

Income Inequality and Environmental Degradation in Egypt: Evidence from Dynamic ARDL Approach

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Income inequality and environmental degradation in Egypt: Evidence from dynamic ARDL approach

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Abstract

This study aims to explore the effect of income inequality on CO₂ emissions in Egypt during the period 1975-2017. " The analysis investigates the validity of the political economy approach compared to the Keynesian approach regarding the inequality-environment nexus. The study applies the novel dynamic autoregressive distributed lags approach (DARDL) to overcome the complications associated with the structure of the ARDL model. The findings showed that the relationship between inequality and CO₂ emissions is not a trade-off relationship. Rather, inequality leads to environmental deterioration in the long term, which supports the political economy approach in explaining the inequality-environment nexus. Hence, the economic development policies adopted in Egypt during the past four decades have led to a negative impact on the environment.

Keywords: CO₂ emissions, cointegration, dynamic ARDL, income inequality, political economy approach.

32 **Introduction**

33 Debates on the relationship between income inequality and environmental degradation have increased in
34 recent years, especially in light of the increasing trends in both carbon dioxide emissions and income
35 inequality in many countries. According to the United Nation Development Program (UNDP, 2019)
36 income inequality has increased in both developed and developing countries during the last four decades.
37 Available statistics on the global environment indicators also indicate that emissions of carbon dioxide
38 have increased worldwide with an annual average of 2.25% during the period 2000-2017 (Ritchie & Roser,
39 2017). These increasing trends in both environmental degradation and income inequality raise several
40 questions about the nexus between them.

41 Despite a large number of recent studies examining the relationship between environmental degradation
42 and income inequality based on different panel data techniques (e.g., Baloch et al., 2020; Chen et al.,
43 2020b; Liu et al., 2019; Uddin et al., 2020a), the studies related to analyzing the effect of inequality on
44 environmental degradation on individual countries are not as numerous as required. Although the results
45 of panel data studies are important in understanding the relationship between these two variables, they
46 failed in reaching a specific global view on the nature of the effect of income inequality on environmental
47 degradation (Hailemariam et al., 2020). The reason for this failure is the great discrepancy in the different
48 panel data results about whether the effect of inequality on environmental degradation is negative or
49 positive. This contradiction in panel data results makes it appropriate in this paper to study the
50 environmental effects of income inequality using the new methods available in the time series
51 econometrics, instead of analyzing a large scale of countries using panel data techniques, so that the results
52 and policy implications drawn from the study will be more specific and clearer.

53 In this context, there are two main approaches, the first is called the political economy approach (Boyce,
54 1994, 2007; Magnani, 2000; Torras & Boyce, 1998), which considers that income inequality deteriorates
55 environmental indicators, so more inequality will be reflected in more environmental degradation.
56 Accordingly, income inequality leads to the existence of corruption that passes and accommodates foreign
57 and local investments in many projects that are harmful to the environment. High inequality supports the
58 economic power of this class that opposes any policies to improve environment indicators, this is because
59 inequality reduces the ability for collective action to reduce environmental degradation. This wealthy class
60 with political influence will not allow laws that limit environmental pollution. Therefore, environmental
61 considerations will not be taken into account in light of the prevalence of an unfair distribution of income
62 and wealth.

63 As for the second approach, it is based on the interpretation of determining the environmental effects of
64 inequality on the Keynesian approach (Heerink et al., 2001; Ravallion et al., 2000), which is concerned
65 with the critical role that aggregate demand exerts on economic activity. This approach believes that
66 inequality in income distribution leads to a transfer of income from the poor to the rich, and thus the
67 demand for environmentally friendly goods and services will increase by the rich, and thus more inequality
68 leads to an improvement in environmental indicators. In explaining this trade-off between inequality and
69 the environment, this approach relies on a high level of the marginal propensity to emit among the poor.
70 If the degree of inequality in income distribution decreases and some income is transferred from the rich

71 to the poor, then an increase in the total consumption of the poor will occur, and this increase will be
72 directed to the consumption of more goods that do not take into account environmental considerations,
73 which puts pressure on resources and ultimately leads to an increase in carbon dioxide emissions. Hence
74 inequality supports the environment according to this approach. Consequently, this study aims to
75 determine which of these two approaches is more capable of explaining the environmental impacts of
76 income inequality in Egypt during the period 1975-2017.

77 The Egyptian economy is a suitable case to study the environmental impacts of income inequality for
78 many reasons. First, Egypt is a developing economy seeking to achieve economic development which
79 affects both environment and the structure of income distribution, and if there is a trade-off between these
80 two variables, then the Egyptian economy will face a difficult choice in its development path. Available
81 statistics on the pattern of income distribution in Egypt indicate that the Gini coefficient, which measures
82 the degree of disparity in income distribution, has increased during the previous decades from an average
83 of 0.37 in the 1970s to 0.39 during the 1980s, then to 0.41 during the 1990s, and finally increased to 0.42
84 during the period 2000-2017 (Solt, 2019). This means that the pattern of economic development that Egypt
85 followed during the previous decades was biased towards the rich.

86 This study seeks to achieve several objectives. First, it aims to identify the appropriate approach to
87 understand the impact of income inequality on environmental degradation in Egypt as an emerging
88 economy. The study employs the dynamic autoregressive distributed lags (DARDL) recently developed
89 by Jordan & Philips (2018) to capture the long and short un relationships between income inequality and
90 environmental degradation in Egypt and avoid the difficulty in explaining the results of the widely known
91 ARDL of (Pesaran et al., 2001) which suffers from a complex structure in the estimation equation that
92 usually includes lags, differences, and difference lags in both short and long terms.

93 Second, the study covers an important period in the path of economic development in Egypt that has
94 begun in 1975, which is the year after the Egyptian government announced a fundamental structural
95 change in its economic orientations from an economy depending on government intervention to an
96 economy facilitated by supply and demand mechanisms and integrated with the world economy
97 (Farzanegan et al., 2020). This new economic orientation has resulted in a change in the pattern of income
98 distribution in the Egyptian economy during the four decades following this year. In this context, it is
99 necessary to determine the effect of the new pattern of income distribution on environmental degradation
100 in Egypt during this relatively long period.

101 Thirdly, in Egypt, not only a deterioration in income distribution has occurred, but an environmental
102 deterioration has also occurred during the past four decades. Carbon dioxide emissions have increased at
103 an annual average of 2.13% during the period 1975-2017 (BP, 2020). Many of the explanations tried to
104 find the main drivers of carbon dioxide emissions in many developed and developing countries, but there
105 is a severe dearth of empirical works on the environmental effects of inequality that have been conducted
106 for the Egyptian economy.

