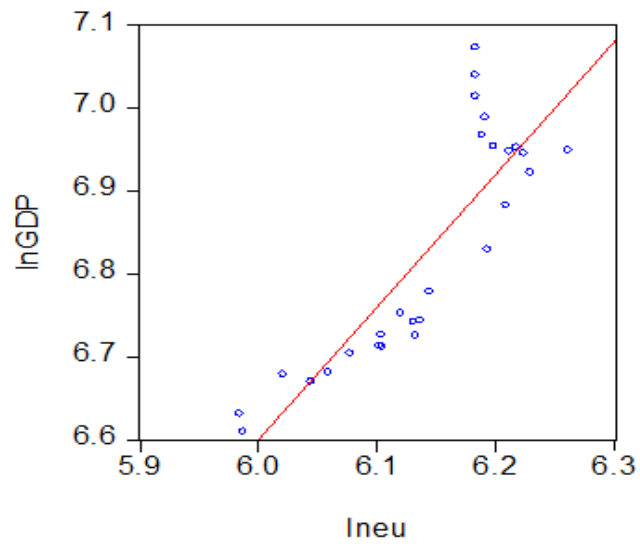


Figure-4. Energy use vs GDP.

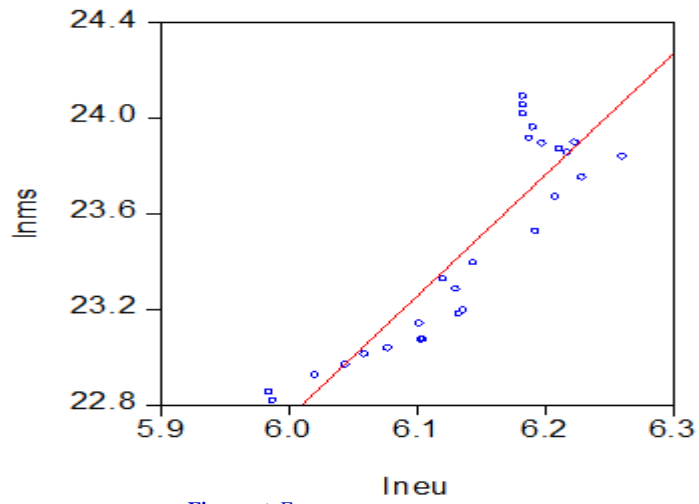


Figure-5. Energy use vs manu. sector.

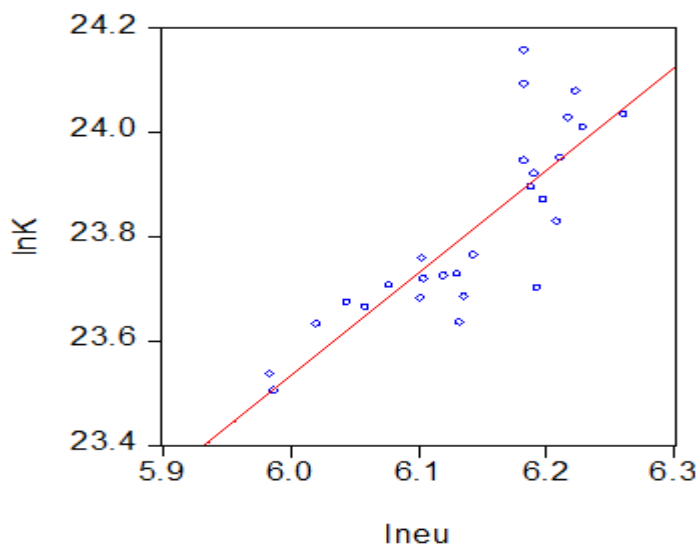


Figure-6. Energy use vs capital.

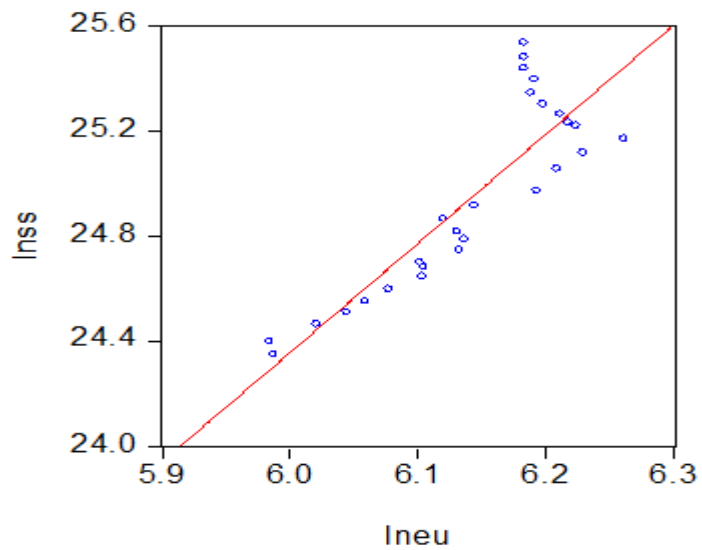


Figure-7. Energy use vs serv. sector.

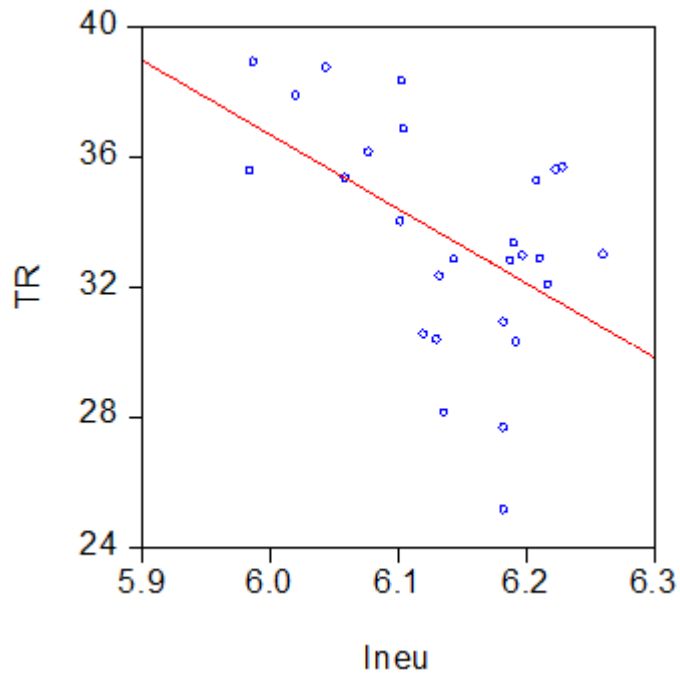


Figure-8. Energy use vs trade.

