### Is the long-run relationship between economic growth, electricity consumption, carbon dioxide emissions and financial development in Gulf Cooperation Council Countries robust?

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#### Abstract

The relationship between carbon dioxide emissions, economic growth, electricity consumption and financial development in the Gulf Cooperation Council (GCC) countries is investigated in this study using panel data for the period of 1980-2012. A number of econometric techniques: dynamic ordinary least squares (DOLS), fully modified ordinary least squares (FMOLS) and the dynamic fixed effect model (DFE) are applied in order to estimate the long-run relationship between the variables. The long-run relationship is found to be robust across these different econometric specifications. No significant short-run significant relationship was observed. Electricity consumption and economic growth have a positive long run relationship with carbon dioxide (CO<sub>2</sub>) emissions whilst a negative and significant relationship was found between CO<sub>2</sub> emissions and financial development. The findings imply that electricity consumption and economic growth stimulate CO<sub>2</sub> emissions in GCC countries while financial development reduces it. Granger causality results reveal that there is a bidirectional causal link between economic growth and CO<sub>2</sub> emissions and a unidirectional causal link running from electricity consumption to CO<sub>2</sub> emissions. However, there is no causal link between financial development and CO<sub>2</sub> emissions. Also, impulse response and variance decomposition analysis outline forecasted impacts of economic growth and electricity consumption on future CO<sub>2</sub> emissions.

**Keywords:** CO<sub>2</sub> emissions, economic growth, electricity consumption, GCC countries, financial development **JEL codes:** Q32, Q43

#### **1. Introduction**

Economic development through economic growth is characterized by the usually close relationship with increasing levels of electricity usage and energy more generally, as well as an associated increase in carbon dioxide emissions ( $CO_2$ ). Additionally, an increase in the level and breadth of a country's financial markets to fund economic development is commonly witnessed. The extent of the influence and the interrelationship of these variables does though vary from country to country.

In this study these relationships are investigated for the six Gulf Cooperation Council countries (GCC): Saudi Arabia, United Arab Emirates (UAE), Qatar, Bahrain, Kuwait and Oman. All of these countries have experienced rapid economic growth over the past 40 years due mainly to their vast oil and gas reserves. These include approximately 40% of the world's proven oil reserves and approximately 25% of the world's natural gas reserves. They contribute approximately 8% of world  $CO_2$  emissions (Al-Saleh et al. 2012). Given the pivotal role of oil and gas in driving the world economy and the rapid increase in economic activity worldwide over the past 40 years, it is perhaps unsurprising that the GCC countries are amongst the highest per capita carbon dioxide ( $CO_2$ ) country emitters (Hertog and Luciani, 2009).

The rapid level of economic development in the GCC countries has been associated with high rates of economic growth, electricity consumption and  $CO_2$  emissions. Rates of economic growth and per capita electricity consumption have surpassed the levels of the major developed economies of the Organization of Economic Cooperation and Development countries (OECD) (World Bank, 2013).

The impact of the 1970s oil shocks upon the energy-mix of the major developed economies was such that it negated the otherwise strong previous association between  $CO_2$  emissions and economic growth as they sought to insulate their economies from high oil and gas prices. This influence was absent in the GCC countries as their governments used their cheap and abundant energy to rapidly develop their economies. This rapidly growing domestic demand for energy in the GCC countries was particularly evidenced for electricity. This situation has now led to three of the six GCC countries being the world's highest  $CO_2$  emitters. Given this scenario, the regions' commitment to sustainable energy policies appears to be a priority.

This study investigates the short and long-run relationships amongst economic growth, electricity consumption,  $CO_2$  emissions and financial development in the region and also determines the causal direction between the variables. Relatively little attention has been paid to the environmental sustainability of this region despite their being significant sources of global energy supply and the potential impacts of this consumption on the environment. This study is an attempt to fill this gap and more importantly, offers a discussion on policy options to achieve sustainability in regional energy systems.

This study contributes primarily by focusing exclusively on the GCC countries and a discussion of solutions to ensure environmental sustainability in the region. A secondary contribution is methodological as it applies a number of sophisticated econometric techniques: the dynamic ordinary least squares (DOLS), the fully modified ordinary least squares (FMOLS) and the dynamic fixed effect model (DFE) to estimate the long-run relationship between the variables. Also panel Granger causality is employed to determine the causal direction between the variables. The robustness of the causal link is checked by the Innovation Accounting Approach (IAA) that consists of impulse response functions and variance decompositions. Therefore, the objective of this study is to assess the interrelationship of the variables of interest in the GCC countries over the period 1980-2012.

The rest of the paper is structured as follows: section 2 contains the literature review while section 3 is dedicated to the discussion of the model and the estimation methods used.

Section 4 presents the results whilst section 5 offers a discussion of the results. The paper ends with conclusions, policy implications and recommendations in section 6.

### 2. Literature Review

### 2.1 Literature review on CO<sub>2</sub> emissions, energy consumption and economic growth

The recent literature has focused on the relationship between energy consumption, environmental pollution and economic growth. Many panel data and time series studies have been done on the relationship between these variables. Saboori et al. (2014) investigated and estimated the bi-directional long-run relationship between energy consumption, carbon dioxide (CO<sub>2</sub>) emissions and economic growth in the road transport sector of all OECD countries. Using the Fully Modified Ordinary Least Squares (FMOLS) method, this study confirmed that there is a positive significant bi-directional relationship between CO<sub>2</sub> emissions and economic growth, road sector energy consumption and economic growth, and between CO<sub>2</sub> emissions and road sector energy consumption. Also the authors found that most of the  $CO_2$  emissions occurred as a result of energy consumption. In addition, the study stressed the need to shift to other options for energy, such as biofuel, renewable and nuclear energy, and the importance of long-run policies that aim to enhance energy efficiency. Hamdi et al. (2014) examined the relationship between electricity consumption, foreign direct investment, capital and economic growth in the case of Bahrain. Their causality analysis supported the feedback effect between electricity consumption and economic growth. Cowan et al. (2014), in a study of the BRICS countries, found support for the neutrality hypothesis for Brazil, India and China, indicating that there is no association between electricity consumption and economic growth. However, regarding the GDP-CO<sub>2</sub> emissions relationship, a feedback hypothesis for Russia, that is a one-way Granger causality running from GDP to CO<sub>2</sub> emissions, and a reverse relationship were found for Brazil both resulting in inconclusive policy implications.