107 The remainder of the study is structured as follows. the second section reviews the literature related to
108 the impact of inequality on the environment, while the third section describes the data and exposes the

109 methodology of the dynamic ARDL model versus the ARDL model, whereas the fourth section discusses
110 the estimation results. Section 5 concludes

111 **2. literature review**

112 Over the past decades, academic attention began to study the determinants of environmental degradation.
113 Empirical works are devoted, in the early stage of research, to estimate the effects of economic growth on
114 the environment, and many studies conducted on developed and developing countries provided evidence
115 on the existence of the inverted U-curve hypothesis in describing the relationship between economic
116 growth and environmental degradation (Grossman & Krueger, 1995). Indeed, the Environmental Kuznets
117 Curve (EKC) implies a nonlinear relationship between the two variables in which the environment begins
118 to deteriorate in the early stages of economic growth and then improves after reaching a specific real
119 income threshold.

120 Despite the importance to investigate the effects of income growth to understand environmental
121 degradation, researchers realized the necessity to monitor the impact of the structure of income distribution
122 by considering the environmental effect of income inequality indicators. In this respect, several
123 approaches have been used to determine the effect of income inequality on environmental degradation.
124 One of them is to capture its effects indirectly by estimating the effects that inequality will have on factors
125 that positively or negatively affect environmental degradation. (Uzar, 2020) presented his analysis of the
126 impact of inequality on one of the main factors that reduce environmental deterioration, renewable energy
127 consumption, in 43 developed and developing countries during the period 2000-2015. He concluded that
128 reducing income inequality will increase the consumption of renewable energy. This result can give us
129 evidence that inequality harms the environment because of its negative effect on renewable energy
130 consumption.

131 Researchers tried to estimate the environmental impacts of income inequality directly by adding one of
132 its indicators into the estimation model. As usual, there are many contradictory results about this effect.
133 Some believe that inequality improves the environment, others believe that inequality deepens
134 environmental degradation. In this context, one can distinguish between two main trends that explain the
135 impact of income inequality on the environment. The first trend is called the Keynesian approach
136 (Ravallion et al., 2000) and the second is called the political economy approach (Boyce, 1994, 2007;
137 Torras & Boyce, 1998). According to the Keynesian approach, more inequality enhances the environment.
138 Differences in marginal propensity to consume between poor and rich lead the rich to consume more in
139 friendly environmental goods and services if their incomes increase at the expense of the poor, so
140 inequality improves the environment. On the other hand, if redistribution policies transfer incomes from
141 the rich to the poor to reduce income inequality, the poor will increase their demand by consuming more
142 unfriendly environmental goods and services and thus inducing environmental degradation. This trade-off
143 relationship between income inequality and the environment has been supported by many empirical
144 studies in developed and developing countries. (Hailemariam et al., 2020) supported this hypothesis by
145 estimating the effects of income inequality on carbon dioxide emissions in OECD countries using the
146 common correlated effects mean group CCEMG to consider the problem of cross-section dependency,
147 they found that an increase in income inequality measured by the Gini coefficient reduced the CO₂

148 emissions. Other methodologies are employed to support the Keynesian or trade-off hypothesis, for
149 example (Hübler, 2017) employed the quantile regression technique instead of the mean regression
150 techniques to estimate the effects of income inequality on different quantiles of carbon emissions. He
151 found that increasing inequality could reduce CO₂ emissions.

152 The second main trend in studying the effects of income inequality on the environment is what is called
153 the political economy approach. According to (Boyce, 2007) high wealth inequalities harm the
154 environment. This view explains the effect of inequality on the environment through the political
155 dimension of inequality which results in the concentration of wealth. More income inequality creates a
156 small class that can deepen its political positions based on its economic power and becomes effective in
157 the decision-making process. Therefore, the concerns of this group, will not concern with projects that
158 support the environment as much as it will focus on influencing investment towards environmentally
159 unfriendly investments. According to (Boyce, 2007), if the parties benefiting from environmental
160 pollution have the economic power backed by political influence, then the parties affected by the negative
161 environmental impacts will not be able to bring about legislative changes that protect the environment.
162 Then the effect of inequality harms the environment as long as this wealthy class exists and influences
163 investment decisions. So, as noted in recent research, more equality in income distribution could reduce
164 CO₂ emissions (Zhang & Zhao, 2014). The political economy explanation of inequality-emissions nexus
165 has been supported also by (Knight et al., 2017), they found that increasing the wealth share of the top
166 10% had a positive effect on the per capita CO₂ emissions. According to this study, the concentration of
167 wealth leads to political influence and economic power preventing any positive actions to protect the
168 environment. The same results are driven by (Uzar & Eyuboglu, 2019) by employing the ARDL
169 methodology to investigate the environmental effects of income inequality in Turkey during 1984-2014.
170 Their results show that inequality harms the environment which makes it one of the main factors affecting
171 the environment policy design that aims to reduce emissions.

172 **3. Data and methodology**

173 **3.1. Data**

174 The present study examines the effect of income inequality on environmental degradation using carbon
175 dioxide emissions (co₂) as a proxy for the state of the environment in Egypt during the period (1975-
176 2017). Despite the availability of data for many independent variables in this study after 2017, the time
177 range of the study stopped in 2017 due to the absence of published data on both co₂ emissions and Gini
178 coefficient after that year.

179 The independent variables include the Gini index (gini) to measure income inequality as used by many
180 studies, real gross domestic product per capita (gdppc) to measure economic growth, people living in
181 urban areas as a share of total population (urb) to capture the urbanization effect, the sum of exports and
182 imports of goods and services measured as a share of gross domestic product to express trade openness,
183 and primary energy consumption per capita (gigajoule per capita) to track energy consumption effect on
184 co₂. All variables will be used in logarithms to overcome the heteroscedasticity problem (Khan, et al.,
185 2019). Tables (1) and (2), and Figure (1) describe the variables used in the study and trace the time path
186 evolution for the series during the period (1975-2017).

187

Table 1. variables definition and data sources

Variables	description	source
co2	Carbon dioxide emissions (kg per capita)	(British Petroleum, 2020)
gini	The Gini index of disposable income inequality, its value ranges from 0 to 1, with 0 means perfect equality and 1 refers to a perfect inequality	(Solt, 2020)
gdppc	Gross domestic product divided by midyear population. Data are in constant 2010 U.S. dollars.	(World Bank, 2020)
urb	Urban population: people living in urban areas as a share of the total population.	(World Bank, 2020)
trade	Trade openness is the sum of exports and imports of goods and services measured as a share of gross domestic product.	(World Bank,2020)
ecpc	Energy consumption per capita (Gigajoule per capita)	(British Petroleum, 2020)

188

189 There are many statistical properties of the data presented in Table 2. Perhaps the most noticeable is the
190 coefficient of variance, which measures the ratio of the standard deviation to the mean. Real GDP per
191 capita has achieved the highest percentage of variation compared to the rest of the variables, reaching
192 (32.5%), which means that there is a large variation or dispersion in incomes among individuals, and this
193 is supported by the rise in the mean of Gini coefficient (0.4). On the other hand, the dependent variable in
194 this study, co2, also showed a great variation that reached 24.8% during the study period. This variation
195 in carbon dioxide emissions and the rest of the variables may justify the need to study the relationship
196 between these variables in the short run and long run.

