

# Technological shocks and its influence on renewable energy consumption-green growth: Do financial institutions mitigate some environmental burden?

**Tan Chao**

Hunan Institute of Engineering

**Xu Yunbao**

Hunan Institute of Engineering

**Dai Chengbo**

China University of Geosciences

**Li Bo**

Hunan Institute of Engineering

**Ahmed Usman** (✉ [voice.of.usman.au@gmail.com](mailto:voice.of.usman.au@gmail.com))

Government College University Faisalabad

---

## Research Article

**Keywords:** Technological shocks, Financial institutions, Renewable energy consumption, Green growth

**Posted Date:** March 14th, 2022

**DOI:** <https://doi.org/10.21203/rs.3.rs-1406685/v1>

**License:**   This work is licensed under a Creative Commons Attribution 4.0 International License.

[Read Full License](#)

---



48 use of clean energy (Luukkanen et al., 2019). The technological invention addresses both demand-based and  
49 production-based carbon emissions, hence contributing as a major determinant of industrial advancement (Yao  
50 et al., 2018). Developing, designing, and executing green technologies improve the sustainability of firms (Chen  
51 et al., 2022). Technological progress is required in supplying and generating energy for fulfilling the needs  
52 domestic, industrial, and transport sectors. Technological shocks promote supply chain and procurement  
53 activities of the industrial sector. This shows that technological shocks interconnected with strategic policies are  
54 fundamental for green growth (Wei & Ullah, 2022).

55         Considering the climate change and global warming issues, the economies throughout the world are  
56 adopting cleaner and newer technologies for producing that are cost-saving and eco-friendly (Horbach et al.,  
57 2012). Green technology occurred from investment in the R&D sector that shifts the economic setup towards the  
58 use of renewable energy sources (Mennicken et al. 2016; Alvarez-Herranz et al., 2017; Shuai & Fan, 2020).  
59 Development of human capital and R&D influence renewable energy consumption at both macro and micro  
60 levels. Well-developed human capital prefers to use energy-efficient products and consume clean energy  
61 sources (Li & Ullah, 2021). Moreover, well-developed human capital prefers to use advanced technologies and  
62 transform the economy to renewable and cleaner energy sources (Yin et al., 2021). Well-developed human  
63 capital influences energy consumption through the channel of technological progress and income. Human  
64 capital is strongly attached to an upsurge in output that in turn affects the consumption of energy (Ullah et al.,  
65 2019). Additionally, human capital also influences technological progress that in turn affects energy  
66 consumption (Zhang et al., 2021). Yao et al. (2019) reported that well-developed human capital reduces the  
67 consumption of dirty energy and promotes the consumption of cleaner and renewable energy.

68         A bulk of research has been done on green growth considering all possible determinants of green  
69 growth in the case of different economies (Popp, 2012). Researchers also assessed the nexus of green growth  
70 and CO<sub>2</sub> emissions (Hao et al. 2021). Guo et al (2017) assessed green growth by considering the mutual impacts  
71 of technological shocks and environmental regulations. Stegemann and Ossewaarde (2018) unveiled the myth of  
72 green growth and energy transformation but isolated the impact of green technologies on the transmission and  
73 generation of energy. Studies examined the effect of technological shocks from different perspectives (Ullah et  
74 al., 2021). These studies confirmed the significance of technological shocks in the battle of environmental  
75 degradation. Mensah et al. (2019) examined the impact of technological innovation in renewable energy sources  
76 and reported that technological innovation contributes effectively to ensuring green growth. In literature,  
77 researchers mostly used research and development expenditures in the energy sector as a proxy variable for  
78 technological change. Existing studies examined the role of technological innovations on CO<sub>2</sub> emissions and the  
79 green environment, but the effect of technological shocks on renewable energy consumption and green growth  
80 lacks attention.

81         Greener technology and renewable energy consumption reduce the dependence on fossil fuel sources  
82 that exert a stronger impact on energy markets. Green technological shocks prevent environmental degradation.  
83 However, the transformation from fossil fuel energy to cleaner renewable energy sources is quite challenging.  
84 The major challenge towards the adoption of green and renewable energy sources is the cost. The adoption  
85 process of renewable energy sources is attached with several financial constraints such as high operating, start-  
86 up, and infrastructure costs. Thus, it is necessary to have a strong financial institutional setup that provides  
87 effective methods of funding, price discovery, risk management, and market liquidity. Furthermore, financial

88 institutions increase capital allocation. Well-developed financial institutions enhance investments in emergent  
89 industries, however, bad quality financial institutions reduce investments in emergent industries (He et al. 2019).  
90 Thus, in an economy, where green growth and renewable energy consumption are highly stimulated, the  
91 contribution of financial institutions can be considerable. Among the green growth literature, there is increasing  
92 interest in examining the determinants that enhance the use of renewable energy consumption. A bulk of the  
93 literature has examined the association between economic growth and renewable energy consumption, the effect  
94 of financial institutions on green growth and renewable energy consumption has not received much attention.

95 Besides financial institutions, environmental awareness and education are key determinants for green  
96 environmental and green growth (Zheng et al., 2021; Li et al., 2022). Akerlof (2017) denoted education as a  
97 major policy tool for sustainable green growth. Low investment in the R&D sector and technological innovation  
98 can be accredited to low educational attainment that might affect environmental performance negatively. Low  
99 level of educational attainment results in restriction of renewable energy consumption and restriction of  
100 dissemination of green environment awareness (Hess and Collins, 2018). Thus, educational attainment might  
101 affect the generation of renewable energy and green growth. Grossman and Krueger (1991) highlighted that  
102 education attainment supports policymakers in the formulation of eco-friendly policies that support in the  
103 achievement of sustainable green growth. Thus, the contribution of education in the attainment of green growth  
104 and the use of clean energy cannot be neglected.