Sbia et al. (2014) investigated the empirical relationship between foreign direct investment, clean energy, trade openness, carbon emissions and economic growth using time series data for the United Arab Emirates (UAE). They found that energy demand had negative relationship with foreign direct investment, trade openness and carbon emissions while economic growth and clean energy stimulated energy consumption. In the case of Saudi Arabia, Alkhatlan and Javid's (2013) study revealed a positive relationship between CO<sub>2</sub> emissions and economic growth. They also concluded that electricity produces less pollution than other sources of energy. Hamdi and Sbia (2013) examined the causal relationship between carbon dioxide emissions, energy consumption and real output for a panel of Gulf Cooperation Council (GCC) countries namely Bahrain, Kuwait, Saudi Arabia, Qatar and UnitedArab Emirates over the period 1980-2009. Their empirical exercise supported the presence of Environmental Kuznets Curve (EKC) hypothesis for these countries only in the long-run. They also found bidirectional causality between carbon emissions and energy usage in the short-run. Ozcan (2013) tested the EKC hypothesis for 12 Middle East countries using panel data for the period 1990-2008. The study provided evidence for a U-shaped EKC for five countries and an inverted U-shaped curve for three countries. No causal link between income and CO<sub>2</sub> emissions was observed for the other four countries. The direction of causality was mixed for different countries. Al-Mulali and Tang (2013) tested the pollution haven hypothesis for the GCC countries. In other words, they investigated the effect of FDI on CO<sub>2</sub> emissions. Their results suggested that increased FDI reduces CO<sub>2</sub> emissions. Also the study reported that energy consumption and GDP growth stimulate CO<sub>2</sub> emissions. Liao and Cao (2013), in a large panel of 132 countries, revealed that factors like urbanisation, population density, trade, energy mix, and economic environment affect the level of CO<sub>2</sub> emissions.

Al-Mulali (2012) undertook a large time series study involving seven different regions – East Asia and Pacific; Eastern Europe and Central Asia; Latin America and the Caribbean; Middle East and North Africa; South Asia; Sub-Saharan Africa; and Western Europe. Using the Fully Modified Ordinary Least Squares (FMOLS) method, the authors investigated the relationship between urbanization, energy consumption and  $CO_2$  emissions. Their results indicated a positive long-run relationship between the variables in six of the regions, while findings varied for the remaining region. Some of the regions demonstrated a negative relationship between the variables, while others – especially the low income countries – did not show any relationship at all. Ozturk and Acaravci (2010) studied the relationship between income, energy consumption,  $CO_2$  emissions and employment in Turkey. They found that neither  $CO_2$  emissions per capita nor energy consumption per capita Granger-cause real GDP growth per capita.

The literature studying the relationship between CO<sub>2</sub> emissions, energy consumption and economic growth involving only the GCC countries is relatively scarce, although there have been several studies on the APEC (Association of Petroleum Exporting Countries), as well as the MENA (Middle East and North African) countries, and high income countries which include all or some of the GCC countries. The study of Omri (2013) on 14 MENA countries found that there is a bi-directional causal link between energy consumption and economic growth in the region. Ozcan (2013) tested 12 Middle East countries and found the EKC hypothesis supported in only three countries and in a further six countries it found no support and no causal link was found in the other three countries. Arouri et al. (2012), in a study of 12 MENA countries, showed that energy consumption has a positive and significant impact on CO<sub>2</sub> emissions and that real GDP demonstrates a quadratic relationship with CO<sub>2</sub> emissions. Narayan and Popp (2012) found that the EKC hypothesis was not supported for a panel of 43 countries including Middle Eastern countries and observed that for the Middle Eastern panel, the income elasticity in the long run is smaller than the short run estimate implying that an increase in income causes a decline in CO<sub>2</sub> emissions. Jaunky (2011) tested the EKC hypothesis for 36 high income countries including three MENA countries – Bahrain, Oman and UAE, with the results indicating that CO<sub>2</sub> emissions decline with a rise in income in the long run. Soytas and Sari (2009) investigated the association between carbon emissions, income, energy and total employment in five selected OPEC countries, including Saudi Arabia, and found a cointegrating relationship between the variables.

#### 2.2 Literature review on CO<sub>2</sub> emissions and financial development

Chang (2015) examined the non-linear effects of financial development and income on energy consumption. The study used five indicators of financial development for a panel of 53 countries for the period 1999-2008. The sample was split into two regimes: high income and non-high income countries. The findings indicate that energy consumption increases with higher levels of financial development when financial development is measured as the share of GDP of private and domestic credit. Ziaei (2015) investigated the effects of two indicators of financial development (credit and stock markets) on energy consumption and CO<sub>2</sub> emissions. The results reveal that financial development reduces CO<sub>2</sub> emissions when the stock market is considered as an indicator of financial development. Boutabba (2014) examined the long-run equilibrium relationship and causal link among CO<sub>2</sub> emissions, financial development, energy consumption and trade openness for India. The findings indicate a positive significant long-run impact of financial development on CO<sub>2</sub> emissions, that is, financial development increases CO<sub>2</sub> emissions in an unidirectional causal link running from financial development to CO<sub>2</sub> emissions.Omri et al. (2014) investigated the causal link between economic growth, financial development and CO<sub>2</sub> emissions in a global panel of 54 countries. Their findings indicate a bi-directional causal link between economic

growth and  $CO_2$  emissions for the sub panels of Middle East, North Africa and Sub Saharan countries and a unidirectional causality running from  $CO_2$  emissions to economic growth for other regions.