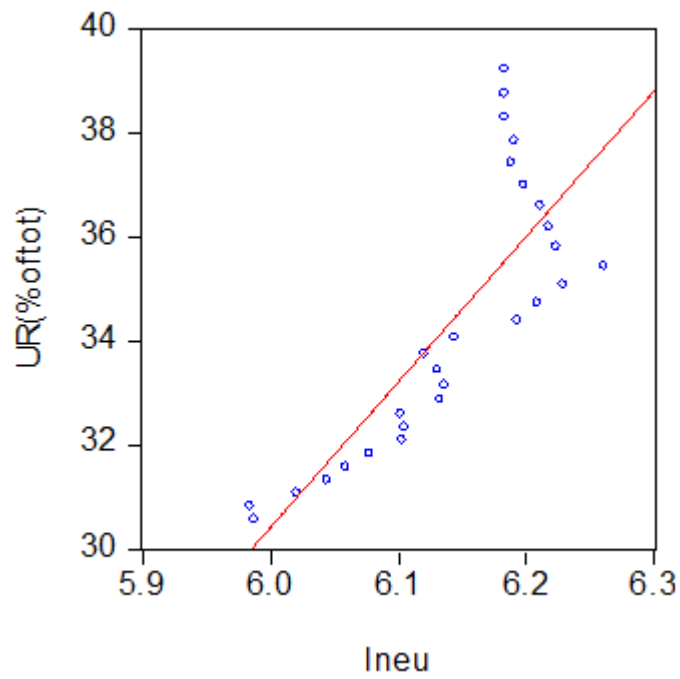


Figure-9. Energy use vs urbanization.

Table-1. Descriptive statistics.

Indicator	LNEU	LNAS	LF	KRATE	K_TR	K_L	LNFD	LNGDP	LNGDpsqr	LNMS	LNSS	TR	LNUR
Mean	6.141543	24.2216	51.52483	2.8444	15.48654	0.462355	3.857081	6.826521	13.65304	23.46551	24.9466	33.46264	34.38519
Median	6.144151	24.20805	51.16713	3.199175	15.2538	0.464314	3.886971	6.778149	13.5563	23.39228	24.91534	32.99043	34.065
Maximum	6.26104	24.57409	54.37153	19.90113	18.35536	0.481903	4.023913	7.072251	14.1445	24.08885	25.53582	38.90949	39.224
Minimum	5.984615	23.75421	49.19157	7.705547	11.16847	0.444266	3.616726	6.609082	13.21816	22.81563	24.34812	25.13914	30.576
Std. dev.	0.075437	0.253745	1.36099	6.762113	1.76638	0.009345	0.113975	0.140222	0.280445	0.429325	0.360725	3.460577	2.613814
Skewness	-0.58545	-0.26714	0.435549	0.551135	-0.30055	-0.10125	-0.66737	0.161627	0.161627	0.004416	0.012432	-0.39825	0.285807
Kurtosis	2.42394	1.901495	2.195486	3.227976	2.874349	2.489324	2.539357	1.588741	1.588741	1.439812	1.738119	2.766747	1.894614
JarqueBera	1.91572	1.678688	1.581811	1.425344	0.424238	0.339518	2.24296	2.358162	2.358162	2.738547	1.792082	0.774928	1.7422
Probability	0.383713	0.431994	0.453434	0.490332	0.808868	0.843868	0.325797	0.307561	0.307561	0.254292	0.408183	0.678776	0.418491

Table-2. Correlation matrix.

Variable	lneu	Ln FD	TR	ln GDP	Ln GD PSQR	LNMS	LNAS	LNSS	LF	lnK	K/L	K.TR	LNUR	KRATE
lneu	1													
lnFD	-0.40844	1												
TR	-0.49789	0.480873	1											
lnGDP	0.86015	-0.17153	-0.57394	1										
lnGDPsqr	0.86015	-0.17153	-0.57394	1	1									
lnms	0.888682	-0.25581	-0.57754	0.992266	0.992266	1								
lnas	0.898947	-0.31033	-0.65019	0.96952	0.96952	0.976291	1							
lnss	0.869575	-0.26594	-0.63719	0.988246	0.988246	0.991693	0.989263	1						
LF	0.762291	-0.17197	-0.50205	0.853262	0.853262	0.848789	0.790053	0.811432	1					
lnK	0.835353	-0.09292	-0.47441	0.936037	0.936037	0.918793	0.887777	0.903808	0.861293	1				
K/L	-0.6833	0.199851	0.47947	-0.76277	-0.76277	-0.76541	-0.70116	-0.72233	-0.97984	-0.74329	1			
K.TR	-0.57559	0.482114	0.987211	-0.65248	-0.65248	-0.65825	-0.71288	-0.70365	-0.62728	-0.55708	0.612068	1		
lnur	0.807579	-0.22955	-0.66117	0.979729	0.979729	0.977007	0.974208	0.992916	0.790883	0.879284	-0.70488	-0.72106	1	
Krate	-0.00744	0.059108	0.067882	0.110239	0.110239	0.067789	0.023597	0.044996	0.288841	0.2102	-0.29051	0.011803	0.047165	1

Table 1 deals with descriptive statistics while Table 2 provides correlation analysis among variables. We have applied ADF unit root test on all variables to find out whether our variables are stationary and in case if they are not stationary, on what difference they will become stationary, in other words, known the order of integration. The results are presented in the following Table 3.

Table-3. ADF unit root test.