197

Table 2. Descriptive Statistics

Variables ^a	Obs.	Mean	Std. Dev.	Std. Dev./mean (%)	Min	Max	Skew.	Kurt.
co ₂	43	455.084	112.832	24.793	267.192	625.962	.213	1.667
gini	43	.405	.021	5.185	.365	.438	-.3	2.063
gdppc	43	1819.222	591.436	32.51	802.749	2817.32	.153	1.871
urb	43	.432	.005	1.157	.427	.44	.426	1.577
trade	43	.508	.112	22.04	.302	.745	.354	2.376
ecpc	43	27.957	8.256	29.53	11.493	40.541	-.074	2.093

^a Statistics are calculated using numbers before taking logarithms.

198

199

Table 3. Matrix of correlations

Variables	lco2	lgini	lgdppc	lurb	ltrade	lecpc
lco2	1.000					
lgini	0.932	0.932				
lgdppc	0.978	0.978	0.978			
lurb	-0.744	-0.744	-0.744	-0.744		
ltrade	-0.259	-0.259	-0.259	-0.259	-0.259	
lecpc	0.961	0.961	0.961	0.961	0.961	0.961

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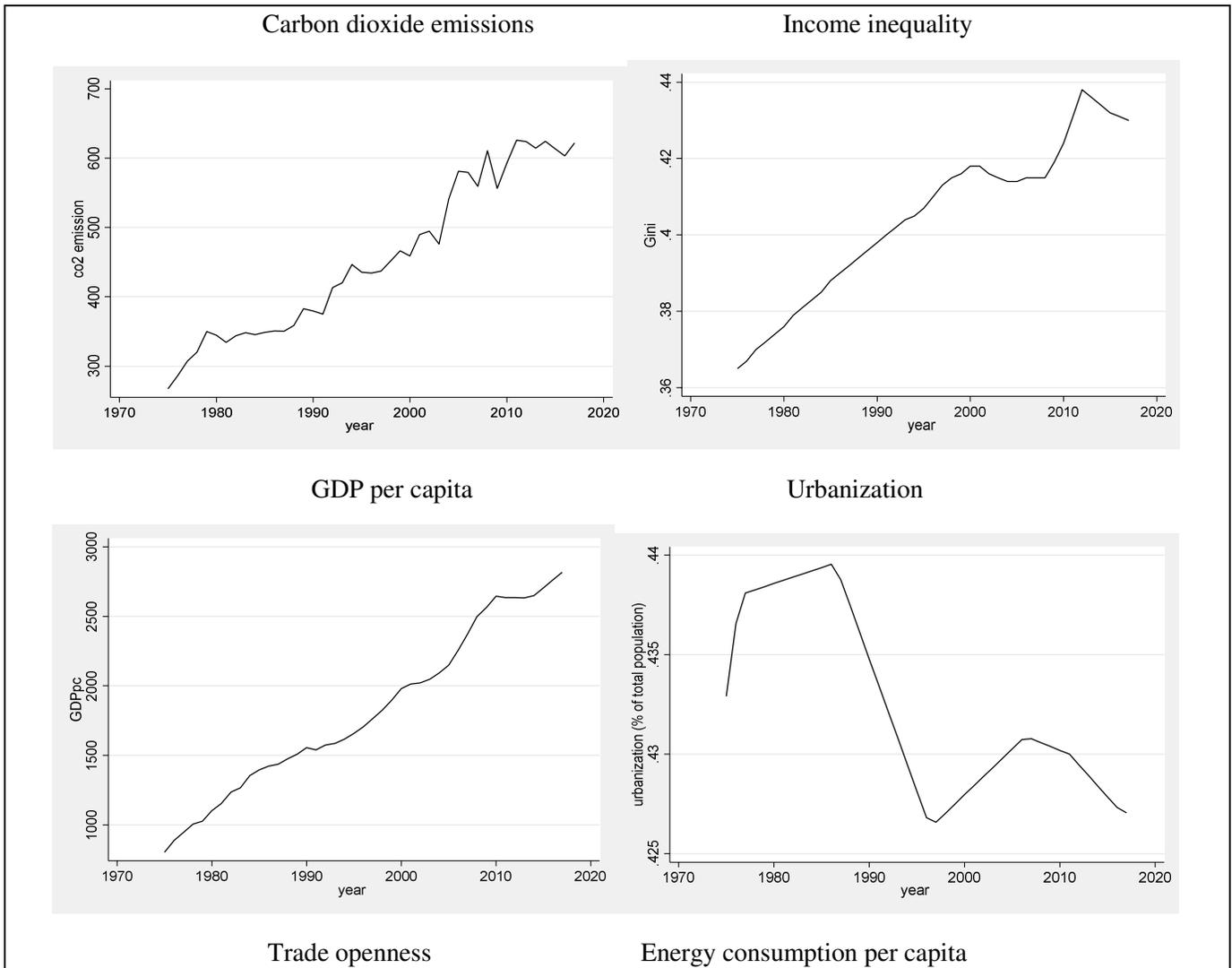
202

Table (3) indicates that there is a strong correlation between all the independent variables and the dependent variable except for trade openness, where these coefficients are more than 90% in all variables,

203 and more than 74% for urbanization. So, it is necessary to place these independent variables in the
204 estimation model of the study. It is noted from Figure (1) that there is an upward trend in carbon dioxide
205 emissions, economic growth, and energy consumption, while the openness index is characterized by
206 volatility and the urbanization index took a downward trend with a relative fluctuation in Egypt during
207 the period from 1975 until 2017.

208

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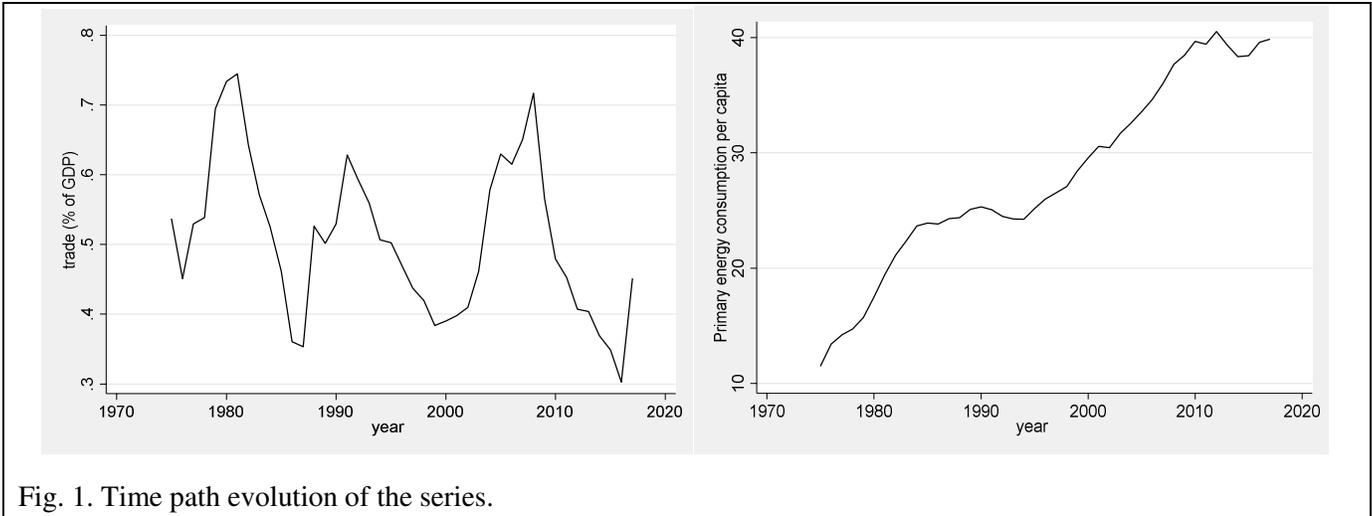


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Fig. 1. Time path evolution of the series.