105 The primary objective of this study is to explore the effect of technological shocks on green growth and  
106 renewable energy consumption. The study also explores the role of education and financial institutions in  
107 determining green growth and renewable energy consumption. The study empirically investigates the impact of  
108 technological shock, financial institutions, and education on green growth and renewable energy consumption  
109 for BRI economies over the period 1991 to 2019. The study employed a cross-sectional ARDL approach for  
110 empirical exploration. The findings of this study will help in evolving policies related to green growth and  
111 renewable energy consumption in BRI economies. The findings of this study not only expand the awareness but  
112 also highlight determinants to reinforce green growth throughout the world.

113

## 114 **Model and methods**

115 Technological innovations are key to fostering economic growth; therefore, it is widely recognized that they can  
116 affect environmental quality and other related factors such as renewable energy consumption and green growth.  
117 Indeed the relationship between technological innovation and economic growth is widely discussed in the  
118 circles of empirics but the mechanism through which technological innovations can affect economic growth is  
119 yet not finalized. However, theorists have highlighted a few mechanisms that can improve a firm's innovations  
120 such as the 'supply push model of innovation', 'demand-pull model of innovation', sudden advancement in  
121 scientific knowledge, and the models that can be explained in clusters (Geroski and Walters 1995). As far as,  
122 decoupling the economic growth from CO<sub>2</sub> emissions is concerned, Grossman and Krueger (1995) suggested  
123 that technology used in the production process helps reduce carbon emissions after a technological change  
124 occurs endogenously. Modern growth theories have also pointed out the various channels such as patents and  
125 innovations through which a technological change can help to attain sustainable economic development (Wang  
126 et al. 2021) necessary to preserve the environment. Hence, we can confer that technological innovation can  
127 significantly help to decouple economic growth and CO<sub>2</sub> emissions (Ullah et al., 2021). Therefore, to capture

128 the impact of technology shock on green growth and renewable energy consumption we have followed Mensah  
 129 et al. (2019) and constructed the following models:

130  
 131 
$$REC_{it} = \eta_0 + \eta_1 Tech_{it} + \eta_2 FIE_{it} + \eta_3 Education_{it} + \eta_4 GS_{it} + \varepsilon_{it} \text{ ----- (1)}$$

132  
 133 
$$GG_{it} = \eta_0 + \eta_1 Tech_{it} + \eta_2 FIE_{it} + \eta_3 Education_{it} + \eta_4 GS_{it} + \varepsilon_{it} \text{ ----- (2)}$$

134  
 135 Where renewable energy consumption (REC) and green growth (GG) in equations (3 & 4) are dependent on  
 136 technological innovations (Tech), financial institutions' efficiency index (FIE), government spending (GS), and  
 137  $\varepsilon_{it}$  is randomly distributed error term. Estimates obtained by estimating equation (1) reflect the long-run effects  
 138 of green growth. To suppose the short-run effects, we re-write equation (1) in an error-correction format. The  
 139 study uses the following linear CS-ARDL regression equation.

140 
$$\Delta REC_{it} = C_i + \lambda_i (REC_{it-1} - \beta_i X_{it-1} - Y_i GS_{it-1} - \phi_{1i} \overline{REC}_{t-1} - \delta_2 \overline{X}_{t-1} - \pi_2 \overline{GS}_{t-1}) + \sum_{j=1}^{p-1} \theta_{ij} \Delta REC_{it-j} +$$
  
 141 
$$\sum_{j=0}^{q-1} \eta_{ij} \Delta X_{it-j} + \sum_{j=0}^{q-1} \tau_{ij} \Delta GS_{it-j} + \eta_{1i} \Delta \overline{REC}_t + \eta_{2i} \Delta \overline{X}_t + \eta_{3i} \Delta \overline{GS}_t + \varepsilon_{it} \text{ ----- (3)}$$

142  
 143 
$$\Delta GG_{it} = C_i + \lambda_i (GG_{it-1} - \beta_i X_{it-1} - Y_i GS_{it-1} - \phi_{1i} \overline{GG}_{t-1} - \delta_2 \overline{X}_{t-1} - \pi_2 \overline{GS}_{t-1}) + \sum_{j=1}^{p-1} \theta_{ij} \Delta GG_{it-j} +$$
  
 144 
$$\sum_{j=0}^{q-1} \eta_{ij} \Delta X_{it-j} + \sum_{j=0}^{q-1} \tau_{ij} \Delta GS_{it-j} + \eta_{1i} \Delta \overline{GG}_t + \eta_{2i} \Delta \overline{X}_t + \eta_{3i} \Delta \overline{GS}_t + \varepsilon_{it} \text{ ----- (4)}$$

145  
 146 Where,  $X_{it-1}$  is set of independent variable in equation (3 & 4). The first step in the empirical estimation of the  
 147 model is to confirm the stationarity of the variables. To that end, we have applied two unit root tests known as  
 148 the Levin–Lin–Chu (2002) and Im–Pesaran–Shin (2003). After the application of the unit root test and  
 149 confirmation of the order of integration of the variables, we move to our main model.