Variable	At level			At first difference			Order
	Cal-value	Critical-value	P-value	Cal-value	Critical-value	P-value	
lnEU	-2.38152	-2.98104	0.1563	-3.73451	-2.98623	0.0098***	I(1)
lnFD	-1.63501	-2.98104	0.4511	-4.16246	-2.98623	0.0036***	I(1)
lnGDP	0.914933	-3.01236	0.9936	-3.04071	-2.98623	0.0447***	I(1)
lnGDPSQR	0.914933	-3.01236	0.9936	-3.04071	-2.98623	0.0447***	I(1)
lnAS	-1.54034	-2.98104	0.4978	-5.6318	-2.98623	0.0001***	I(1)
lnMS	-0.66283	-2.98623	0.8386	-2.97619	-2.98623	0.0510**	I(1)
lnSS	-0.08327	-2.98623	0.9411	-3.1085	-2.98623	0.0388***	I(1)
lnLF	-0.51608	-2.98104	0.8727	-4.58869	-2.98623	0.0013***	I(1)
lnTR	-1.46633	-2.98104	0.5343	-6.08777	-2.98623	0.0000***	I(1)
lnUR	-0.5968	-1.95568	0.4482	-2.60835	-1.95568	0.0115***	I(1)
lnK_L	-0.99888	-2.98104	0.7383	-5.1374	-2.98623	0.0003***	I(1)
lnK.TR	-1.2174	-2.98104	0.6511	-6.13611	-2.98623	0.0000***	I(1)
Krate	-3.58643	-2.98104	0.0133***	-	-	-	I(0)

Table-4. Lag order selection criteria.

Model 1						
Lag	LogL	LR	FPE	AIC	SC	HQ
0	302.7398	NA	1.97e-18	-23.73918	-23.44665	-23.65805
1	540.7985	342.8045	2.07e-25	-39.90388	-37.85616	-39.33593
2	640.1691	95.39587*	2.33e-27*	-44.97353*	-41.17064*	-43.91877*
Model 2						
Lag	LogL	LR	FPE	AIC	SC	HQ
0	156.4659	NA	1.51e-14	-11.95727	-11.61599	-11.86262
1	351.3361	265.0234	1.52e-19	-23.62689	-20.89661	-22.86962
2	482.0801	104.5952*	6.89e-22*	-30.16641*	-25.04713*	-28.74654*
Model 3						
Lag	LogL	LR	FPE	AIC	SC	HQ
0	255.9239	NA	8.36e-17	-19.99391	-19.70138	-19.91277
1	376.8753	174.1701	1.02e-19	-26.79002	-24.74231	-26.22208
2	446.4181	66.76113*	1.25e-20*	-29.47345*	-25.67056*	-28.41869*
Model 4						
Lag	LogL	LR	FPE	AIC	SC	HQ
0	119.8341	NA	3.14e-10	-10.53037	-10.33200	-10.48364
1	193.3696	113.6458	1.73e-12	-15.76087	-14.76902	-15.52722
2	218.2210	29.36984*	9.21e-13	-16.56555	-14.78020	-16.14497
3	236.6076	15.04360	1.22e-12	-16.78251	-14.20368	-16.17502
4	282.0807	20.66955	3.17e-13*	-19.46188*	-16.08956*	-18.66746*

* indicates optimal lags selected by specified criterion.

Table 4 deals with lag order selection criteria. Since we are following the results of ADF test, it concludes that all variables are stationary at first difference, I(1), for which ARDL technique to Cointegration can be applied. Since we have selected optimal lag criteria based on AIC, which is two optimal lags for models 1, 2 and 3 while four optimal lag for model 4. Bound Test is used in order to analyze the long run relationships and examine whether Cointegration exist or not. So Bound-test is applied on four models and results are presented in following table, which shows calculated f-values along with lower and upper bounds critical values. Table shows that for model (1) calculated F-value is 13.19193, lower bound 2.62 and upper bound 3.79. When all variables are order of integration I(0), then decision should be made on lower bound, whereas, if all variable of order I(1), we should decide on upper

bound I(1). Since our all variables are stationary at 1st difference or say are of order of integration I(1), we compare our calculated f-value with upper bound critical value. However, it can be concluded that there exists long run relationship among variables in our estimated model (1) i.e. calculated F-value is greater than upper bound. Table 5 provides Bound test results at 5% significance level.

Table-5. Bound test at 5% significance level.

	Calculated F-value	Critical lower bound I(0)	Critical upper bound I(1)	Result
Model 1	13.19193	2.62	3.79	Cointegration
Model 2	9.110683	2.45	3.61	Cointegration
Model 3	4.158414	2.62	3.79	Cointegration
Model 4	4.260630	3.23	4.35	Cointegration

4.1. Results of Long-Run Estimates (ARDL Model)

This sub-section reports results for long run estimates of ARDL model. Following table shows explanatory variables along with their respective coefficients, t-statistics and probability.

Table-6. Long run coefficients.

Model 1 dependent variable (lnEU)			
Variable	Coefficient	t-statistic	Prob.
LNFD	-0.09559	-3.43978	0.0088
LNGDP	28.34371	9.719116	0.0000
SQRLNGDP	-1.99597	-9.27861	0.0000
LNTR	0.027455	0.944657	0.3725
LNUR	-0.59673	-5.17139	0.0009
C	-91.6268	-9.39514	0.0000
Adjusted R ²	0.994019		
D. Watson statistics	2.957943		
Model 2 dependent variable (lnEU)			
Variable	Coefficient	t-statistic	Prob.
LNFD	0.072666	1.874132	0.1100
LNAS	0.296844	3.230479	0.0179
LNMS	0.227504	4.867763	0.0028
LNSS	-0.37203	-4.59649	0.0037
LnUR	-0.17439	-4.84166	0.0029
LnTR	0.000905	0.535597	0.6115
C	3.112022	3.613225	0.0112
Adjusted R ²	0.989947		
D. Watson statistics	3.077529		
Model 3 dependent variable (lnEU)			
Variable	Coefficient	t-statistic	Prob.
LNGDP	55.02006	13.66525	0.0000
LNGDPSQR	-4.00228	-13.5212	0.0000
K_L	-9.9747	-4.20679	0.0023
LnTR	-0.11315	-3.65109	0.0053
K_TR	0.249181	3.687336	0.0050
C	-178.339	-13.7252	0.0000
Adjusted R ²	0.988735		
D. Watson statistics	2.49295		
Model 4 dependent variable (lnEI)			
Variable	Coefficient	t-statistic	Prob.
LNGDP	-0.115540	-6.858480	0.0010
K_L	-0.825229	-2.968843	0.0312
KRATE	-0.001281	-3.530680	0.0167
C	2.071748	9.873514	0.0002
Adjusted R ²	0.976843		
D. Watson statistics	2.333951		

Energy consumption (*lnEU*) is used as dependent variable for models (1,2 and 3). Reporting results for model (1) describes that coefficient of financial development has negative sign while GDP is positive. Squared term of GDP also significantly negative along with Urbanization at 1% level of significance. Coefficient of trade openness has positive sign but it is insignificant variable indicating that trade openness does not significantly affect energy consumption. Value of adjusted R² for model (1) indicates model is pretty appropriate and fit as it explains 0.994019 variation in the model that is model predicts responses for new observations. Table 6 indicates the long run coefficients of model 1.