216

217 **3.2. Model specification: Dynamic ARDL versus ARDL**

218 The main objective of the study is to find out the effect of income inequality on the environmental
 219 pollution in Egypt proxied by CO₂ emission, and then the Gini coefficient represents the main independent
 220 variable in the study. To avoid model specification bias, the study adds control variables depending on the
 221 results of many studies in the relevant literature concerning CO₂ drivers. In this context, economic growth
 222 is often considered an important variable that affects carbon dioxide emissions (e.g., Munir et al., 2020).
 223 Its effect is usually studied within the environmental Kuznets curve hypothesis, which indicates that
 224 economic growth deteriorates the environment in the early stages of development and then turns into
 225 enhancing the environment in the advanced levels of economic development.

226 The study also includes urbanization index as a control variable as it is recently noted in the literature
 227 of CO₂ drivers. Several new studies have begun to be concerned with the impact of urbanization on the
 228 environment as one of the main sources affecting carbon dioxide emissions (Dong et al., 2019). This
 229 concern of the environmental effect of urbanization in the empirical literature prompted us to include it as
 230 an independent variable in the study, especially since many of the results captured a significant effect of
 231 urbanization on carbon dioxide emissions (Ali et al., 2019). On the other hand, urbanization is an important
 232 indicator to know the development that is taking place at the level of structural change, especially in a
 233 developing country seeking to achieve economic development like Egypt.

234 Because the Egyptian economy applied many economic policies during the study period (1975-2017),
 235 which are consistent with the nature of economic policy programs of both the International Monetary Fund
 236 and the World Bank, and which were aimed largely at liberalizing foreign trade, it was appropriate to
 237 study the impact of these policies on carbon dioxide emissions by adding an index of trade openness,
 238 which measures the ratio of total exports and imports to GDP, to analyze the environmental effect of trade
 239 openness in Egypt. Tracking the impact of trade openness on the environment is a common practice in
 240 much empirical research such as (Leal & Marques, 202).

241 Despite the importance of energy consumption in influencing economic growth, the effect of primary
 242 energy use on carbon dioxide emissions is difficult to ignore when formulating the econometric model in

243 this study, especially in a developing economy such as the Egyptian economy. Accordingly, the
 244 econometric model can be formulated in a logarithmic form as follows:

$$245 \quad lco_{2t} = \beta_0 + \beta_1 lgin_i_t + \beta_2 lgdppc_t + \beta_3 lurb_t + \beta_4 ltrade_t + \beta_5 lecp_c_t + e_t \quad (1)$$

246 in which $\beta_1, \beta_2, \beta_3, \beta_4,$ and β_5 account for the elasticity of carbon dioxide emissions lco_2 in response to a
 247 1% change in income inequality ($lgini$), GDP per capita ($lgdppc$), urbanization ($lurb$), trade openness
 248 ($ltrade$), and energy consumption per capita ($lecp_c$) respectively. $\beta_0,$ and e_t denote constant and the error
 249 term respectively, t refers to the year and l means logarithm. All variables are defined in Table 1.

250 The main feature of the ARDL methodology according to (Pesaran et al., 2001) is the possibility of
 251 testing cointegration among time series that differ in the degree of integration so that it does not exceed
 252 I(1) and provided that the dependent variable is integrated after taking the first difference I(1). Thus, as
 253 long as there is no time series is integrated in the second-order I(2), it is possible to test the relationship
 254 between the variables not only in the long run but rather to monitor the behavior of these variables in the
 255 short run and to capture the adjustment process by estimating the error correction term.

256 The ARDL model can be formulated as follows:

$$257 \quad \Delta lco_{2t} = a_0 + \sum_{i=1}^p a_{1i} \Delta lco_{2t-i} + \sum_{i=0}^{q1} a_{2i} \Delta lgin_i_{t-i} + \sum_{i=0}^{q3} a_{3i} \Delta lgdppc_{t-i} + \sum_{i=0}^{q4} a_{4i} \Delta lurb_{t-i}$$

$$258 \quad + \sum_{i=0}^{q5} a_{5i} \Delta ltrade_{t-i} + \sum_{i=0}^{q6} a_{6i} \Delta lecp_c_{t-i} + b_1 lnco_{2t-1} + b_2 lgin_i_{t-1} + b_3 lgdppc_{t-1}$$

$$259 \quad + b_4 lurb_{t-1} + b_5 ltrade_{t-1} + b_6 lecp_c_{t-1} + \epsilon_t \quad (2)$$

260 where Δ refers to the first difference, p and qi denote the lag length determined by Akaike information
 261 criterion (AIC), and ϵ_t represents the error term. The long-run relationships will be captured by b , whereas
 262 the short-run relationships will be explored by a . Long-run equilibrium relationship among variables is
 263 testing using F-statistic based on the null hypothesis of no cointegration $H_0: a_1 = a_2 = a_3 = a_4 = a_5 =$
 264 $a_6 = 0$ and against the alternative hypothesis of the existence of cointegration of $H_1: a_1 \neq a_2 \neq a_3 \neq$
 265 $a_4 \neq a_5 \neq a_6 \neq 0$. According to (Pesaran et al., 2001), if the value of the F-statistic is greater than the
 266 upper bound of the critical values, then there is a long-run equilibrium relationship among the variables,
 267 and if the calculated value of the F-statistic is less than the lower bound then there is a long-run equilibrium
 268 relationship among the variables. This study depends on the critical bounds values provided by (Kripfganz
 269 & Schneider, 2019)

270 The error correction model can be estimated according to Eq. (3) whereas ECT_{t-1} represent a one lagged
 271 error correction term and δ is the adjustment coefficient that must be statistically significant and negative
 272 to guarantee the convergence towards the equilibrium path in the long run.

$$\begin{aligned}
273 \quad \Delta lco_{2t} = a_0 + \sum_{i=1}^p a_{1i} \Delta lco_{2t-i} + \sum_{i=0}^{q1} a_{2i} \Delta lgini_{t-i} + \sum_{i=0}^{q3} a_{3i} \Delta lgdppc_{t-i} + \sum_{i=0}^{q4} a_{4i} \Delta lurbt_{t-i} \\
274 \quad + \sum_{i=0}^{q5} a_{5i} \Delta ltrade_{t-i} + \sum_{i=0}^{q6} a_{6i} \Delta lecpc_{t-i} + \delta ECT_{t-1} + \epsilon_t \quad (3)
\end{aligned}$$

275 Despite the simple implementation and the wide use of ARDL models in studying the relationship
276 between variables in the short and long runs, there are many complications in understanding and
277 interpreting the results of these models as they usually contain multiple lags or difference lags. The
278 complex structure of ARDL models leads to difficulty in quantifying the short and long-run effects of the
279 independent variables on the dependent variable which may lead to a lack of clarity for the decision-maker
280 to apply the appropriate economic policy. In addition to the importance of knowing the extent of a
281 relationship between the variables, as well as the direction of this relationship, the policy-maker also needs
282 a clear answer to the following question: What is the effect of an increase or decrease in one of the
283 independent variables in the ARDL model on the response variable while the other variables remain
284 constant? To overcome these difficulties, (Jordan & Philips, 2018) presented a Dynamic Autoregressive
285 Distributed Lags model using simulation to predict the effect of a positive or negative change in one of
286 the regressors on the dependent variable while keeping the rest of the variables constant (Khan, Teng,
287 Khan, et al., 2019).