150 For getting short and long-run estimates of the variables such as REC, GG, Tech, FIE, Education, and  
 151 GS. We have relied on the cross-sectional augmented autoregressive distributive lag order (CS-ARDL) model  
 152 proposed by Chudik and Pesaran (2013). This method has various benefits over other methods. The first-  
 153 generation estimation techniques such as FMOLS, DOLS, etc can't address the issue of cross-sectional  
 154 dependence and provide biased and inefficient estimates (Chen et al., 2022). Whereas, due to the inclusion of  
 155 lagged dependent variable and cross-sectional averages, the CS-ARDL can deal with the most celebrated  
 156 problems of the cross-sectional dependence and also deal with an additional issue of slope heterogeneity; hence,  
 157 providing efficient results (Chudik and Pesaran, 2013). Another benefit of this method is that it can provide both  
 158 short and long-run estimates simultaneously. Further, pre-unit root testing is not mandatory during the  
 159 application of this method because it can also account for the variables that are integrated at different orders.  
 160 Lastly, this method can provide efficient results even in the case of a limited number of observations across  
 161 time.

162 To check the robustness of our results, we have applied PMG-ARDL and the quantile regression  
 163 model. As compared to ordinary least square (OLS), which defines the variation in the dependent variables with  
 164 the averages of the explanatory variables. Nevertheless, in the case of panel data analysis, some of the basic  
 165 assumptions of OLS (zero mean, homoscedasticity, and normal distribution) are violated resulting in biased  
 166 estimates. Therefore, quantile regression analysis is appropriate because in this method dependent variable is the

167 conditional quantile of all independent variables and thereby providing regression for all quantiles. Therefore,  
 168 quantile regression is better as compared to the OLS in the case of panel data due to its ability to analyze the  
 169 impact of explanatory variables on the range of variation and conditional diffusion of the dependent variable.

170

171 **Data**

172 This study explores the impact of technological shocks, financial institutions, and education on green growth  
 173 and renewable energy consumption. Table 1 displays the information regarding abbreviations and definitions of  
 174 variables and descriptive statistics. In this research, green growth (GG) is measured through environmentally  
 175 adjusted multifactor productivity while renewable energy consumption (REC) is measured in terms of the total  
 176 consumption of energy in quad Btu. Technological shocks are measured through two variables. These proxies  
 177 are patent applicants and eco-innovation (EI). Eco-innovation is taken as development of environment-related  
 178 technologies as percentage of all technologies. Financial institutions' role is measured through financial  
 179 institutions efficiency index (FIE). However, education impact is measured by average years of schooling.  
 180 Government spending (GS) is used as a control variable in this study. It is measured as general government final  
 181 consumption expenditures as a percent of GDP. The data for all the variables have been collected from the  
 182 OECD, EIA, Barro-Lee, World Bank and the IMF.

183

184 **Table 1: Definitions and descriptive**

| Variables | Definitions   | Mean  | Median | Maximum | Minimum | Std. Dev. | Skewness | Kurtosis |
|-----------|---|-------|--------|---------|---------|-----------|----------|----------|
| GG        | Environmentally adjusted multifactor productivity                       | 4.376 | 4.142  | 13.14   | -6.233  | 3.752     | -0.045   | 2.780    |
| REC       | total energy consumption from nuclear, renewables, and other (quad Btu) | 5.917 | 3.285  | 19.17   | 0.800   | 5.553     | 1.107    | 2.616    |
| Patent    | Patent applications, total (residents and non-residents)                | 11.56 | 12.05  | 14.24   | 8.139   | 1.503     | -0.290   | 2.013    |
| EI        | Development of environment-related technologies, % all technologies     | 8.583 | 8.470  | 15.71   | 3.360   | 2.370     | 0.200    | 2.327    |
| FIE       | Financial institutions efficiency index                                 | 0.614 | 0.631  | 0.799   | 0.265   | 0.136     | -0.604   | 2.554    |
| Education | Average years of schooling  | 3.531 | 3.893  | 4.487   | 1.032   | 0.945     | -1.012   | 2.868    |
| GS        | General government final consumption expenditure (% of GDP)             | 15.29 | 15.21  | 21.06   | 9.802   | 2.848     | -0.199   | 2.254    |

185

186 **Empirical results**

187 Panel data analysis begins with the investigation of cross-sectional dependence, and for this task, the study  
 188 employed Pesaran's (2004) cross-sectional dependence test. It is expected that the measures adopted by one  
 189 economy influence the nearby economies due to liberalization and globalization. Thus, it is mandatory to test  
 190 the cross-sectional dependence in order to avoid the biased value of cointegration and unit root analysis  
 191 (Westerlund, 2007). The results of cross-sectional dependence are given in Table 2. It is reported that most of

192 the selected variables are dependent cross-sectionally. Any variation in variables like REC, GG, Patent, FIE,  
 193 education, and GS in one of the selected countries will affect other countries in the sample.

194 After confirming the cross-sectional dependence, the next step is to examine the order of integration of  
 195 variables. For this task, the study used LLC and IPS unit root tests. Table 3 reports the findings of both unit root  
 196 tests. The results of LLC test report that GG, patent, education, and GS are level stationary variables, however,  
 197 REC and FIE are integrated at first difference. The findings of IPS unit root test show that GG is stationary at  
 198 level and the remaining variables are stationary at first difference. The next step is to check the cointegration  
 199 among variables. In this regard, Westerlund panel cointegration test is used. Table 4 displays the outcomes of  
 200 panel cointegration test. The panel cointegration test results are collected from the panel group and statistics  
 201 group  $(P_t, P_a)$  and  $(G_t, G_a)$ . The findings display that cointegration exists among all the variables of REC  
 202 model and GG model. It reveals that long-run association exists among the selected variables.