For model (2), variables such as agriculture share, manufacturing share, services share and urbanization are significant at 1% level of significant while financial development and trade openness are insignificant. Value of adjusted R² is appropriate suggesting that model explains variation and responses to new observation well. Findings for model (3) show that, all variables used in the models are significant at 1% level of significance. Similarly, for model (4) all variables are significant at 1% level of significance and all variables have negative signs.

4.2. Results of Error Correction Model (ECM)

We have extracted short run coefficients using error correction model which are reported in the following table. Error correction term (ECM) has vital importance in case of short run, since it shows speed of adjustment or say convergence, to put it in simpler words, it tells how long it will take for variable to converge.

Table-7. Short run coefficients (ECM).

Model (1) dependent variable = $\Delta\ln eu$			
Regressors	Coefficients	t-values	Probability
$\Delta\ln eu(-1)$	0.206676	1.498820	0.1723
$\Delta(\text{LNGDP})$	10.436003	1.351704	0.2134
$\Delta(\text{SqrlnGDP})$	-0.700256	-1.238103	0.2508
$\Delta(\text{lnFD})$	-0.055113	-1.362606	0.2101
$\Delta(\text{LNTR})$	-0.085653	-2.959916	0.0181
$\Delta(\text{LNUR})$	63.038532	2.998724	0.0171
CointEq(-1)	-0.692655	-6.706688	0.0002
Model (2) dependent variable = $\Delta\ln EU$			
$\Delta(\ln EU(-1))$	-0.171864	-0.768660	0.4713
$\Delta(\text{LNFD})$	0.016844	0.451528	0.6675
$\Delta(\text{LNAS})$	0.134042	2.345304	0.0574
$\Delta(\text{LNMS})$	0.210007	2.772848	0.0323
$\Delta(\text{LNSS})$	0.450135	1.917656	0.1036
$\Delta(\text{URGR})$	0.384149	1.810363	0.1202
$\Delta(\text{TR})$	-0.003095	-2.642852	0.0384
CointEq(-1)	-0.171864	-5.241138	0.0019
Model (3) dependent variable = $\Delta\ln EU$			
$\Delta(\text{LNEU}(-1))$	-0.150361	-1.018429	0.3351
$\Delta(\text{LNGDP})$	22.770081	1.932933	0.0853
$\Delta(\text{LNGDPSQR})$	-1.651934	-1.892680	0.0909
$\Delta(\text{K}_L)$	-9.451071	-3.786609	0.0043
$\Delta(\text{TR})$	-0.124388	-3.879681	0.0037
$\Delta(\text{K}_L \text{TR})$	0.261387	3.767464	0.0044
CointEq(-1)	-1.048985	-4.511555	0.0015
Model (4) dependent variable = $\Delta\ln EI$			
$\Delta(\text{LNEI}(-1))$	0.221720	0.551194	0.6052
$\Delta(\text{LNGDP})$	-0.033037	-0.929384	0.3953
$\Delta(\text{K}_L)$	-0.168731	-1.073350	0.3322
$\Delta(\text{KRATE})$	-0.000260	-1.840263	0.1251
CointEq(-1)	-0.602722	-2.458190	0.0574

For model (1), ECM has value -0.692655 at 1% level of significance in short run. It has implication that any shock will be corrected if it occurs in energy consumption by taking 69 percent speed in course of one year. Similarly, for model (2) value of ECM is -0.171864 at 1% level of significance. As well, model (3) has ECM value of -1.048985 at 1% level of significance indicating any shock will be adjusted in energy consumption by speed of 105% in course of one year. For model (4), ECM has value -0.602722 at 1% level of significance in short run. It shows that any shock will be adjusted if it occurs in energy intensity by taking speed of 60 percent in course of one year. Table 7 describes short run analysis of the model 1.

4.3. Encompassing Analysis

This section reports the results of encompassing analysis which are done to find out sensitivity and robustness of variables and to mitigate specification bias problem as shown in following tables.

Table-8. Model 1 dependent variable (EU).

Variables	Equation 1	Equation 2	Equation 3	Equation 4	Base eq.
LNFD	-0.43355*** (-3.87384)	-0.71209*** (-3.86366)	-0.419452*** (-2.387171)	7.190318 -0.025545	-0.09559*** (-3.43978)
LNGDP		1.342625*** -3.976537	16.289056 -1.251343	272.358376 -0.030342	28.34371*** -9.719116
SQRLNGDP			-1.114634 (-1.160006)	-20.330684 (-0.03016)	-1.99597*** (-9.27861)
LNTR				-0.090987 (-0.02679)	0.027455 -0.944657
LNUR					-0.59673*** (-5.17139)

Above Table 8 reports coefficient of model 1 for variables of financial development (LNFD), gross domestic product (LNGDP), squared term of GDP (SQRLNGDP), trade openness (LNTR) and urbanization (LNUR), whereas, energy use is used as dependent variable. Coefficient of financial development is negative and significant at 1% level of significance through all equations except Equation 4 where it is positive and insignificant. GDP is positive and significant for Equation 2 and base eq. while for Equation 3 & Equation 4 it is insignificant. Variable of squared GDP is negative throughout all equations, and significant at 1% of level of significance in base equation. Trade openness is insignificant throughout all equations, while urbanization is negative and significant at 1% level of significance.