Ozturk and Acaravci (2013) found that an increase in foreign trade to GDP ratio results an increase in per capita CO<sub>2</sub> emissions and the financial development variable has no significant effect on per capita CO<sub>2</sub> emissions in the long run for Turkey-. These results also support the validity of the EKC hypothesis in the Turkish economy. Shahbaz et al. (2013a) in a time series study applied the ARDL bounds testing approach to cointegration to examine the influence of financial development on CO<sub>2</sub> emissions in Malaysia. Their findings indicate a positive and significant relationship between the variables. Their findings imply that an economy with more developed financial markets tend to attract more investment and thus facilitate more industrialization which contributes towards higher level of energy consumption eventually leading to higher level of CO<sub>2</sub> emissions. Shahbaz et al. (2013b) in another study investigated the relationship amongst economic growth, energy consumption, financial development, trade openness and CO<sub>2</sub> emissions in Indonesia. Their results confirms a long-run cointegrating relationship among the variables. The study further reports that financial development reduces CO<sub>2</sub> emissions, in other words, financial development improves environmental degradation levelsIn a recent panel study (Al-Mulali et al., 2014) found that financial development was one of the factors that increased energy consumption in GCC countries. The results further observed a cointegrating relationship between GDP, urbanization, total trade and financial development. Financial development was found to stimulate energy consumption and CO<sub>2</sub> emissions in sub Saharan African countries (Al Mulali, 2012) through an increase of investment in energy intensive industries whilst Shahbaz and Lean (2012) obtained similar results for Tunisia.

Zhang (2011) investigated the impact of financial development on  $CO_2$  emissions for China and found financial development was a significant factor. Sadorsky (2011) examined the effect of financial development on energy consumption for a panel of nine Central and Eastern European economies. The findings supported the positive influence of financial development on energy consumption. Sadorsky observed similar findings in an earlier study (Sadorsky, 2010) that investigated the effect of financial development on energy consumption in emerging economies. Tamazian and Rao (2010) recognized financial development as an important driver of environmental performance. They argued that a more financially developed market would provide more resources for environmental projects at a cheaper price. Tamazian et al. (2009) found that a high degree of financial development is associated with better environmental conditions. Jalil and Feridun (2011) found that financial development reduces CO<sub>2</sub> emissions whilst Zhang and Cheng (2009) found the opposite in these two China studies. Yuxiang and Chen (2010) argued that a country with a more developed and sound financial system would enable industries to adopt and use advanced state-of-the-art technologies that are less carbon intensive. They further suggested that financial development helps economies enforce environmentally friendly regulations.

Overall the review on the relationship between financial development and  $CO_2$  emissions suggests that the results are mixed although most of the investigations support the view that higher levels of financial development is positively associated with declining levels of  $CO_2$  emissions. As seen from the literature, there is a limited number of studies for the GCC countries on this issue. Therefore, this study aims to fill this gap in the literature.

#### **3.** Empirical model and econometric methods

An econometric model of the following form is estimated:

$$C_{it} = \beta_i + \beta_1 \ln E_{it} + \beta_2 \ln Y_{it} + \beta_3 F D_{it} + \varepsilon_{it}$$
(1)

The coefficients,  $\beta_1$  and  $\beta_2$  represent the long run elasticity estimates of CO<sub>2</sub> emissions with respect to energy consumption and per capita GDP, as an increase in electricity consumption and income are expected to cause an increase in CO<sub>2</sub> emissions. The effect of financial development on CO<sub>2</sub> emissions cannot be anticipated at this stage as the literature offers inconclusive evidence about this relationship.

To estimate the model, the following actions were taken in a step wise process: (i) a cross-sectional dependence (CD) test was performed to verify whether there is cross-sectional dependence across the panel; (ii) once cross-sectional dependence is observed, an appropriate panel unit root test (i.e. CIPS) was conducted to examine the stationarity of the series; (iii) the Pedroni cointegration test which verifies the long-run relationship among the variables was then conducted; (iv) panel DOLS and panel FMOLS were employed to estimate the long-run relationship while the DFE estimation technique was applied to estimate the short-run and long-run relationships among the variables v) a VECM Granger causality test was conducted to assess causality between the variables and finally vi) the robustness of the causal direction of the relationship was checked by using an Innovation Accounting Approach (IAA) through impulse response functions and variance decomposition analysis.

To investigate the relationships, data for the following variables were sourced:

- Per capita CO<sub>2</sub> emissions (C)
- Per capita electricity consumption (E)
- Per capita real GDP (Y)

- Financial development (FD) - domestic credit available to the private sector as share of GDP.

The World Development Indicators database 2013 was the source of the data for all six countries (World Bank, 2013). Real GDP per capita (Y) which is measured at constant 2000 US\$ was used, per capita electricity use (kWh) and per capita  $CO_2$  emissions were estimated by dividing total electricity and  $CO_2$  emissions by the mid-year population. The variables were then transformed into natural logs. This transformation was intended to overcome the problem of heteroscedasticity between the variables.

#### 3.1. Testing for Unit Roots

It is argued that (Asteriou, 2009) long-run parameters are likely to demonstrate cointegrating relationships among a set of I(1) variables. In other words, it is expected that the macroeconomic variables in the model will be characterized by a unit root process (Nelson and Plosser, 1982). Therefore, determining the order of integration of the variables is the next priority in estimation and the conducting of unit root tests for all variables achieves that aim. Cross-sectional dependence is to be expected amongst this group of <u>six</u> homogenous<u>six</u> countries. An examination of the presence of contemporaneous correlation across the countries was achieved by implementing a cross-sectional dependence (CD) test developed by Pesaran (2004) who defines the CD statistic as:

$$CD = \left[\frac{TN(N-1)}{2}\right]^{1/2} \overline{\hat{\rho}},$$
(2)

where

$$\overline{\hat{\rho}} = \left(\frac{2}{N(N-1)}\right) \sum_{i=1}^{N-1} \sum_{j=i+1}^{N} \hat{\rho}_{ij}$$

in which  $\hat{\rho}_{ij}$  is the pair-wise cross-sectional correlation coefficients of residuals from the conventional ADF regression; T and N are sample and panel sizes, respectively.