205 **Table 2: Cross-sectional dependence tests**

|                | REC   | GG    | Patent   | FIE     | Education | GS     |
|----------------|-------|-------|----------|---------|-----------|--------|
| Pesaran's test | 0.141 |       | 3.242*** | -1.774* | -0.919    | -1.198 |
| Off-diagonal   | 0.291 |       | 0.287    | 0.272   | 0.336     | 0.252  |
| Pesaran's test |       | 1.455 | -2.390** | -1.892* | -1.821*   | -1.354 |
| Off-diagonal   |       | 0.165 | 0.278    | 0.298   | 0.282     | 0.455  |

206 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

208 **Table 3: Panel unit root tests**

|           | LLC       |           |          | IPS       |           |          |
|-----------|-----------|-----------|----------|-----------|-----------|----------|
|           | I(0)      | I(1)      | Decision | I(0)      | I(1)      | Decision |
| REC       | 2.425     | -7.007*** | I(1)     | 0.947     | -4.747*** | I(1)     |
| GG        | -6.689    |           | I(0)     | -3.994*** |           | I(0)     |
| Patent    | -1.639*   |           | I(0)     | -1.365    | -4.374*** | I(1)     |
| FIE       | 0.441     | -6.300*** | I(1)     | -1.971    | -7.478*** | I(1)     |
| Education | -4.548*** |           | I(0)     | -1.389    | -3.088*** | I(1)     |
| GS        | -3.018*** |           | I(0)     | -1.980    | -3.873*** | I(1)     |

209 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

211 **Table 4: Panel cointegration tests**

| Statistic | REC       |         |         | GG        |         |         |
|-----------|-----------|---------|---------|-----------|---------|---------|
|           | Value     | Z-value | P-value | Value     | Z-value | P-value |
| Gt        | -3.348**  | 2.181   | 0.015   | -4.702*** | 5.129   | 0.000   |
| Ga        | -11.11    | 0.767   | 0.779   | -10.31    | 1.450   | 0.926   |
| Pt        | -8.875*** | 4.168   | 0.000   | -13.27*** | 8.254   | 0.000   |
| Pa        | -15.83**  | 1.770   | 0.038   | -16.28    | 1.208   | 0.113   |

212 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

213

214 After confirming the existence of long-run relationship between variables, the study evaluated the long-  
 215 run and short-run relationship by employing a cross-sectional ARDL approach. The coefficient estimates of the  
 216 renewable energy consumption model are given in Table 5. However, the coefficient estimates of green growth  
 217 models are reported in Table 6. In the renewable energy consumption model, the positive and significant  
 218 coefficient estimate of the patent variable means that an increase in technological innovation brings significant  
 219 increase in renewable energy consumption in the long-run. It implies that a 1 percent upsurge in patent increases  
 220 renewable energy consumption by 0.532 percent in the long-run. Financial institution's efficiency reports a  
 221 significant and positive impact on renewable energy consumption in the long-run. It reveals that 1 percent  
 222 improvement in financial institutions efficiency enhances renewable energy consumption by 1.210 percent in  
 223 the long-run.

224 Education and government spending report statistically insignificant impact on renewable energy  
 225 consumption in the long-run. In the short run, findings display that technology innovation reports no impact on  
 226 renewable energy consumption as shown by a statistically insignificant coefficient estimate of patent variable.  
 227 Financial institution's efficiency and education report statistically significant increasing effect on renewable  
 228 energy consumption in the short-run. In contrast, government spending does not produce any significant effect  
 229 on renewable energy consumption in the short-run. The results of technique-based and variable-based methods  
 230 to measure robustness are consistent in direction and magnitude. In the end, the ECM value is negative and  
 231 significant, which indicates the possibility of convergence towards equilibrium in the long-run. Likewise, in  
 232 Table 7, the estimates of our concerned variables are positively significant in most quantiles, particularly, at  
 233 higher quantiles.

234

235 **Table 5: Short and long-run estimates of renewable energy consumption**

|                  | Basic model |        | Variable based-robustness |        | Method based-robustness |        |
|------------------|-------------|--------|---------------------------|--------|-------------------------|--------|
|                  | CS-ARDL     |        | CS-ARDL                   |        | ARDL-PMG                |        |
|                  | Coefficient | Z-Stat | Coefficient               | Z-Stat | Coefficient             | t-Stat |
| <b>Long-run</b>  |             |        |                           |        |                         |        |
| Patent           | 0.532**     | 2.300  |                           |        | 0.568**                 | 2.589  |
| EI               |             |        | 0.941***                  | 8.960  |                         |        |
| FIE              | 1.210*      | 1.860  | 1.300**                   | 2.140  | 1.703**                 | 2.079  |
| Education        | 0.914       | 0.710  | 0.542                     | 0.190  | 0.979*                  | 1.935  |
| GS               | 0.201       | 0.950  | 0.113***                  | 3.340  | 0.200***                | 2.936  |
| <b>Short-run</b> |             |        |                           |        |                         |        |
| Patent           | 1.532       | 0.860  |                           |        | 1.380                   | 1.338  |
| D(Patent(-1))    |             |        |                           |        | 0.230                   | 0.355  |
| EI               |             |        | 0.525                     | 0.550  |                         |        |
| FIE              | 0.978*      | 1.710  | 1.052                     | 0.500  | 1.161**                 | 2.400  |
| D(FIE(-1))       |             |        |                           |        | 0.689                   | 0.366  |
| Education        | 0.875*      | 1.940  | 0.625                     | 0.310  | 0.238**                 | 2.099  |
| D(Education(-1)) |             |        |                           |        | 0.147                   | 0.255  |
| GS               | 0.142       | 0.460  | 0.107*                    | 1.800  | 0.149                   | 1.938  |
| D(GS(-1))        |             |        |                           |        | 0.366                   | 0.455  |

|         |          |       |          |       |          |       |
|---------|----------|-------|----------|-------|----------|-------|
| C       |          |       |          |       | 1.803    | 1.203 |
| ECM(-1) | -0.365** | 2.365 | -0.398** | 2.012 | -0.570** | 2.165 |