Similarly, the following Table 9 reports coefficients of model 2 for variables of financial development (LNFD), agriculture sector (LNAS), manufacturing sector (LNMS), services sector (LNSS), trade openness (LNTR), urbanization (LNUR). Coefficient of financial development is negative from Equation 1 to Equation 4. It is positive for base and Equation 4. It is significant only for Equation 1 and Equation 2 at 1 percent level of significance. Coefficient of agricultural sector is positive throughout the equations. It is negative and significant for Equation 2, 5 and base eq. at 1 percent level of significance. Coefficient of manufacturing sector is positive throughout the equations except for Equation 3. It is significant only in the base eq at 1 percent level of significance. Likewise, the coefficient of services sector is negative in all the equations and it is significant only in the base equation. Coefficients of urbanization and trade openness are negative and insignificant except for urbanization in base equation, which is significant at 1% level of significance.

Table-9. Model 2 dependent variable (EU).

Variable	Equation 1	Equation 2	Equation 3	Equation 4	Equation 5	Base eq.
LNFD	-0.43355***	-0.33248***	-0.00483	-0.100025	0.132704	0.072666
	(-3.87384)	(-4.59406)	(-0.004342)	(-0.695157)	-1.550735	-1.874132
LNAS		0.124465***	0.815514	0.108518	0.299078***	0.296844***
		(-3.386085)	(-0.501844)	(-0.10617)	(-3.40624)	(-3.230479)
LNMS			-0.794101	0.267889	0.142194	0.227504***
			(-0.462831)	-0.965477	-1.125352	-4.867763
LNSS				-0.235836	-0.043118	-0.37203***
				(-0.251744)	(-0.081365)	(-4.59649)
LnUR					-0.288477	-0.17439***
					(-0.496461)	(-4.84166)
LnTR						0.000905
						-0.535597

Table 10 reports findings for model 3, coefficients of GDP and squared GDP are significant at 1% level of significance and are positive and negative respectively, for equations Equation 4 and baseline Eq. while coefficient of capital and labor ratio has negative sign and significant only in baseline eq. Similarly, trade openness has negative sign and significant at 1% level of significance, however, comparative advantage variable is found to be positive and significant showing that 1% increase in comparative advantage leads to 0.249% increase in energy use.

Table-10. Model 3 dependent (EU).

Variable	Equation 1	Equation 2	Equation 3	Equation 4	Baseline eq.
LNGDP	-0.33663	-11.804512	25.42732	45.76374***	55.02006***
	(-0.82148)	(-0.162415)	-1.442257	-8.219856	-13.66525
LNGDPSQR		0.818114	-1.84153	-3.31596***	-4.00228***
		(-0.157902)	(-1.44276)	(-8.11493)	(-13.5212)
K_L			0.002858	0.177591	-9.9747***
			(-0.023486)	(-1.455)	(-4.20679)
LnTR				-0.04931	-0.11315***
				(-0.5181)	(-3.65109)
K_TR					0.249181***
					(-3.687336)

5. CONCLUSION

A bulk of studies have endeavored to examine the linkages between energy consumption and economic growth, however, no consensus has emerged. The study investigates dynamic relationships between economic growth and energy consumption via incorporating different variables such as trade openness, financial development, urbanization. Four different models are estimated, first three models are estimated for energy use, whereas, model 4 is estimated for energy intensity. The study employs ARDL bound test approach to discover long run relationships and concludes that there exists long run relationship for all four models. It concludes that trade openness positively related to energy use that is when country engage in trade it needs production of goods to export which leads industries to produce more and consume more energy while urbanization impacts negatively energy use for Pakistan suggesting that in urban areas are likely to adopt energy efficient technology. Economic growth is shown to have larger and positive impact on energy use, while financial development has negative impact on energy use. Since it is likely that financial development leads to energy and cost-efficient technologies in practical use. Among shares of economy, agriculture and manufacturing share has positive impact on energy use because these sectors need energy to produce. However, services share is shown to have negative effect on energy use, it leads to decrease in energy use. Capital to labor ratio and comparative advantage impact energy use negatively and positively.

The policy makers around country can look for the empirical results of this study, since it provides stages of energy use and economic growth relationship. We have witnessed a huge significant positive impact of GDP on energy use, suggesting that as GDP grows it significantly increases energy consumption. We have also found

inverted U-shaped relationship between GDP and energy use, indicating that initially, as GDP grows it leads to significant increase in energy use and after achieving certain point GDP grows but energy use tends to decline. However initial impact is larger. For Pakistan, it is unaffordable to lose or restrict growth since it is main driver of development, therefore, use of cleaner and pollution-efficient energy should be promoted all over country to mitigate negative and hazardous outcomes occurring because of massive consumption of energy usage. Moreover, government of Pakistan should consider above situation (stages) while devising and policies related to energy.

Trade openness and urbanization have negative significant impact on energy use, indicating that trade brings energy efficient and eco-friendly technology, therefore, trade should be promoted, and government should design policies to increase our trade with other countries. While urbanization leads to improvement in efficient use of public infrastructure, such as local public transport, in this way it lowers energy use, thus energy use causing pollution can be reduced if government takes serious measures to improve quality of public infrastructure. Financial development is also seen to lower use of energy, argument is well-developed financial markets accelerate home investment which attracts foreign inflows along with know-how and advanced and energy-efficient technology, thus reducing energy use by improving energy efficiency. Policy makers should pay heed to encourage loans and attempt to boost financial markets, which is also good for development.

Policy makers should also take into consideration economy's sectors, i-e agriculture, manufacturing and services. Agriculture sector and manufacturing sector are seen to increase energy use while services sector is seen to lower energy use, Government should introduce energy efficient and advanced technology and different sources for energy in agriculture and manufacturing sectors to save energy resources and usage. Growth rate of capital lowers energy use, since as capital grows, it is possible it grows with advancements of technologies, so government may target on capital, and policy may be devised to promote growth of capital which ultimately would lower use of energy. In long run, emphasis should be given to adopting energy saving methods, such as energy mitigation and energy mix choices, investment in renewable energy resources should also be focused. The major goal should be to achieve efficiency in overall energy use by improving energy infrastructure and promoting financial development, trade openness.