Because the CD test indicates the presence of cross-sectional dependence in the panel, the following cross-sectionally augmented Dickey-Fuller (CADF) regression was used:

 $\Delta y_{it} = \alpha_i + \kappa_i t + \beta_i y_{it-1} + \gamma_i \bar{y}_{t-1} + \phi_i \Delta \bar{y}_t + \varepsilon_{it}, t = 1....T \text{ and } i=1....N$ (3) where,

 $\bar{y}_t = N^{-1} \sum_{i=1}^N y_{it}$ 

is the cross-sectional mean of  $y_{it}$ . The purpose of including the cross-sectional mean in the above equation is to control for contemporaneous correlation among  $y_{it}$ . This is a modified version of the IPS test (Im et al., 2003) and is referred to as the cross-sectionally augmented IPS (CIPS) test (Pesaran, 2007). The null hypothesis of the test can be expressed as  $H_0$ :  $\beta_1=0$  for all i against the alternative hypothesis  $H_0$ :  $\beta < 0$  for some i. The test statistic provided by Pesaran (2007) is given as:

$$\operatorname{CIPS}(N,T) = N^{-1} \sum_{i=1}^{N} t_i(N,T)$$

where  $t_i$  (N, T) is the t statistic of  $\beta_i$  in equation (2). The critical values of CIPS (N, T) are available in Table II(c) of Pesaran (2007).

### 3.2. Panel Cointegration

If the results from the CIPS unit root test indicate a cointegrating relationships in the dataset then several panel cointegration tests suggested by Pedroni (1997, 1999) need to be conducted. The Pedroni cointegration test controls for country size and heterogeneity which allows for multiple regressors of the cointegration vector to vary across various panel sections (Pedroni, 1999). Seven panel cointegration statistics for seven tests are obtained. Four are within-dimension tests, whilst three are between-dimension or group statistics approach. The methodology used for the panel cointegration is reproduced in Salahuddin and Gow (2014, 49-50).

#### 3.3. Estimation of panel cointegration regression

If a cointegrating relationship between the variables is found, the next step is to estimate the long-run parameters. Since in the presence of cointegration, OLS leads to spurious coefficients, a number of alternative econometric methods are proposed. One such method is the panel dynamic OLS (DOLS) which is believed to provide better results for cointegrated panels. However, one major weakness of DOLS (Kao and Chiang, 2000) is that it does not consider the cross sectional heterogeneity issue. Pedroni (2000; 2001) proposed the fully modified OLS (FMOLS) estimator for cointegrated panels which takes into account the cross sectional heterogeneity and serial correlation problems. The FMOLS technique is also believed to provide consistent estimates in small samples (Pedroni, 2001).

#### 3.4. Dynamic Fixed Effect Model

One common shortcoming of both the DOLS and FMOLS methods is that they do not estimate short-run relationships (Murthy, 2007). Alternative methods such as the pooled mean group (PMG) regression, mean group regression (MG) and the dynamic fixed effect (DFE) model are available to consider different levels of heterogeneity across countries while estimating both the short-run and the long-run effects simultaneously. The DFE imposes homogeneity restrictions on the long-run and short-run coefficients while allowing the intercept to vary. Since GCC countries are characterized by similar macroeconomic structures (oil-based economies), the application of the DFE model is justified in this case. There may be different types of temporary shocks in different GCC countries due to local laws, regulations and political regimes and this heterogeneity is captured by country-specific intercepts.

In practice, contemporaneous correlation across residuals arises from omitted common factors. To eliminate the influence of these common factors allowance for time-specific effects in the estimated regressions are made. In order to comply with the requirements for

standard estimation and inference, equation 1 is embedded into an ARDL (p, q) model. In error correction form, this can be written as follows:

$$\Delta(y_{i})_{t} = \sum_{j=1}^{p-1} \gamma^{i}_{j} \Delta(y_{i})_{t-1} + \sum_{j=0}^{q-1} \delta^{i}_{j} \Delta(x_{i})_{t-1} + [(y_{i})_{t-1} - \beta^{i}_{1}(X_{i})_{t-1}] + \beta^{i}_{0} + \mu_{t} + \varepsilon_{it}$$
(4)

where  $y_i$  is the dependent variable (CO<sub>2</sub> emissions), Xs are independent variables (electricity consumption, economic growth and financial development),  $\gamma^i_j$  and  $\delta^i_j$  are short run coefficients,  $\beta^i_{1}$  are the long- run coefficients,  $\beta^i_{0}$ ,  $\mu_t$  and  $\epsilon_{it}$  are country-specific fixed effects, time-specific effects and stochastic error term respectively.

#### 3.5 Panel Granger Causality test

If the variables are found to be first difference stationary [I(1)], then to assess the causal direction of the relationship between them further tests are required (Granger, 1969). Information about the exact direction of the causal link enables a more nuanced discussion of the policy implications of the findings (Shahbaz et al., 2012).