236 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

237

238 In the green growth model, long-run findings display that technological innovation tends to improve  
239 green growth in the long-run as shown by a significant and positive coefficient estimate of patent variable. It  
240 infers that a 1 percent upsurge in patent improves green growth by 1.665 percent in the long-run. Financial  
241 institutions' efficiency seems to have a significant and positive impact on green growth in the long-run. It is  
242 reported as a 1 percent improvement in financial institution's efficiency results in enhancing green growth by  
243 1.754 percent in the long-run. Education is positively and significantly associated with green growth in the long-  
244 run. It seems that a 1 percent rise in education level brings 1.542 percent upsurge in green growth in the long-  
245 run. Government spending also reports significant and positive association with green growth in the long-run. It  
246 shows that a 1 percent expansion in government spending leads to 0.875 percent improvement in green growth  
247 in the long-run. In the short-run, findings report that the role of technology innovation and financial institution's  
248 efficiency is significant and positive in enhancing green growth. In contrast, education and government  
249 spending does not report any significant effect on green growth in the short-run. The coefficient estimates of  
250 technique-based and variable-based robustness models are also consistent as shown by their symbols and  
251 magnitude. The coefficient estimate of ECM term is negative and significant revealing the possibility of  
252 reasonable convergence in the long-run. Lastly, Table 7 revealed that technology innovation and financial  
253 institutions increase green growth in high quantiles.

254

255 **Table 6: Short and long-run estimates of green growth**

|                  | Basic model |        | Variable based-robustness |        | Method based-robustness |        |
|------------------|-------------|--------|---------------------------|--------|-------------------------|--------|
|                  | CS-ARDL     |        | CS-ARDL                   |        | ARDL-PMG                |        |
|                  | Coefficient | Z-Stat | Coefficient               | Z-Stat | Coefficient             | t-Stat |
| <b>Long-run</b>  |             |        |                           |        |                         |        |
| Patent           | 1.665**     | 2.370  |                           |        | 1.176**                 | 2.265  |
| EI               |             |        | 0.732**                   | 8.120  |                         |        |
| FIE              | 1.754*      | 1.790  | 1.564**                   | 2.230  | 1.780**                 | 2.398  |
| Education        | 1.542*      | 1.720  | 1.325*                    | 2.470  | 0.992***                | 2.769  |
| GS               | 0.875*      | 1.860  | 0.785                     | 0.830  | 0.536***                | 2.968  |
| <b>Short-run</b> |             |        |                           |        |                         |        |
| D(Patent)        | 0.665**     | 2.150  |                           |        | 0.497*                  | 1.719  |
| D(Patent(-1))    |             |        |                           |        | 0.762                   | 0.877  |
| EI               |             |        | 0.265***                  | 2.990  |                         |        |
| D(FIE)           | 1.356**     | 2.340  | 1.154*                    | 1.740  | 1.663*                  | 1.762  |
| D(FIE(-1))       |             |        |                           |        | 0.899                   | 0.428  |
| D(Education)     | 0.354       | 0.450  | 0.900                     | 0.500  | 1.526                   | 0.451  |
| D(Education(-1)) |             |        |                           |        | 1.001                   | 1.100  |
| D(GS)            | 0.355       | 0.480  | 0.632                     | 0.890  | 1.120**                 | 2.715  |
| D(GS(-1))        |             |        |                           |        | 1.163                   | 1.541  |
| CC               |             |        |                           |        | 6.110                   | 2.700  |
| ECM(-1)          | -0.504*     | 1.875  | -0.409***                 | 3.025  | -0.644***               | 2.823  |

256 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

257

258 **Results discussions**

259 Our findings suggest that technological shocks help improve green growth and renewable energy  
260 consumption, which is also supported by some previous studies such as Chen et al. (2019) and Danish and  
261 Ulucak (2020). A positive change in technology improves the efficiency of production techniques and  
262 equipment and also helps to transform the energy structure which is crucial for decoupling economic growth and  
263 CO2 emissions. The development of innovative environmental ideas fosters the growth of environment-related  
264 technologies, in particular, and the overall technological process, in general. Similarly, the development of  
265 energy-related technologies improves the energy intensity and increases the share of renewable energy sources  
266 in the total energy mix, thereby increasing renewable energy consumption. Technological innovations are  
267 mandatory for the development of technologies, particularly to improve energy efficiency, energy conservation,  
268 and by substituting non-renewable energy sources with renewable ones. The importance of technological  
269 innovations in the development of renewable energy projects can't be ignored because more modern and  
270 sophisticated technologies can significantly improve energy efficiency and reduce energy consumption (Danish  
271 and Ulucak, 2020). These findings are also supported by the previous works (Silva et al., 2013; Sohag et al.,  
272 2015).