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REFERENCES

- Akarca, A.T. and I.T.V. Long, 1979. Energy and employment: A time-series analysis of the causal relationship. *Resources and Energy*, 2(2-3): 151-162. Available at: [https://doi.org/10.1016/0165-0572\(79\)90027-6](https://doi.org/10.1016/0165-0572(79)90027-6).
- Aqeel, A. and M.S. Butt, 2001. The relationship between energy consumption and economic growth in Pakistan. *Asia-Pacific Development Journal*, 8(2): 101-110.
- Asafu-Adjaye, J., D. Byrne and M. Alvarez, 2016. Economic growth, fossil fuel and non-fossil consumption: A pooled mean group analysis using proxies for capital. *Energy Economics*, 60: 345-356. Available at: <https://doi.org/10.1016/j.eneco.2016.10.016>.
- Bekaert, G. and C.R. Harvey, 2000. Foreign speculators and emerging equity markets. *The Journal of Finance*, 55(2): 565-613.
- Brown, R.L., J. Durbin and J.M. Evans, 1975. Techniques for testing the constancy of regression relationships over time. *Journal of the Royal Statistical Society: Series B (Methodological)*, 37(2): 149-163. Available at: <https://doi.org/10.1111/j.2517-6161.1975.tb01532.x>.
- Çoban, S. and M. Topcu, 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>.

- Engle, R.F. and C.W. Granger, 1987. Co-integration and error correction: Representation, estimation, and testing. *Econometrica: Journal of the Econometric Society*, 55(2): 251-276. Available at: <https://doi.org/10.2307/1913236>.
- Erol, U. and E.S. Yu, 1987. On the causal relationship between energy and income for industrialized countries. *The Journal of Energy and Development*, 13(1): 113-122.
- Fan, J.-L., Y.-J. Zhang and B. Wang, 2017. The impact of urbanization on residential energy consumption in China: An aggregated and disaggregated analysis. *Renewable and Sustainable Energy Reviews*, 75: 220-233. Available at: <https://doi.org/10.1016/j.rser.2016.10.066>.
- Farhani, S. and S.A. Solarin, 2017. Financial development and energy demand in the United States: New evidence from combined cointegration and asymmetric causality tests. *Energy*, 134: 1029-1037. Available at: <https://doi.org/10.1016/j.energy.2017.06.121>.
- Furuoka, F., 2015. Financial development and energy consumption: Evidence from a heterogeneous panel of Asian countries. *Renewable and Sustainable Energy Reviews*, 52: 430-444. Available at: <https://doi.org/10.1016/j.rser.2015.07.120>.
- Jena, P.R. and U. Grote, 2008. Growth-trade-environment nexus in India. Available from <https://www.econstor.eu/handle/10419/39895>.
- Johansen, S. and K. Juselius, 1990. Maximum likelihood estimation and inference on cointegration—with applications to the demand for money. *Oxford Bulletin of Economics and Statistics*, 52(2): 169-210. Available at: <https://doi.org/10.1111/j.1468-0084.1990.mp52002003.x>.
- Kalimeris, P., C. Richardson and K. Bithas, 2014. A meta-analysis investigation of the direction of the energy-GDP causal relationship: Implications for the growth-degrowth dialogue. *Journal of Cleaner Production*, 67: 1-13. Available at: <https://doi.org/10.1016/j.jclepro.2013.12.040>.
- Kraft, J. and A. Kraft, 1978. On the relationship between energy and GNP. *The Journal of Energy and Development*: 401-403.
- Ling, C.H., K. Ahmed, R.B. Muhamad and M. Shahbaz, 2015. Decomposing the trade-environment nexus for Malaysia: What do the technique, scale, composition, and comparative advantage effect indicate? *Environmental Science and Pollution Research*, 22(24): 20131-20142. Available at: <https://doi.org/10.1007/s11356-015-5217-9>.
- Metcalf, G.E., 2008. An empirical analysis of energy intensity and its determinants at the state level. *Energy Journal*, 29(3): 3-1.
- Mirza, F.M., N. Fatima and K. Ullah, 2019. Impact of China-Pakistan economic corridor on Pakistan's future energy consumption and energy saving potential: Evidence from sectoral time series analysis *Energy Strategy Reviews*, 25: 34-46. Available at: <https://doi.org/10.1016/j.esr.2019.04.015>.
- Nasreen, S., S. Anwar and I. Ozturk, 2017. Financial stability, energy consumption and environmental quality: Evidence from South Asian economies. *Renewable and Sustainable Energy Reviews*, 67: 1105-1122. Available at: <https://doi.org/10.1016/j.rser.2016.09.021>.
- Pesaran, M.H., 1997. The role of economic theory in modelling the long run. *The Economic Journal*, 107(440): 178-191. Available at: <https://doi.org/10.1111/1468-0297.00151>.
- Pesaran, M.H. and Y. Shin, 1998. An autoregressive distributed-lag modelling approach to cointegration analysis. *Econometric Society Monographs*, 31: 371-413. Available at: <https://doi.org/10.1017/ccol521633230.011>.
- Pesaran, M.H., Y. Shin and R.J. Smith, 2001. Bounds testing approaches to the analysis of level relationships. *Journal of Applied Econometrics*, 16(3): 289-326. Available at: <https://doi.org/10.1002/jae.616>.
- Sadorsky, P., 2011. Financial development and energy consumption in central and eastern European frontier economies. *Energy Policy*, 39(2): 999-1006. Available at: <https://doi.org/10.1016/j.enpol.2010.11.034>.
- Sbia, R., M. Shahbaz and H. Hamdi, 2014. A contribution of foreign direct investment, clean energy, trade openness, carbon emissions and economic growth to energy demand in UAE. *Economic Modelling*, 36: 191-197. Available at: <https://doi.org/10.1016/j.econmod.2013.09.047>.
- Tugcu, C.T. and M. Topcu, 2018. Total, renewable and non-renewable energy consumption and economic growth: Revisiting the issue with an asymmetric point of view. *Energy*, 152: 64-74. Available at: <https://doi.org/10.1016/j.energy.2018.03.128>.

Voigt, S., E. De Cian, M. Schymura and E. Verdolini, 2014. Energy intensity developments in 40 major economies: Structural change or technology improvement? Energy Economics, 41: 47-62. Available at: <https://doi.org/10.1016/j.eneco.2013.10.015>.

Yu, E.S. and J.-Y. Choi, 1985. The causal relationship between energy and GNP: An international comparison. The Journal of Energy and Development, 10(2): 249-272.

APPENDIX

Table-11. Variables summary.