#### 3.6 Impulse response and variance decomposition

One major weakness of the VECM Granger causality test is that it is unable to provide reliable estimates of the causal strength of relationship between variables beyond the selected sample period. Another limitation is that it provides only the direction of the relationship, not the corresponding sign. To overcome these limitations, this study applies the Innovation Accounting Approach (IAA) which consists of variance decomposition and generalized impulse response functions. The generalized impulse response function is preferred over the simple Choleski fractionalization impulse response analysis as the generalized impulse response function is insensitive to the order of the VECM. It also indicates whether the impacts of innovations are positive or negative or whether they have a short-run or long-run effect. The general representation of this procedure is available in the seminal works of Sims (1980, 1986) and Bernanke (1986). Although impulse response function traces the effect of a one standard deviation shock on the current and future values of all the endogenous variables through the dynamic structure of VECM, it doesn't provide the magnitude of such effect. Consequently, the variance decomposition method is employed to examine this magnitude.

Variance decomposition (Pesaran and Shin, 1999) measures the percentage contribution of each innovation to h-step ahead of the forecast error variance of the dependent variable and provides a means to determine the relative importance of shocks in explaining the variation in the dependent variable. Engle and Granger (1987) argued that the variance decomposition approach produces more reliable results as compared to those from other traditional approaches.

#### 4. Results

Table 1 reports the summary statistics which shows that the data are well behaved. The standard deviations show that the data are homogeneous.

<Please insert table 1 here>

Table 2 presents the results of the CD test and CIPS unit root test. The CD test results confirms cross sectional dependence in two of the three series (GDPC and energy consumption). The CIPS unit root test proves all variables to be stationary at first difference, i.e. I(1).

<Please insert table 2 here>

The Pedroni panel cointegration test results are presented in table 3. Six of seven tests reject the null hypothesis of no cointegration. The group rho statistic has the best statistical power of all the tests (Gutierrez, 2003), and it also rejects the null of no cointegration. Therefore, there is evidence that there is a long run cointegrating relationship among the variables.

### <Please insert table 3 here>

Table 4 presents the results from the DOLS estimates. The estimates suggest positive and significant long-run relationships of electricity consumption and economic growth with  $CO_2$  emissions. Financial development demonstrates a negative and significant association with  $CO_2$  emissions.

### <Please insert table 4 here>

FMOLS estimates which produced similar results to DOLS but with slightly different coefficient values are reported in table 5. The FMOLS results indicate a positive and highly significant relationship between electricity consumption and economic growth with  $CO_2$  emissions. Financial development has a highly significant negative effect on  $CO_2$  emissions which means financial development reduces  $CO_2$  emissions.

### <Please insert table 5 here>

Table 6 provides the results from the DFE estimation. Overall results suggest that the long run coefficient of  $CO_2$  emissions to electricity consumption is 0.61 and this is significant at the 5% level. In other words, a 1% increase in electricity consumption enhances  $CO_2$  emissions by 0.61% in the long run. There is also a significant and positive long run relationship between economic growth and  $CO_2$  emissions. The long run coefficient of  $CO_2$  emissions to economic growth is 0.40 which means a 1% increase in real GDP per capita causes a 0.40% increase in  $CO_2$  emissions. There is no significant short-run relationship among these variables.

### <Please insert table 6 here>

In table 7, the panel vector error correction model (VECM) Granger causality findings are reported. There is a bidirectional causal link between energy consumption and  $CO_2$  emissions and a unidirectional causal link from economic growth to energy consumption. The relationship between economic growth and  $CO_2$  emissions has no causal link.

### <Please insert table 7 here>

From Figure 1, it can be seen that the standard deviation of per capita  $CO_2$  emissions leads to a positive increase in future per capita  $CO_2$  emissions. The response of per capita  $CO_2$ emissions to the increases in electricity consumption and per capita GDP demonstrate the expected signs but with different magnitudes. The accumulated response of per capita  $CO_2$ emissions to electricity consumption is positive and significant and to GDP per capita is also positive and significant. The response of per capita  $CO_2$  emissions to future shocks of financial development is negative and significant. Thus, these findings are supportive of all earlier econometric estimations.

### <Please insert figure 1 here>

Results from the variance decomposition analysis are reported in Table 8. The study allows for a 32 year forecasting horizon. Interestingly, at the 5-year forecasting horizon, about 94% of the one-step forecast variance in per capita  $CO_2$  emissions is accounted for by its own innovations and altogether 6% is accounted for by economic growth, electricity consumption and financial development. In the long-run, the response to own innovative shocks declines to around 65% while the response of per capita electricity consumption to the shocks in per capita  $CO_2$  emissions, economic growth and financial development are expected to rise to 35% from the first 5-year forecast horizon of 6%. Amongst the 35% of the variance, approximately 23% of variance is due to the shocks in per capita electricity consumption and around 11% variations are attributed to GDP per capita while the rest, 0.76%, is due to the shock in financial development. The findings reinforce that while per capita electricity consumption is likely to have a very strong forecasted impact on per capita  $CO_2$  emissions, the impact of economic growth is also likely to be evident in the future. However, the forecasted impact of financial development seems to be weak.

<Please insert table 8 here>

## 5. Discussion

This study investigated the effects of economic growth, electricity consumption and financial development on CO<sub>2</sub> emissions in GCC countries using panel data for the period of 1980-2012. CIPS panel unit root tests were conducted that account for cross sectional dependence and find that all variables first difference stationary. The Pedroni cointegration test confirms a cointegrating relationship among the variables. Group DOLS andFMOLS were employed to estimate the long-run relationship among the variables. The panel econometric technique, the DFE model, was estimated to examine both the short-run and the long-run relationship between CO<sub>2</sub> emissions and economic growth, electricity consumption and financial development. Group DOLS and FMOLS were also employed to test the robustness of the long-run relationship among the variables. Economic growth and electricity consumption were found to have a positive significant impact on CO<sub>2</sub> emissions in the longrun while no significant short-run relationship between these variables was observed. The findings of the long-run association between electricity consumption and economic growth are in contrast with the results of a recent study on GCC countries (Hamdi and Sbia, 2014). Financial development was found to reduce CO<sub>2</sub> emissions in the long-run. A bi-directional causal link was found between economic growth and CO<sub>2</sub> emissions. This implies that although the GDPs of the GCC countries are largely oil based, their oil based revenues generate high incomes for their citizens and a massive influx of foreign workers both of which leads to a sharp rise in energy demand. To meet the growing energy demand, enormous amounts of electricity are generated, mostly from fossil fuel sources.