273 As far as the impact of financial institutions efficiency on green growth and renewable energy  
274 consumption are concerned it appears to be positive. According to Frankel and Romer (1999), efficient financial  
275 institutions can provide necessary funds at an affordable cost for research and development activities, necessary  
276 to boost technological innovations, which are significant for the complete separation of economic growth and  
277 CO2 emissions. Similarly, an increase in the efficiency of financial institutions provides funds for the promotion  
278 of pro-environment activities and technologies, which will ultimately be crucial for attaining green growth by  
279 reducing energy-related pollution also highlighted by Claessens and Feijen (2007), Zhang et al. (2021), and  
280 Zhou et al. (2022). Further, the transformation of the energy sector from dirty to clean energy sources require  
281 too much initial cost because the deployment of renewable energy projects is costly. To meet the high initial  
282 cost of renewable energy projects, the efficiency of financial institutions can play a crucial and significant role  
283 (Anton and Nucu, 2020).

284 Human capital is a by-product of education, skill, training, and experience. Therefore, it is used as a  
285 crucial input in the production function and spurs economic growth. Most of the developed economies have  
286 transformed their production process by substituting physical-capital intensive techniques with human-capital  
287 intensive techniques which have turned their production process into environmentally friendly and green  
288 (Mahalick et al. 2021). Further, the rise in the education level makes people more aware and conscious about  
289 their environment and nature; hence, their response towards the environment is more careful. Similarly, a more  
290 educated common individual and firm manager want to consume more clean energy, thus increasing literacy  
291 rate is positively linked to renewable energy consumption.

292 **Table 7: Panel quantile regression**

|                                     | 0.05   | 0.10  | 0.20  | 0.30  | 0.40    | 0.50     | 0.60     | 0.70     | 0.80     | 0.90     | 0.95     |
|-------------------------------------|--------|-------|-------|-------|---------|----------|----------|----------|----------|----------|----------|
| <b>Renewable energy consumption</b> |        |       |       |       |         |          |          |          |          |          |          |
| Patent                              | 0.445* | 0.383 | 0.440 | 0.658 | 1.463** | 3.150*** | 3.199*** | 3.037*** | 2.852*** | 2.594*** | 2.700*** |

|                     |          |         |          |          |          |          |          |          |          |          |          |
|---------------------|----------|---------|----------|----------|----------|----------|----------|----------|----------|----------|----------|
|                     | (1.715)  | (1.181) | (1.046)  | (1.330)  | (2.282)  | (8.198)  | (9.103)  | (9.867)  | (7.715)  | (7.359)  | (6.782)  |
| FIE                 | 1.862*   | 1.347   | 1.561    | 1.195    | 1.540*   | 2.561*** | 2.692*** | 2.209*** | 2.507*** | 1.450**  | 1.752*   |
|                     | (1.805)  | (1.374) | (1.110)  | (1.270)  | (1.862)  | (4.328)  | (4.030)  | (3.653)  | (2.618)  | (2.008)  | (1.768)  |
| Education           | 0.609*   | 0.711*  | 0.597    | 0.430    | -0.187   | -0.741   | -0.038   | 0.384    | 1.597*** | 1.999*** | 1.702*** |
|                     | (1.821)  | (1.774) | (1.175)  | (0.743)  | (0.272)  | (0.820)  | (0.044)  | (0.500)  | (2.966)  | (4.113)  | (3.133)  |
| GS                  | 0.151*   | 0.140   | 0.135    | 0.161    | 0.346*   | 0.736**  | 1.041*** | 1.114*** | 1.397*   | 1.400*** | 1.294*** |
|                     | (1.771)  | (1.376) | (1.044)  | (1.079)  | (1.948)  | (2.530)  | (3.828)  | (4.567)  | (1.825)  | (3.441)  | (11.60)  |
| <b>Green growth</b> |          |         |          |          |          |          |          |          |          |          |          |
| Patent              | 0.981    | 1.255*  | 1.367*** | 1.803*** | 2.130*** | 1.856*** | 0.439    | 0.223    | 0.732*** | 0.774**  | 0.078    |
|                     | (1.211)  | (1.965) | (3.223)  | (4.406)  | (4.950)  | (3.352)  | (0.875)  | (0.588)  | (2.708)  | (2.354)  | (0.126)  |
| FIE                 | 0.977    | -0.365  | -0.128   | -2.298   | -4.277   | -4.034   | 5.061*   | 6.025**  | 6.670**  | 4.090    | 3.601*   |
|                     | (0.186)  | (0.090) | (0.031)  | (0.547)  | (0.988)  | (0.729)  | (1.856)  | (2.467)  | (2.571)  | (1.136)  | (1.775)  |
| Education           | 0.998    | 2.431   | 3.295*   | 4.116*   | 4.201**  | 3.836*** | 1.331    | 0.920    | 1.466**  | 2.206*** | 1.163    |
|                     | (0.490)  | (1.437) | (1.848)  | (1.942)  | (2.532)  | (4.121)  | (1.539)  | (1.360)  | (2.502)  | (4.143)  | (1.369)  |
| GS                  | -        | -       | -0.151   | -0.172   | -0.284*  | -0.093   | -0.044   | 0.028    | 0.241*   | 0.319**  | 0.406*** |
|                     | 0.594*** | 0.326** |          |          |          |          |          |          |          |          |          |
|                     | (3.255)  | (2.398) | (1.088)  | (1.109)  | (1.662)  | (0.444)  | (0.236)  | (0.183)  | (1.766)  | (2.361)  | (2.860)  |

293 **Note:** \*\*\*p<0.01; \*\*p<0.05; \*p<0.1

## 294 **Conclusion and implications**

295 The notion of green growth implies that during the process of economic growth natural resources must  
296 be used with great caution so that extra burden on the environment should be avoided and the environment can  
297 be preserved for future generations. Therefore, the idea of green growth has got paramount importance in the  
298 issue of sustainability. Another important factor that has become popular in achieving the goal of sustainable  
299 economic development in renewable energy consumption. Renewable energy consumption is widely considered  
300 to be the most influential factor in mitigating the effects of CO<sub>2</sub> emissions. Taking into consideration the above  
301 factors, we aim to investigate the determinants of green growth and renewable energy consumption. Among  
302 important determinants of green growth and renewable energy consumption, we have included technological  
303 shocks, financial institutions, and education. To investigate this relationship empirically we have applied the  
304 CS-ARDL model that can take into consideration the issue of cross-sectional dependence. Further, to check the  
305 robustness of our results we have applied PMG-ARDL model and quantile regression model.