288 According to (Danish & Ulucak, 2020; Jordan & Philips, 2018) the error correction term estimated by
289 dynamic ARDL can be calculated according to Eq. (4):

$$\begin{aligned}
290 \quad \Delta lco_{2t} = a_0 + \omega_0 lco_{2t-1} + \varphi_1 \Delta lgini_t + \omega_1 lgini_{t-1} + \varphi_2 \Delta lgdppc_t + \omega_2 lgdppc_{t-1} + \varphi_3 \Delta lurbt_t \\
291 \quad + \omega_3 lurbt_{t-1} + \varphi_4 \Delta ltrade_t + \omega_4 ltrade_{t-1} + \varphi_5 \Delta lecpc_t + \omega_5 lecpc_{t-1} + \epsilon_t \quad (4)
\end{aligned}$$

292
293 By generating 5000 simulations, the dynamic ARDL not only can estimate coefficients in the short and
294 long runs but also predict, using graphs, the response of carbon dioxide emissions to positive and negative
295 shocks in every independent variable while other regressors are constant (Khan et al., 2020).
296

297 **4. Results and discussion**

298 **4.1. Unit root tests**

299 Many criticisms have been directed to the results of unit root tests when ignoring the presence of structural
300 breaks (Perron, 1989). During the study period 1975-2017, fundamental changes have been occurred in
301 macroeconomic policies based on different IMF programs implemented in the Egyptian economy. These
302 programs included many structural adjustment policies like privatization and exchange rate liberalization.
303 So, it is necessary to test the stationarity of time series considering the presence of structural breaks. The
304 study applies Clemente–Montañés–Reyes unit root test with two endogenous structural breaks (Clemente
305 et al., 1998) to capture the sudden changes in means of the series which called the additive outliers model
306 (AO), along with exploring the gradual changes which called innovative outliers (IO). According to the

307 results of the Clemente–Montañés–Reyes unit root test listed in Table (4) all series are integrated of order
 308 one I(1), meaning that the dynamic ARDL cointegration test could be performed.

Table 4. Clemente–Montañés–Reyes unit root test with two structural breaks

Variables	Additive outlier model (AO)			Innovative outlier model (IO)		
	t- statistic	Break 1	Break 2	t-statistic	Break 1	Break 2
level						
<i>Lco2</i>	-4.993	1989	2001	-5.218	1990	2002
<i>lgini</i>	-2.651	1992	2013	-2.342	2007	2010
<i>lgdppc</i>	-3.225	1985	2001	-3.147	1994	2004
<i>lurb</i>	-3.433	1998	2015	-5.467	1985	2011
<i>ltrade</i>	-3.659	1987	2013	-3.485	1988	2010
<i>lecpc</i>	-3.737	1983	2002	-4.570	1977	1997
First difference						
<i>d.lco2</i>	-6.918**	2003	2007	-8.808**	2002	2006
<i>d.lgini</i>	-5.618**	2009	2014	-6.107**	1999	2000
<i>d.lgdppc</i>	-5.977**	1982	2006	-5.574**	1984	1993
<i>d.lurb</i>	-5.564**	1989	1999	-6.698**	1985	1995
<i>d.ltrade</i>	-5.500**	2002	2006	-5.516**	1985	2007
<i>d.ecpc</i>	-5.759**	1982	2010	-6.010**	1983	1993

Notes: The critical value at 5% significance level is - 5.490

309

310 **4.2. Bounds cointegration test**

311 According to the results of unit root tests tabulated in Table (4), the bounds test for cointegration is
 312 performed to explore the eventual existence of long run relationships among variables. Table (5) includes
 313 F-statistic and T-statistic values compared with critical values of (Kripfganz & Schneider, 2019) and
 314 approximate p-values. The results indicate that the null hypothesis of no cointegration can be rejected as
 315 both F and T calculated values are greater than the critical values of the upper bounds I(1) at 10% and 5%
 316 levels of significance respectively. This result confirms the existence of a long-run equilibrium
 317 relationship between CO2, income inequality, economic growth, urbanization, trade openness, and
 318 primary energy consumption in Egypt during the period (1975-2017).

319

320 **Table 5. Bounds test for cointegration**

Calculated values		Kripfganz and Schneider (2019) critical values and approximate p-values							
		10%		5%		1%		p-value	
		I(0)	I(1)	I(0)	I(1)	I(0)	I(1)	I(0)	I(1)
F- statistic	4.082*	2.451	3.760	2.957	4.440	4.167	6.052	0.011	0.072
T- statistic	-4.554**	-2.504	-3.812	-2.864	-4.240	-3.597	-5.110	0.001	0.029

Notes: *** p<0.01, ** p<0.05, and * p<0.1

321

322 **4.3. Estimation results**

323 Table (6) displays the estimation results of both dynamic simulated ARDL model estimation and the
 324 ARDL model used as a robustness check model, whereas Figures (2-6) depict the effects of 1 % decrease

325 and 1% increase in predicted income inequality, economic growth, urbanization, trade openness, and
 326 energy consumption on CO2 emissions in Egypt respectively. Results indicate that the value of the error
 327 correction term ECT_{t-1} is negative and statistically significant. This means that, in the short run, the model
 328 corrects its deviation (error) and converges to the long-run equilibrium path at a speed of 75% annually.