A unidirectional causal link running from electricity consumption to  $CO_2$  emissions is found to exist. No causal link was found between financial development and  $CO_2$  emissions. Impulse response functions and variance decomposition analysis reveal that per capita electricity consumption and economic growth will continue to impact  $CO_2$  emissions significantly into the future while the impact of financial development is expected to be of little magnitude. Therefore, the GCC countries will have to look for alternative sources of power generation as well as undertaking measures to reduce  $CO_2$  emissions. The overall results imply that economic growth and electricity consumption contribute towards  $CO_2$ emissions in the GCC countries. No such relationship was found for financial development.

### 6. Conclusions and policy implications

This study aimed to examine the effects of economic growth, electricity consumption and financial development on  $CO_2$  emissions in GCC countries using panel data for the period of 1980-2012. Unit root test that account for cross sectional dependence was conducted. Pedroni cointegration test confirmed a cointegrating relationship between the variables. A panel econometric technique, the DFE model, was estimated to examine both the short-run and the long-run relationship between  $CO_2$  emissions and economic growth, electricity consumption and financial development. Economic growth and electricity consumption were found to have a positive significant impact on  $CO_2$  emissions in the long-run while no significant short-run relationship between these variables was observed. Financial development was found to reduce  $CO_2$  emissions in the long-run. Group DOLS and FMOLS provided evidence in support of the DFE results.

The Granger causality results suggested a bidirectional causal link between economic growth and  $CO_2$  emissions. A unidirectional causal link running from electricity consumption

to  $CO_2$  emissions was found to exist. No causal link was found between financial development and  $CO_2$  emissions. Impulse response functions and variance decomposition analysis revealed that per capita electricity consumption and economic growth would continue to impact  $CO_2$  emissions significantly into the future while the impact of financial development is expected to be of little magnitude. Overall results demonstrated that economic growth and electricity consumption contributed towards  $CO_2$  emissions in the GCC countries. No such relationship was found for financial development.

The findings of this study have very important policy implications for GCC countries for not only to be able to efficiently deal with current climate challenges but also for their post-oil future. Emissions are already causing sea levels to rise and affecting coastlines and marine lives resulting in increasing levels of salinity. This situation will eventually cause a scarcity of the availability of fresh water. The GCC countries are already running a large number of desalinization plants which are very expensive to operate and are also harmful to the environment as they need huge amounts of electricity to run.

Also, as these countries' energy supply is predicted to reduce with the passage of time, the opportunity cost of huge government subsidies on current energy consumption is likely to be more and more financially unsustainable. Although a trade-off between these opportunity costs and the political reality of these countries, which are mostly ruled by monarchies, may be difficult to envisage. Since, these countries are under potential threat as a result of their alarming levels of emissions and their responses to combating emissions appear to be inadequate so far, they can't afford to waste time. There is a need to act promptly to promote energy efficiency and the use of renewable resources, in other words, they must do everything possible to reduce their economies dependence on fossil fuels and to introduce newer more environmentally friendly technologies to meet their energy needs.

Based on the findings of the study, it is suggested that the GCC countries should reduce  $CO_2$  emissions by a variety of measures. There are alternative potential measures for electricity generation that will enable the region to achieve higher levels of energy efficiency. It is already evident that GCC countries can reduce  $CO_2$  emissions and gain energy efficiency in three ways: a) promoting CCUS (carbon capture, utilization and storage) plants, b) promoting the use of renewable resources and c) building nuclear energy plants.

The CCUS method has already proved its potential to reduce  $CO_2$  emissions in the region (Al Saleh et al. 2012). Another recent study of Saudi Arabia also has recognized the potential of CCUS to significantly reduce  $CO_2$  emissions (Mansouri et al. 2013). The GCC countries also have clear advantage over the rest of the world in renewable resources, especially solar and wind energy. The region is characterized by an enormous amount of sunlight and wind and for more than 80% of days in a year its sky is cloud free or clear. The average solar radiation of the region is 2200kWh (th)/m<sup>2</sup> (Hertog, 2010). Therefore, solar and wind are the two most significant potential renewable sources for energy in the region. Although, Saudi Arabia and the UAE have already been pursuing research on this potential, other GCC countries also need to recognize and tap this opportunity. It is already evident that use of solar photovoltaic (PV) can significantly save  $CO_2$  emissions in the electricity sector of Saudi Arabia and the UAE (Mansouri et al., 2013; Mondal et al., 2014). Solar PV is a very good technology option for long term investment in the power sector. It will potentially enable GCC countries to achieve their renewable generation targets (Mondal et al. 2014).

Building nuclear energy plants is another viable option for the GCC countries to combat emissions. Since all these countries' economies are characterized by large foreign capital reserves thanks to their oil revenues, investment in such projects should not be considered too ambitious for them. The UAE has already decided to integrate nuclear energy into its electricity generation portfolio and a recent study by (Alfarra and Abu-Hijley, 2012)

showed through a number of scenario analysis that the use of nuclear energy would not only reduce  $CO_2$  emissions but also reduce per unit electricity generation costs.

Finally, this study recommends that the GCC countries need to significantly boost investment for research in clean energy technologies and build energy expertise. This is not only to address the prevailing climate challenges and meet their current renewable energy targets only but also to deal with further challenges in the post-oil age. Long-term investment in building a university under the potential name of 'GCC University of Energy Research and Technology' could be a vital and sustainable contribution towards the achievement of such goal.