306 From the results of the CS-ARDL model, we confer that the estimates attached to the Patent and eco-  
307 innovation are positively significant, confirming that the positive technological shock help to increase renewable  
308 energy consumption. Similarly, the estimates of the Patent and eco-innovation are positively significant in green  
309 growth models, implying that a positive technology shock is beneficial in attaining green growth. On the other  
310 side, the development of financial institutions and an increase in education also intensify renewable energy  
311 consumption and help to achieve green growth. The estimates of the robust ARDL-PMG model also follow the  
312 results of the baseline model and confirm that a positive technology shock, financial institution development,  
313 and increase in education all exert a positive impact on renewable energy consumption and green growth.  
314 Likewise, the estimates of our concerned variables are positively significant in most quantiles, particularly, at  
315 higher quantiles.

316 Generally, our findings imply that a positive shock in technological innovations helps to achieve green  
317 growth and increase renewable energy consumption. These results are important for policy implications in the  
318 issue of environmental sustainability. Therefore, the policymakers should focus on investing in research and

319 development activities which are crucial for increasing the technology-related innovations in the economy. In  
320 order to boost environmental related technologies, the policymakers may entice the people and firms with the  
321 provision of easy credit facilities through financial institutions. Further, the financial and scientific policy should  
322 be integrated with environmental policy to attain the goals of environmental sustainability. Because financial  
323 institutions can provide credits and funds for the deployment of renewable energy plants which require too  
324 much initial investment. Similarly, the research and development activities also require too much investment  
325 that can be fulfilled by strengthening the financial infrastructure. Lastly, the policymakers should also focus on  
326 increasing the formal literacy rate which would be crucial for improving technological innovations and financial  
327 institution development.

328 The literature suggests that the activities related to technological innovation may increase during a  
329 boom and may halt during recessions; however, in this study, we have not addressed this issue empirically.  
330 Therefore, future studies should also take into account the positive as well as negative shocks in technological  
331 innovations, explicitly, and try to examine the nexus between technology, green growth, and renewable energy  
332 consumption by using asymmetric analysis.

333

334

335 **Acknowledgement:** Research on the mechanism and path of mass entrepreneurship and Innovation Incubation  
336 Base promoting employment and entrepreneurship in Colleges and universities in the new era, Grant No.  
337 XJK21CGD035, project No. ND213254

338 **Authors Contributions:** This idea was given by Tan Chao. Tan Chao, Xu Yunbao, and Dai Chengbo analyzed  
339 the data and wrote the complete paper. While Ahmed Usman and Li Bo read and approved the final version.

340 **Availability of data and materials:** The datasets used and/or analyzed during the current study are available  
341 from the corresponding author on reasonable request.

342 **Ethical Approval:** Not applicable

343 **Consent to Participate:** I am free to contact any of the people involved in the research to seek further  
344 clarification and information

345 **Consent to Publish:** Not applicable

346 **Competing interests:** The authors declare that they have no conflict of interest.

347

348

## 349 **References**

350 Álvarez-Herránz, A., Balsalobre, D., Cantos, J. M., & Shahbaz, M. (2017). Energy innovations-GHG emissions  
351 nexus: fresh empirical evidence from OECD countries. *Energy Policy*, *101*, 90-100.

352 Anton, S. G., & Nucu, A. E. A. (2020). The effect of financial development on renewable energy consumption.  
353 A panel data approach. *Renewable Energy*, *147*, 330-338.

354 Chen, J., Rojnruttikul, N., Kun, L. Y., & Ullah, S. (2022). Management of Green Economic Infrastructure and  
355 Environmental Sustainability in One Belt and Road Initiative Economies. *Environmental Science and  
356 Pollution Research*, 1-11.

357 Chudik, A., & Pesaran, M. H. (2013). *Large Panel Data Models with Cross-Sectional Dependence: A*  
358 *Survey* (No. 4371). CESifo.

359 Claessens, S., & Feijen, E. (2007). *Financial sector development and the millennium development goals* (No.  
360 89). World Bank Publications.

361 Frankel, J. A., & Romer, D. H. (1999). Does trade cause growth?. *American economic review*, 89(3), 379-399.

362 Geroski, P. A., & Walters, C. F. (1995). Innovative activity over the business cycle. *The Economic*  
363 *Journal*, 105(431), 916-928.

364 Grossman, G. M., & Krueger, A. B. (1991). *Environmental Impacts of a North American Free Trade*  
365 *Agreement* (No. 158).

366 Grossman, G. M., & Krueger, A. B. (1995). Economic growth and the environment. *The quarterly journal of*  
367 *economics*, 110(2), 353-377.

368 Guo, L., Qu, Y., & Tseng, M. L. (2017). The interaction effects of environmental regulation and technological  
369 innovation on regional green growth performance. *Journal of cleaner production*, 162, 894-902.

370 Hao, L. N., Umar, M., Khan, Z., & Ali, W. (2021). Green growth and low carbon emission in G7 countries: how  
371 critical the network of environmental taxes, renewable energy and human capital is?. *Science of The*  
372 *Total Environment*, 752, 141853.