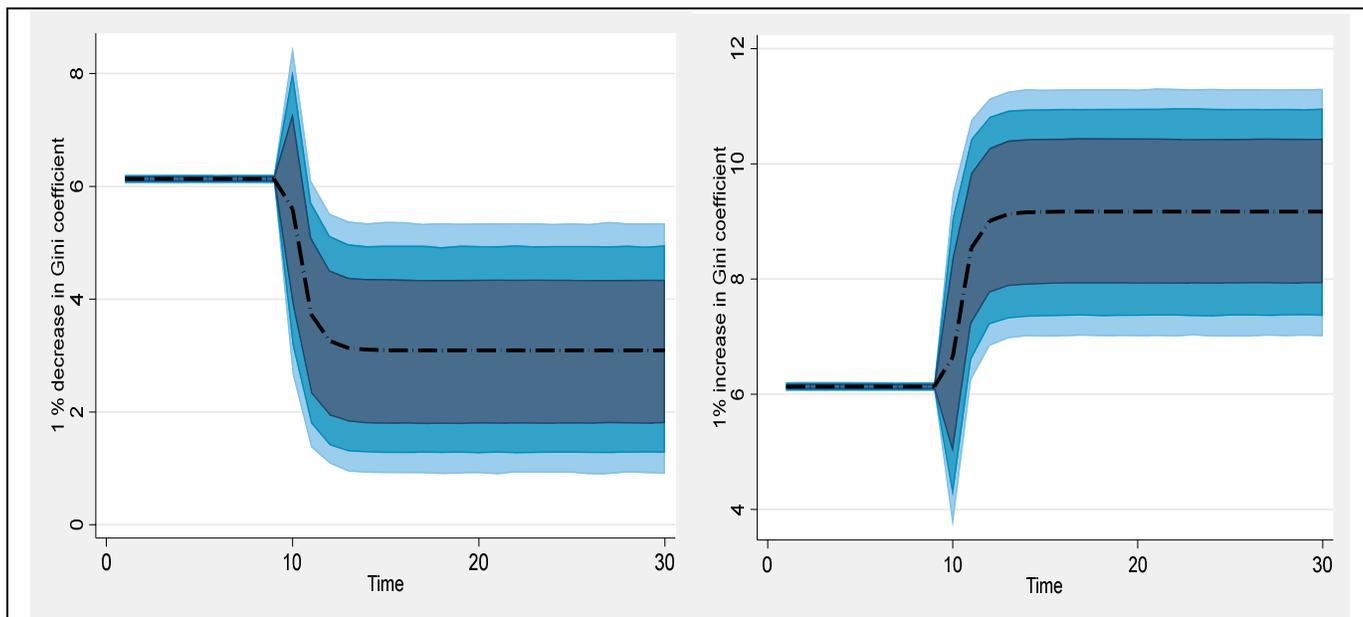
Table 6. Dynamic ARDL and ARDL results

VARIABLES	Dynamic ARDL	ARDL
ECT_{t-1}	-0.750*** (0.165)	-0.663*** (0.140)
$\Delta l g i n i$	0.526 (1.487)	
$l g i n i_{t-1}$	2.285* (1.216)	3.102** (1.442)
$\Delta l g d p p c$	0.538 (0.469)	
$\Delta l u r b$	11.61** (4.358)	8.300* (4.207)
$\Delta l t r a d e$	0.139*** (0.0487)	
$\Delta l e c p c$	-0.536* (0.309)	
$l g d p p c_{t-1}$	1.039*** (0.293)	1.347*** (0.342)
$l u r b_{t-1}$	3.990* (2.144)	4.902* (2.716)
$l t r a d e_{t-1}$	0.188*** (0.0513)	0.232*** (0.0572)
$l e c p e_{t-1}$	-0.761*** (0.229)	-0.987*** (0.339)
Constant	4.893 (3.250)	4.239 (2.660)
R ²	0.527	0.454
Simulations	5000	
Diagnostic Tests^b		
Jarque-Bera	.3392[0.844]	1.414 [0.493]
Breusch-Pagan / Cook-Weisberg test for heteroskedasticity	0.06[0.807]	0.02[0.90]
Breusch-Godfrey LM test for autocorrelation	2.052[0.321]	0.218[0.6405]
Ramsey RESET test	0.24[0.8671]	1.79[0.1698]
Cumulative sum test for parameter stability ^a	0.567	0.518
Durbin-Watson d-statistic	2.298	2.073

329 Notes: Standard errors in parentheses *** p<0.01, ** p<0.05, and * p<0.1. ^a Critical value are: 1.143, 0.9479, and 0.850 for
 330 1%,5%, and 10% levels of significance respectively. ^b Square brackets include p-value.

331
 332 Several diagnostic tests are performed to confirm the power and reliability of the dynamic ARDL and
 333 ARDL assessment results. According to the Jarque-Bera test residuals are normally distributed, and the
 334 models don't suffer from the serial correlation according to the Breusch-Godfrey LM test for
 335 autocorrelation, and also there is no problem of heteroskedasticity as indicated by the results of the
 336 Breusch-Pagan /Cook-Weisberg test for heteroskedasticity. On the other hand, the diagnostic tests indicate
 337 that models are stable and don't suffer from specification bias according to the cumulative sum test for
 338 parameter stability and Ramsey RESET tests respectively.

339 Results indicate that in the short run there is no significant relationship between income inequality and
 340 carbon dioxide emissions, whereas, there is a statistically significant and positive relationship between
 341 them in the long term, as a 1% rise in income inequality increases carbon dioxide emissions in the long
 342 run by 2.285%. This effect of income inequality in Egypt in long run is consistent with other results in
 343 recent works like (Baloch et al., 2020; Wolde-Rufael & Idowu, 2017) and contradicts with others
 344 (Grunewald et al., 2017).

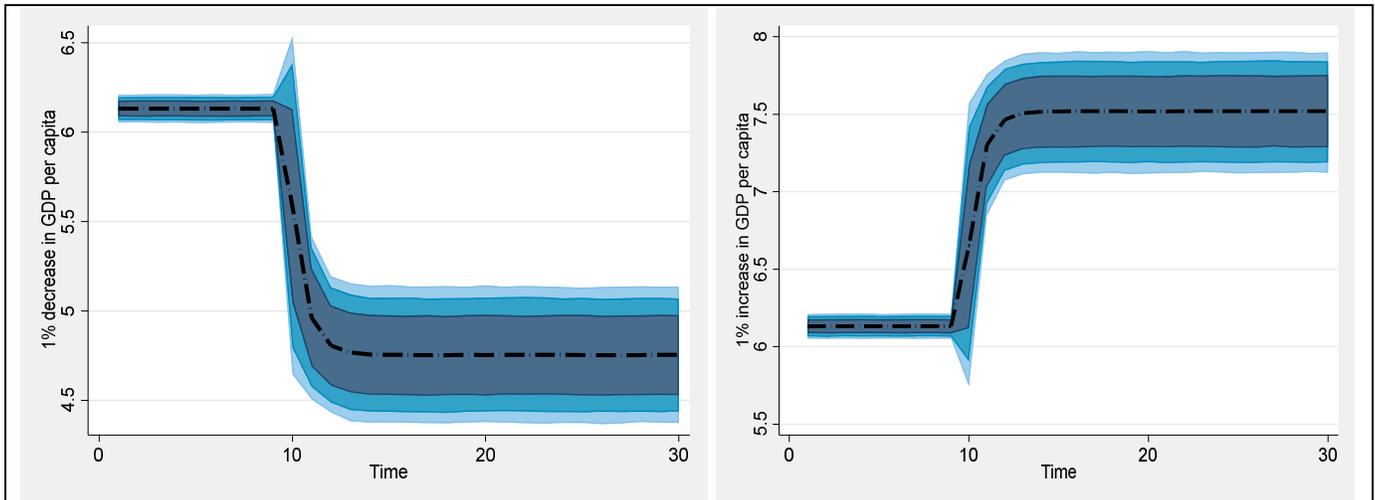


345 Fig. 2. Effects of 1 % decrease and a 1% increase in predicted Gini coefficient on CO_2 in Egypt. The average predicted values
 346 are indicated by the dashed line. The shaded area shows (from darkest to lightest) 75%, 90%, and 95% confidence intervals.
 347