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# Figure 1. Impulse responses of CO<sub>2</sub> emissions to independent variables

|              | LCO2      | LGDPC    | LFD       | LEPC      |
|--------------|-----------|----------|-----------|-----------|
| Mean         | 3.044914  | 10.00954 | 1.534847  | 3.844567  |
| Median       | 3.117565  | 9.864770 | 1.551185  | 3.918153  |
| Maximum      | 4.227340  | 11.31380 | 1.971023  | 4.241647  |
| Minimum      | 1.492100  | 8.711520 | 0.833438  | 2.794986  |
| Std. Dev.    | 0.614045  | 0.640451 | 0.209025  | 0.291089  |
| Skewness     | -0.461721 | 0.116449 | -0.453239 | -1.042773 |
| Kurtosis     | 2.913923  | 1.591193 | 3.473983  | 4.047534  |
|              |           |          |           |           |
| Jarque-Bera  | 6.737876  | 15.97199 | 8.196517  | 42.66683  |
| Probability  | 0.034426  | 0.000340 | 0.016602  | 0.000000  |
|              |           |          |           |           |
| Sum          | 572.4439  | 1881.794 | 288.5513  | 722.7785  |
| Sum Sq. Dev. | 70.50864  | 76.70310 | 8.170278  | 15.84502  |
|              |           |          |           |           |
| Observations | 188       | 188      | 188       | 188       |

## Table 1: Summary statistics of the variables

# Table 2: CIPS unit root test results

|                  | Р     | CD       | CIPS   | CIPS (1 <sup>st</sup> Diff.) |
|------------------|-------|----------|--------|------------------------------|
| LGDPC            | 0.488 | 2.09**   | -1.501 | -2.637***                    |
| LCO <sub>2</sub> | 0.298 | -0.52    | -1.747 | -2.664***                    |
| LEPU             | 0.832 | 18.50*** | -1.283 | -2.860***                    |
| FD               | 0.506 | 11.03*** | -1.811 | -3.006***                    |

# Table 3: Pedroni Residual Cointegration Test

| Alternative hypothesis: common AR coefs. (within-dimension) |           |                  |               |                  |        |
|---|-----------|------------------|---------------|------------------|--------|
|   |           |                  |               | Weighted         |        |
|   |           | <u>Statistic</u> | <u>Prob.</u>  | <u>Statistic</u> | Prob.  |
| Panel v-Statistic   |           | 2.679511         | 0.0037        | 1.777153         | 0.0378 |
| Panel rho-Statistic   |           | -1.588999        | 0.0560        | -0.742916        | 0.2288 |
| Panel PP-Statistic  |           | -2.961571        | 0.0015        | -1.791329        | 0.0366 |
| Panel ADF-Statistic   |           | -1.357293        | 0.0873        | -0.834728        | 0.2019 |
| Alternative hypoth  | esis: ind | ividual AR c     | coefs. (betwe | en-dimensio      | n)     |
|   |           | Statistic        | Prob.         |                  |        |
| Group rho-Statistic   |           | -0.907086        | 0.1822        |                  |        |
| Group PP-Statistic  |           | -2.939609        | 0.0016        |                  |        |
| Group ADF-Statistic   |           | -1.579396        | 0.0571        |                  |        |

| Variable           | Coefficient | Std. Error         | t-Statistic | Prob.    |
|--------------------|-------------|--------------------|-------------|----------|
| LEC                | 0.648209    | 0.100480           | 6.451117    | 0.0000   |
| LGDPC              | 0.390644    | 0.202823           | 1.926033    | 0.0562   |
| LFD                | -0.006200   | 0.002251           | -2.754503   | 0.0067   |
| R-squared          | 0.949922    | Mean dependent var |             | 3.041422 |
| Adjusted R-squared | 0.926752    | S.D. dependent var |             | 0.603062 |
| S.E. of regression | 0.163215    | Sum squared resid  |             | 3.569641 |
| Long-run variance  | 0.039696    |                    |             |          |

Table 4: Results from panel Dynamic Ordinary Least Squares (DOLS) estimation

# Table 5: Results from Fully Modified Ordinary Least Squares (FMOLS) estimation

| Variable           | Coefficient | Std. Error         | t-Statistic | Prob.    |
|--------------------|-------------|--------------------|-------------|----------|
| LGDPC              | 0.405094    | 0.131383           | 3.083311    | 0.0024   |
| LFD                | -0.655508   | 0.168888           | -3.881318   | 0.0001   |
| LEPC               | 1.338018    | 0.169979           | 7.871661    | 0.0000   |
| R-squared          | 0.907085    | Mean dependent var |             | 3.050169 |
| Adjusted R-squared | 0.902738    | S.D. dependent var |             | 0.602068 |
| S.E. of regression | 0.187766    | Sum squared resid  |             | 6.028805 |
| Durbin-Watson stat | 1.597326    | Long-run variance  |             | 0.075695 |

| Table 6: Results from D | vnamic Fixed Effect ( | (DFE) and Mean | Group (MG | ) estimations |
|-------------------------|-----------------------|----------------|-----------|---------------|
|                         |                       |                |           | ,             |

|                                     | Dynamic Fixed Effect |           | Mean Group |           |
|-------------------------------------|----------------------|-----------|------------|-----------|
| Variables                           | Long Run             | Short Run | Long Run   | Short Run |
| Error Correction                    |                      | -0.402*** |            | -0.499*** |
|                                     |                      | (0.0562)  |            | (0.161)   |
| $\Delta$ GDP Per Capita             |                      | 0.217     |            | 0.525***  |
|                                     |                      | (0.254)   |            | (0.158)   |
| $\Delta$ Electric Power Consumption |                      | 0.885***  |            | 0.163     |
|                                     |                      | (0.120)   |            | (0.220)   |
| $\Delta$ Financial Development      |                      | 0.104     |            | 0.0378    |
|                                     |                      | (0.177)   |            | (0.0784)  |
| GDP Per Capita                      | 0.405*               |           | 3.372      |           |
|                                     | (0.013)              |           | (2.570)    |           |
| Electric Power Consumption          | 0.617***             |           | -0.737     |           |
|                                     | (0.107)              |           | (0.850)    |           |
| Financial Development               | -0.131               |           | -0.0354    |           |
|                                     | (0.058)              |           | (0.0761)   |           |
| Constant                            |                      | -2.448*** |            | -4.168**  |
|                                     |                      | (0.785)   |            | (2.044)   |
| Observations                        | 192                  | 192       | 192        | 192       |