373 He, L., Zhang, L., Zhong, Z., Wang, D., & Wang, F. (2019). Green credit, renewable energy investment and  
374 green economy development: Empirical analysis based on 150 listed companies of China. *Journal of*  
375 *cleaner production*, 208, 363-372.

376 Hess, D. J., & Collins, B. M. (2018). Climate change and higher education: Assessing factors that affect  
377 curriculum requirements. *Journal of Cleaner Production*, 170, 1451-1458.

378 Horbach, J., Rammer, C., & Rennings, K. (2012). Determinants of eco-innovations by type of environmental  
379 impact—The role of regulatory push/pull, technology push and market pull. *Ecological economics*, 78,  
380 112-122.

381 Li, X., & Ullah, S. (2021). Caring for the environment: how CO2 emissions respond to human capital in BRICS  
382 economies?. *Environmental Science and Pollution Research*, 1-11.

383 Li, X., Ozturk, I., Majeed, M. T., Hafeez, M., & Ullah, S. (2022). Considering the asymmetric effect of financial  
384 deepening on environmental quality in BRICS economies: Policy options for the green  
385 economy. *Journal of Cleaner Production*, 331, 129909.

386 Luukkanen, J., Kaivo-Oja, J., Vähäkari, N., O'Mahony, T., Korkeakoski, M., Panula-Ontto, J., ... & Hogarth, N.  
387 (2019). Green economic development in Lao PDR: A sustainability window analysis of Green Growth  
388 Productivity and the Efficiency Gap. *Journal of cleaner production*, 211, 818-829.

389 Mahalik, M. K., Mallick, H., & Padhan, H. (2021). Do educational levels influence the environmental quality?  
390 The role of renewable and non-renewable energy demand in selected BRICS countries with a new  
391 policy perspective. *Renewable Energy*, 164, 419-432.

392 Mennicken, L., Janz, A., & Roth, S. (2016). The German R&D program for CO2 utilization—innovations for a  
393 green economy. *Environmental Science and Pollution Research*, 23(11), 11386-11392.

394 Mensah, C. N., Long, X., Dauda, L., Boamah, K. B., Salman, M., Appiah-Twum, F., & Tachie, A. K. (2019).  
395 Technological innovation and green growth in the Organization for Economic Cooperation and  
396 Development economies. *Journal of Cleaner Production*, 240, 118204.

397 Popp, D. (2012). The role of technological change in green growth. *Policy Research Working Paper Series*,  
398 (6239).

399 Shuai, S., & Fan, Z. (2020). Modeling the role of environmental regulations in regional green economy  
400 efficiency of China: Empirical evidence from super efficiency DEA-Tobit model. *Journal of*  
401 *environmental management*, 261, 110227.

402 Stegemann, L., & Ossewaarde, M. (2018). A sustainable myth: A neo-Gramscian perspective on the populist  
403 and post-truth tendencies of the European green growth discourse. *Energy research & social*  
404 *science*, 43, 25-32.

405 Ullah, S., Majeed, M. T., & Hafeez, M. (2019). Education, experience, social network and firm survival: the  
406 case of the electrical fittings cluster in Sargodha, Pakistan. *Decision*, 46(3), 267-278.

407 Ullah, S., Ozturk, I., Majeed, M. T., & Ahmad, W. (2021). Do technological innovations have symmetric or  
408 asymmetric effects on environmental quality? Evidence from Pakistan. *Journal of Cleaner*  
409 *Production*, 316, 128239.

410 Wang, K. H., Umar, M., Akram, R., & Caglar, E. (2021). Is technological innovation making world "Greener"?  
411 An evidence from changing growth story of China. *Technological Forecasting and Social*  
412 *Change*, 165, 120516.

413 Wei, L., & Ullah, S. (2022). International tourism, digital infrastructure, and CO2 emissions: fresh evidence  
414 from panel quantile regression approach. *Environmental Science and Pollution Research*, 1-8.

415 Yao, Y., Ivanovski, K., Inekwe, J., & Smyth, R. (2019). Human capital and energy consumption: Evidence from  
416 OECD countries. *Energy Economics*, 84, 104534.

417 Yin, Y., Xiong, X., Ullah, S., & Sohail, S. (2021). Examining the asymmetric socioeconomic determinants of  
418 CO2 emissions in China: challenges and policy implications. *Environmental Science and Pollution*  
419 *Research*, 28(40), 57115-57125.

420 Zhang, D., Mohsin, M., Rasheed, A. K., Chang, Y., & Taghizadeh-Hesary, F. (2021). Public spending and green  
421 economic growth in BRI region: mediating role of green finance. *Energy Policy*, 153, 112256.

422 Zhang, L., Godil, D. I., Bibi, M., Khan, M. K., Sarwat, S., & Anser, M. K. (2021). Caring for the environment:  
423 how human capital, natural resources, and economic growth interact with environmental degradation in  
424 Pakistan? A dynamic ARDL approach. *Science of The Total Environment*, 774, 145553.

425 Zheng, G. W., Siddik, A. B., Masukujjaman, M., & Fatema, N. (2021). Factors Affecting the Sustainability  
426 Performance of Financial Institutions in Bangladesh: The Role of Green  
427 Finance. *Sustainability*, 13(18), 10165.

428 Zhou, G., Zhu, J., & Luo, S. (2022). The impact of fintech innovation on green growth in China: Mediating  
429 effect of green finance. *Ecological Economics*, 193, 107308.

430