348

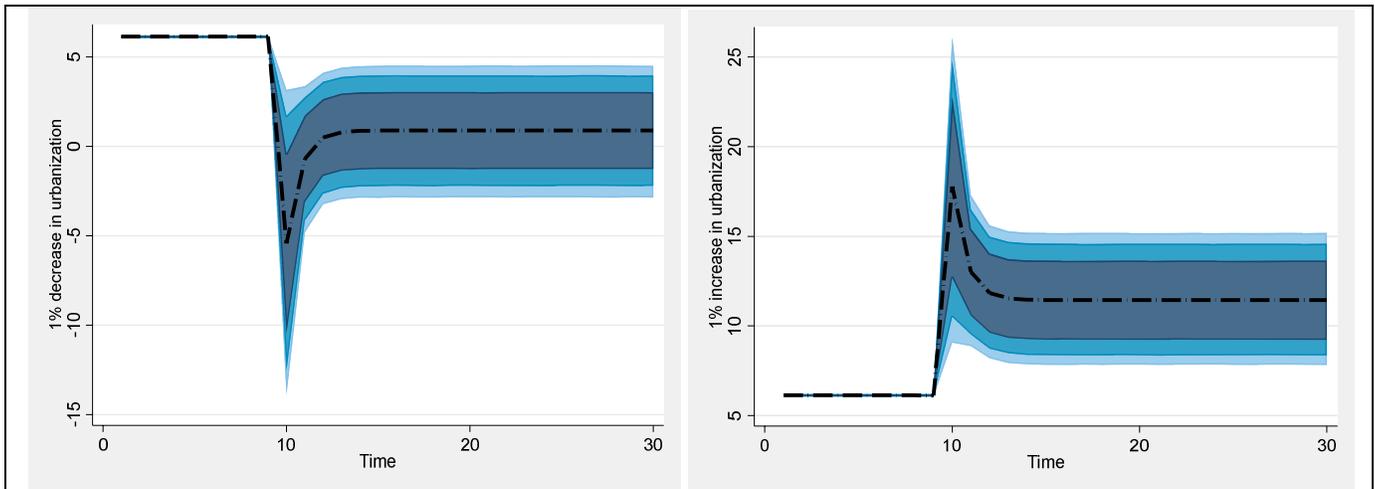
349 This finding is inconsistent with the inequality-environment trade-off approach which depends on the
 350 high marginal propensity to emit for the poor. Despite the high ratio of private consumption to GDP in
 351 Egypt that reached 88% in 2017, which means that the marginal propensity to consume in Egypt is very
 352 high and accordingly the expected effect of income inequality will enhance the environment, results
 353 suggest that more inequality increases CO_2 emission in Egypt. The harmful effect of income inequality on
 354 the environment in the long term means that the political economy approach to understanding the effect
 355 of inequality on environmental pollution is valid in the Egyptian economy, and deepening inequality
 356 through transferring incomes to the rich at the expense of the poor in Egypt will not lead to an increase in
 357 the rich's demand for environmentally friendly goods, but it will lead to an increase in the demand of the
 358 poor to purchase less environmentally friendly and cheaper goods, which harms the environment. On the
 359 other hand, the continued concentration of income and wealth in the hands of the few people or rich class
 360 as a result of the continuing increase in income inequality, as what happened in the increase of the Gini
 361 coefficient in Egypt during the past four decades. In the long term, this class would create political centers
 362 based on its economic power that would enable them to block any laws taking environmental
 363 considerations into account, whether at the level of domestic production or imports from abroad. Although
 364 economic growth doesn't affect environmental pollution in the short term, its effect, in the long run, is

365 positive and significant. Results show that an increase in the real per capita GDP by 1% will lead to an
 366 increase in carbon dioxide emission rates in Egypt by 1.04% in the long run. This can be explained by
 367 relying on the fact that the real GDP per capita in Egypt is still low ranging from 802.7 to 2817.3 according to Table
 368 (2) during the study period (1975-2017), so any increase in real income will be directed to more consumption at the
 369 expense of environmental degradation. Results refer to a very harmful environmental effect of urbanization
 370 in the short and long terms in Egypt as supported by recent works on urbanization-emissions nexus (Das
 371 Neves Lopes et al., 2020; Ullah et al., 2020).



372
 373 Fig. 3. Effects of 1 % decrease and a 1% increase in predicted GDP per capita on CO_2 in Egypt. The average predicted values are
 374 indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.

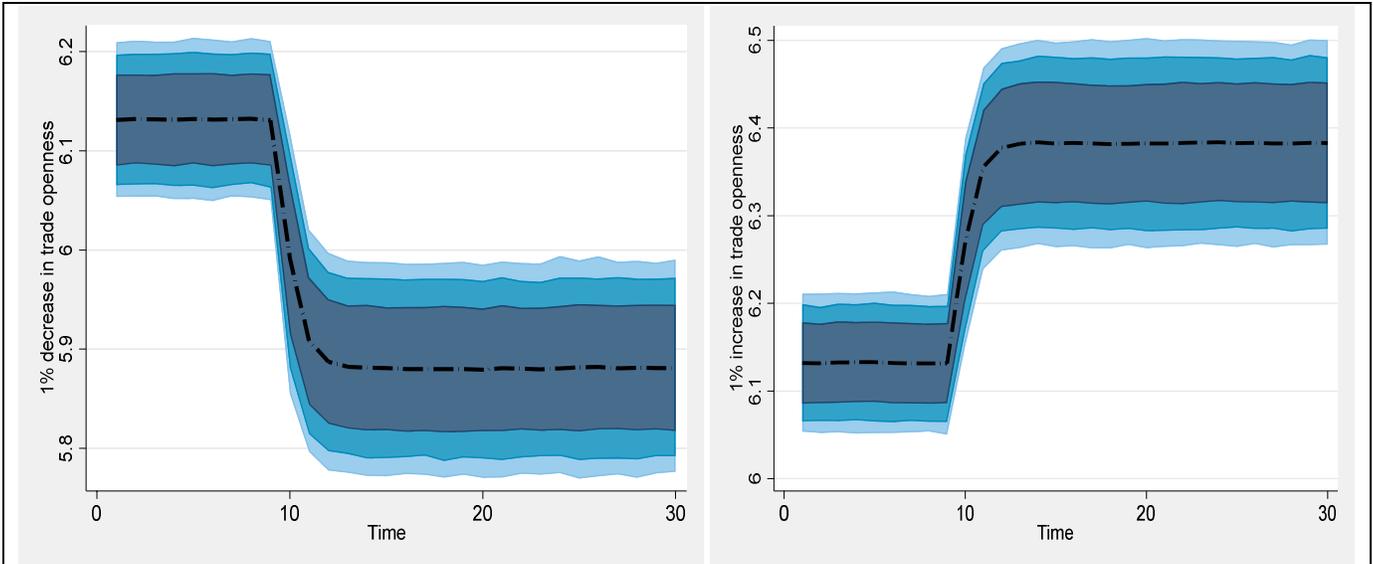
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376
 377 Fig. 4. Effects of 1 % decrease and a 1% increase in predicted urbanization on CO_2 in Egypt. The average predicted values are
 378 indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.