| Table 7: Panel VECN | A Granger Causality |
|---------------------|---------------------|
|---------------------|---------------------|

| Dependent variable: D | D(LCO2)  |    |        |
|-----------------------|----------|----|--------|
| Excluded              | Chi-sq   | df | Prob.  |
| D(LEPU)               | 5.287331 | 1  | 0.2215 |
| D(LGDPC)              | 0.329650 | 1  | 0.5659 |
| D(FD)                 | 6.121831 | 1  | 0.0134 |
| All                   | 11.23275 | 3  | 0.0105 |
| Dependent variable: D | O(LEPU)  |    |        |
| Excluded              | Chi-sq   | df | Prob.  |
| D(LCO2)               | 7.352636 | 1  | 0.0067 |
| D(LGDPC)              | 5.152295 | 1  | 0.0232 |
| D(FD)                 | 0.142911 | 1  | 0.7054 |
| All                   | 11.69896 | 3  | 0.0085 |
| Dependent variable: D | D(LGDPC) |    |        |
| Excluded              | Chi-sq   | df | Prob.  |
| D(LCO2)               | 0.882288 | 1  | 0.3476 |
| D(LEPU)               | 1.985224 | 1  | 0.1588 |
| D(FD)                 | 0.867994 | 1  | 0.3515 |
| All                   | 2.547835 | 3  | 0.4667 |
| Dependent variable: D | D(FD)    |    |        |
| Excluded              | Chi-sq   | df | Prob.  |
| D(LCO2)               | 2.618363 | 1  | 0.1056 |
| D(LEPU)               | 2.679576 | 1  | 0.1016 |
| D(LGDPC)              | 0.196280 | 1  | 0.6577 |
| All                   | 3.735083 | 3  | 0.2915 |

|        |          |          |          |          | DOMESTIC_CREDIT_TO_ |
|--------|----------|----------|----------|----------|---------------------|
| Period | S.E.     | LCO2     | LEU      | LGDPC    | PRIVA               |
| 1      | 0.141899 | 100.0000 | 0.000000 | 0.000000 | 0.000000            |
| 2      | 0.184542 | 99.69382 | 0.117580 | 0.146193 | 0.042407            |
| 3      | 0.207631 | 98.49373 | 1.021031 | 0.386978 | 0.098266            |
| 4      | 0.222774 | 96.56938 | 2.526875 | 0.763157 | 0.140583            |
| 5      | 0.234046 | 94.24441 | 4.339306 | 1.250446 | 0.165842            |
| 6      | 0.243119 | 91.78623 | 6.232153 | 1.805594 | 0.176026            |
| 7      | 0.250761 | 89.36774 | 8.066943 | 2.389722 | 0.175592            |
| 8      | 0.257369 | 87.08277 | 9.772029 | 2.975648 | 0.169548            |
| 9      | 0.263173 | 84.97231 | 11.31842 | 3.547021 | 0.162251            |
| 10     | 0.268323 | 83.04643 | 12.70142 | 4.095226 | 0.156919            |
| 11     | 0.272926 | 81.29900 | 13.92888 | 4.616515 | 0.155613            |
| 12     | 0.277064 | 79.71645 | 15.01417 | 5.109955 | 0.159422            |
| 13     | 0.280802 | 78.28272 | 15.97245 | 5.576120 | 0.168714            |
| 14     | 0.284191 | 76.98166 | 16.81866 | 6.016312 | 0.183360            |
| 15     | 0.287276 | 75.79829 | 17.56666 | 6.432122 | 0.202927            |
| 16     | 0.290094 | 74.71916 | 18.22883 | 6.825195 | 0.226808            |
| 17     | 0.292674 | 73.73249 | 18.81608 | 7.197110 | 0.254327            |
| 18     | 0.295044 | 72.82801 | 19.33787 | 7.549329 | 0.284796            |
| 19     | 0.297227 | 71.99686 | 19.80240 | 7.883180 | 0.317560            |
| 20     | 0.299240 | 71.23135 | 20.21677 | 8.199860 | 0.352015            |
| 21     | 0.301102 | 70.52484 | 20.58709 | 8.500444 | 0.387625            |
| 22     | 0.302827 | 69.87154 | 20.91864 | 8.785902 | 0.423923            |
| 23     | 0.304427 | 69.26639 | 21.21600 | 9.057107 | 0.460509            |
| 24     | 0.305915 | 68.70496 | 21.48314 | 9.314856 | 0.497045            |
| 25     | 0.307299 | 68.18335 | 21.72352 | 9.559875 | 0.533255            |
| 26     | 0.308589 | 67.69810 | 21.94016 | 9.792832 | 0.568911            |
| 27     | 0.309793 | 67.24614 | 22.13569 | 10.01434 | 0.603833            |
| 28     | 0.310917 | 66.82473 | 22.31241 | 10.22498 | 0.637879            |
| 29     | 0.311968 | 66.43142 | 22.47235 | 10.42529 | 0.670943            |
| 30     | 0.312951 | 66.06400 | 22.61730 | 10.61576 | 0.702943            |
| 31     | 0.313872 | 65.72048 | 22.74883 | 10.79687 | 0.733827            |
| 32     | 0.314734 | 65.39905 | 22.86833 | 10.96906 | 0.763557            |

 Table 8. Variance Decomposition of CO2 emission for GCC countries: 1980-2012