Indicator name	Long definition	Unit	Source
Energy use	“Energy use refers to use of primary energy before transformation to other end-use fuels, which is equal to indigenous production plus imports and stock changes, minus exports and fuels supplied to ships and aircraft engaged in international transport”.	(kg of oil equivalent per capita)	IEA Statistics OECD/IEA (http://www.iea.org/stats/index.asp), subject to https://www.iea.org/t&c/termsandconditions/
Energy intensity	“Ratio of energy consumption to gross domestic product”.	kt of CO ₂ equivalent	WDI
GDP per capita	“GDP per capita is gross domestic product divided by midyear population. GDP is the sum of gross value added by all resident producers in the economy plus any product taxes and minus any subsidies not included in the value of the products”.	(constant 2010 US\$)	WDI
Trade	“Trade is the sum of exports and imports of goods and services measured as a share of gross domestic product.”	(% of GDP)	WDI
Urban population growth	“Urban population refers to people living in urban areas as defined by national statistical offices. It is calculated using World Bank population estimates and urban ratios from the United Nations World Urbanization Prospects”.	(% annual)	WDI
Domestic credit to private sector	“Domestic credit to private sector refers to financial resources provided to the private sector by financial corporations, such as through loans, purchases of nonequity securities, and trade credits and other accounts receivable, that establish a claim for repayment”.	(% of GDP)	IFS
Manufacturing, value added	“Manufacturing refers to industries belonging to ISIC divisions 15-37.”	(constant 2010 US\$)	WDI
Agriculture, value added	“Agriculture corresponds to ISIC divisions 1-5 and includes forestry, hunting, and fishing, as well as cultivation of crops and livestock production.”	(constant 2010 US\$)	WDI
Services, etc., value added	“Services correspond to ISIC divisions 50-99. They include value added in wholesale and retail trade (including hotels and restaurants), transport, and government, financial, professional, and personal services such as education, health care, and real estate services. Also included are imputed bank service charges, import duties, and any	(constant 2010 US\$)	WDI

	statistical discrepancies noted by national compilers as well as discrepancies arising from rescaling. “		
Labor force participation rate, total	“Labor force participation rate is the proportion of the population ages 15 and older that is economically active: all people who supply labor for the production of goods and services during a specified period”.	(% of total) (modeled ILO estimate)	WDI
Gross fixed capital formation	“Gross fixed capital formation (formerly gross domestic fixed investment) includes land improvements (fences, ditches, drains, and so on); plant, machinery, and equipment purchases; and the construction of roads, railways, and the like, including schools, offices, hospitals, private residential dwellings, and commercial and industrial buildings.”	(constant 2010 US\$)	WDI
Capital-Labor ratio	“capital is divided by labor force to get capital-labor ratio”	(%)	WDI
Comparative advantage	“Capital-labor ratio multiplied by Trade openness to get comparative advantage”	(%)	WDI
Gross fixed capital formation	“Average annual growth of gross fixed capital formation based on constant local currency. Aggregates are based on constant 2010 U.S. dollars.”	(annual % growth)	WDI

Table-12. Diagnostic test results.

Model (1)		
Test	F-statistics	Prob.
Jarque-Bera test	0.735054	0.692445
ARCH test for Hetero	0.787691	0.4685
Autocorrelation LM test	19.95133	0.0022
Model (2)		
Jarque-Bera test	1.344169	0.510643
ARCH test for Hetero	0.356434	0.7045
Autocorrelation LM test	3.492315	0.1326
Model (3)		
Jarque-Bera test	3.380483	0.184475
ARCH test for Hetero	0.019048	0.9811
Autocorrelation LM test	2.237430	0.1773
Model (4)		
Jarque-Bera Test	1.462779	0.481240
ARCH test for Hetero	0.621148	0.6549
Autocorrelation LM test	6.684165	0.2814

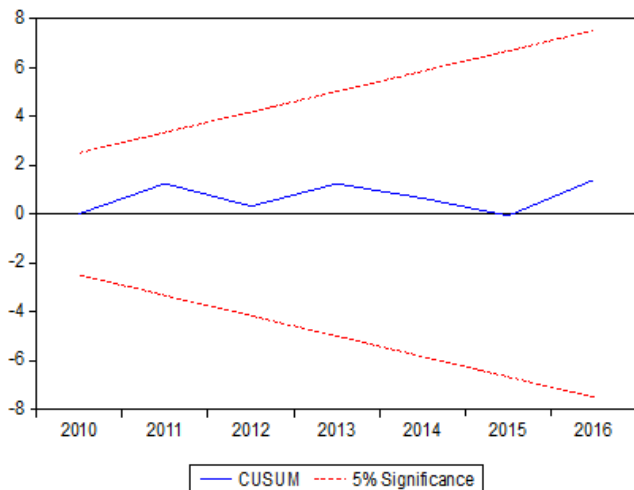


Figure.(a) CUSUM test (Model 1).

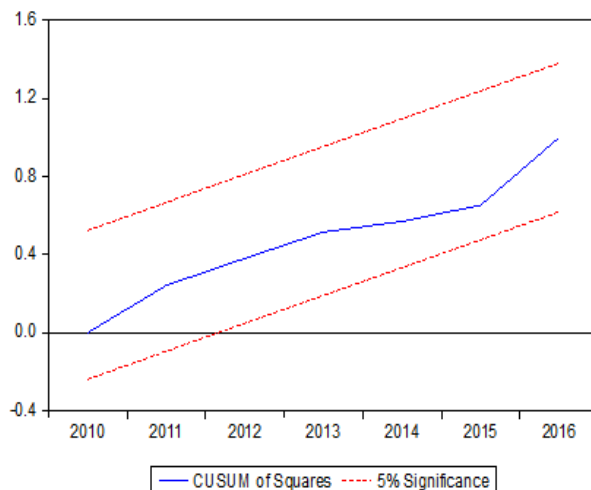


Figure. (b) Squares test (Model 1).

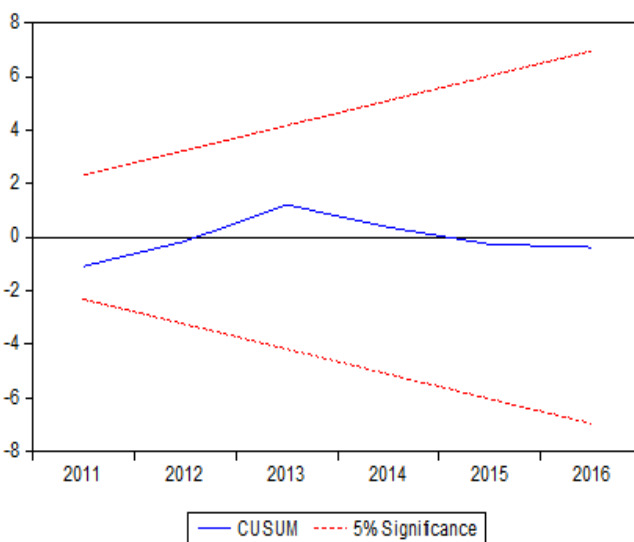


Figure.(c) CUSUM test (Model 2).

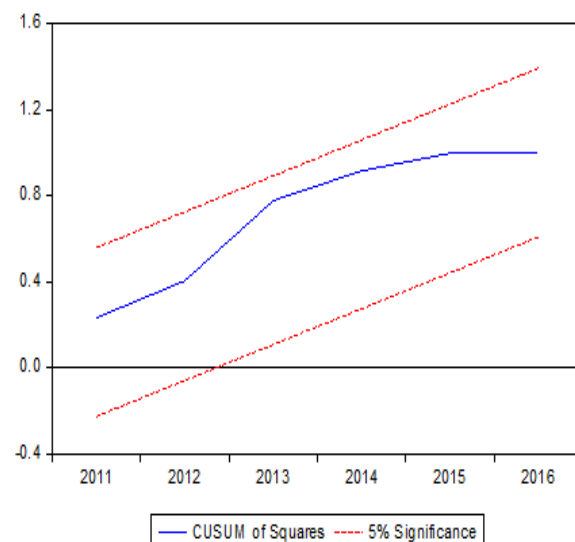


Figure.(d) Squares test (Model 2).

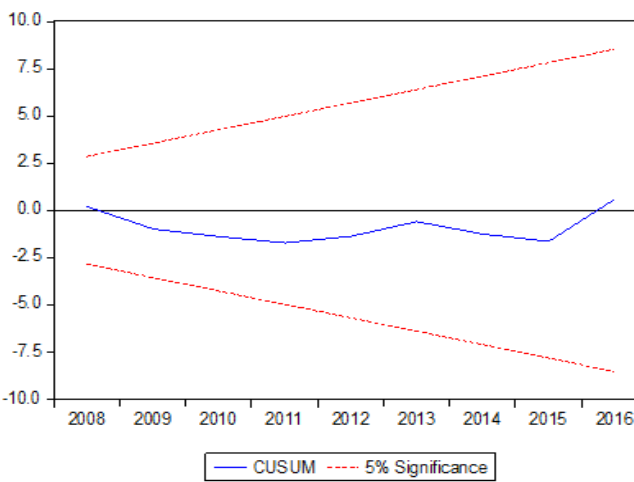


Figure.(e) CUSUM test (Model 3).

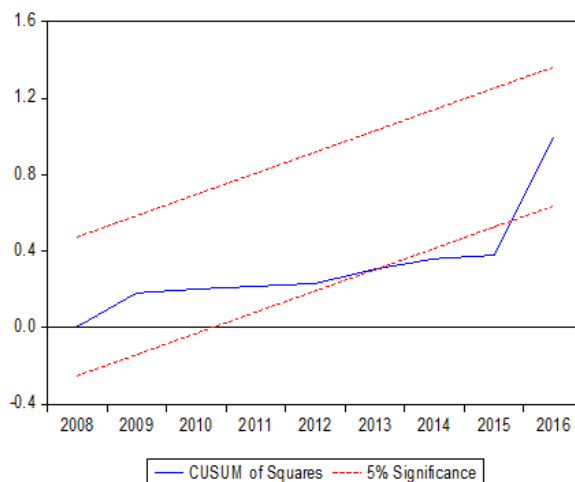


Figure.(f) Squares test (Model 3).

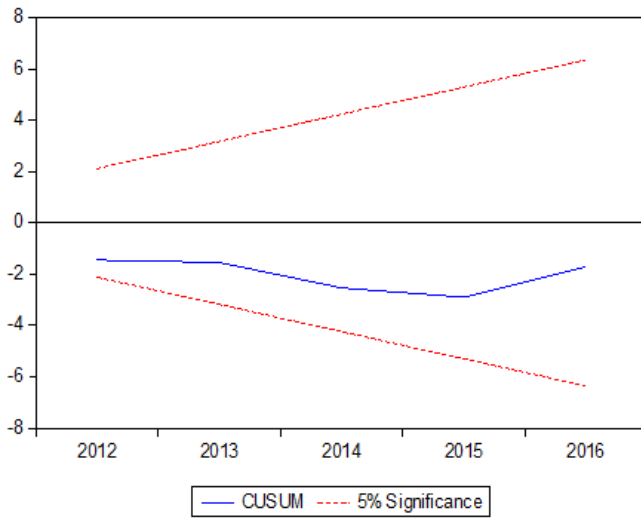


Figure.(g) CUSUM test (Model 4).

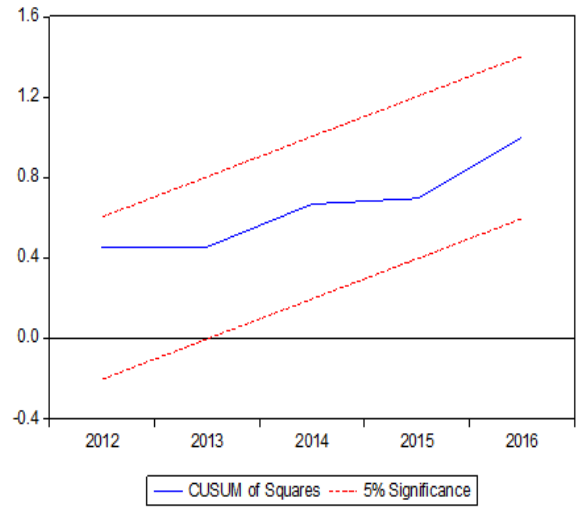


Figure.(h) Squares test (Model 4).

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