379

380



381 Fig. 5. Effects of 1 % decrease and a 1% increase in predicted trade openness on CO_2 in Egypt. The average predicted values are
 382 indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.
 383

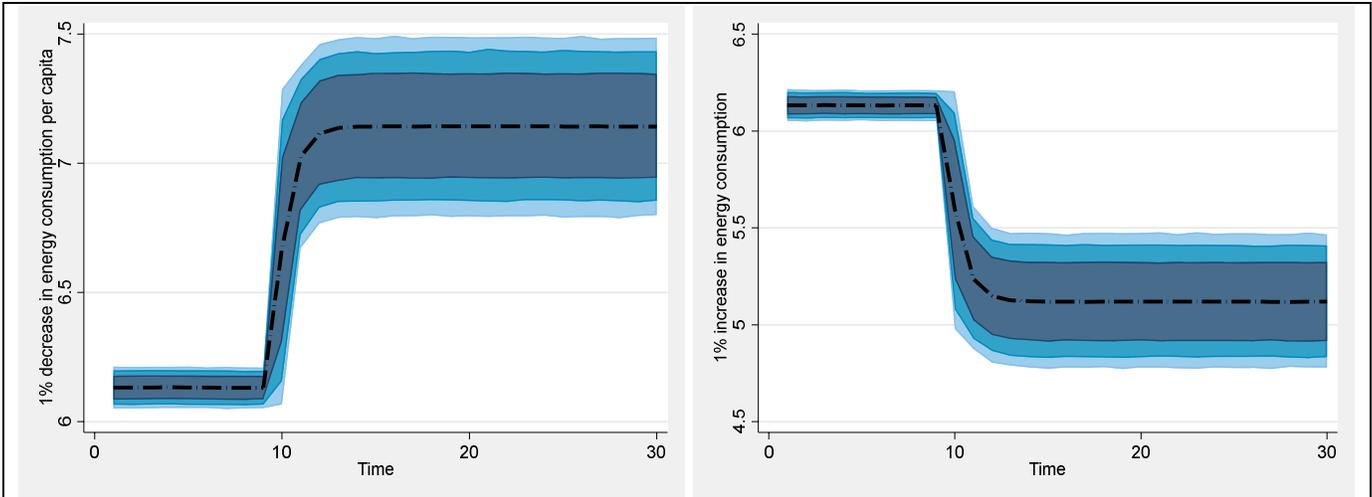
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385 As for the impact of trade openness on the environment in Egypt, results show a harmful effect
 386 environment, which contradicts with (Shahbaz et al., 2013) and recently (Zubair et al., 2020), but
 387 consistent with other works (Asongu, 2018; Sannasse & Seetanah, 2016). A 1% increase in trade
 388 openness results in an increase in carbon dioxide emissions by 0.14% and 0.19% in the short and long
 389 terms respectively. This result supports the political economy approach in explaining the harmful effect
 390 of inequality on the environment. In many developing economies, trade openness and capital account
 391 liberalization led to an increase in imports of goods and encouraged foreign direct investments that work
 392 in anti-environment and many dirty industries which faced many environmental restrictions in developed
 393 countries since the early 1970s (Al-Ayouty et al., 2017). Results also indicate significant and negative
 394 effects of energy consumption on CO_2 emission in both short and long terms. This result contrasts with
 395 the positive effect captured in many recent empirical works in developing countries and needs further
 396 investigation.

397

398





399 Fig. 6. Effects of 1 % decrease and a 1% increase in predicted primary energy consumption per capita on CO_2 in Egypt.
 400 The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%,
 401 and 95% confidence intervals.
 402

403
 404 **5. Conclusion and policy implications**

405 The previous analysis is an extension of the empirical studies assessing the relationship between income
 406 inequality and environmental degradation in developing economies. The study aimed to examine the
 407 extent to which economic policy-makers in Egypt could rely on the political economy approach in
 408 determining the effect of income inequality on environmental degradation during the period (1975-2017).
 409 To this end, the study employed the dynamic ARDL methodology initiated by (Jordan & Philips, 2018)
 410 to estimate the environmental impacts of Gini effects during the short-run and long run. The main finding
 411 of the study is that income inequality harms the environment in Egypt in the long run despite the absence
 412 of a significant relationship between them in the short run. This is consistent with the political economy
 413 approach which believes that income inequality harms the environment. The study also employed the
 414 ARDL model for robustness check and its results are compatible with the Dynamic ARDL results.

415 There are several implications for both economic policy and environmental policy. Firstly, the pattern
 416 of economic development in Egypt was not only biased towards the rich at the expense of the poor, but
 417 rather led to a negative impact on the environment reflected by an increase in carbon dioxide emissions.
 418 This is a complex problem the Egyptian economy suffers, and it can be mitigated by adopting economic
 419 policies that redistribute income in favor of the poor and improve the environmental indicators in Egypt.
 420 Secondly, when studying the factors affecting environmental pollution, it is necessary not to be satisfied
 421 with the real income growth index as a measure of economic activity, but it is also important to consider
 422 the effect of the income distribution structure accompanying the growth process in the estimation model,
 423 especially in developing economies that seek to achieve economic development and the requirements of
 424 sustainable development.

425 Thirdly, it is noticeable according to the previous analysis that it is wrong to continue pursuing the biased
426 economic growth policies for the rich at the expense of the poor in Egypt claiming that inequality improves
427 the environment. The analysis showed that more inequality deteriorated the environment despite the rise
428 in private consumption in Egypt to rates that reached more than 80% of the GDP, which means that the
429 increase in the incomes of the poor will go to more consumption and then more pressures on the
430 environment. This result means that the Keynesian approach in explaining the relationship between
431 inequality and the environment is not valid in the Egyptian economy.

432 Finally, the previous analysis illustrated that there is no trade-off relationship between income inequality
433 and environmental degradation, and therefore economic policy in Egypt will not be subject to choosing
434 one of them and sacrificing the other. Rather, an improvement in income distribution in Egypt will lead
435 to an improvement in the environment and the achievement of some sustainable development goals.

436

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438 1- **Ethics approval and consent to participate:** Not applicable.

439 2- **Consent to Publish:** “Not applicable” for this paper.

440 3- **Availability of data and materials:** The datasets are available in the World Development Indicators
441 data of the World Bank, British Petroleum Statistics (2020), and (Solt, 2020)

442 4- **Competing interests:** The author declares that he has no competing interests.

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444 6- **Author Contributions:** The author contributed to all parts of this article.

445

446

447 **References**

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Figures

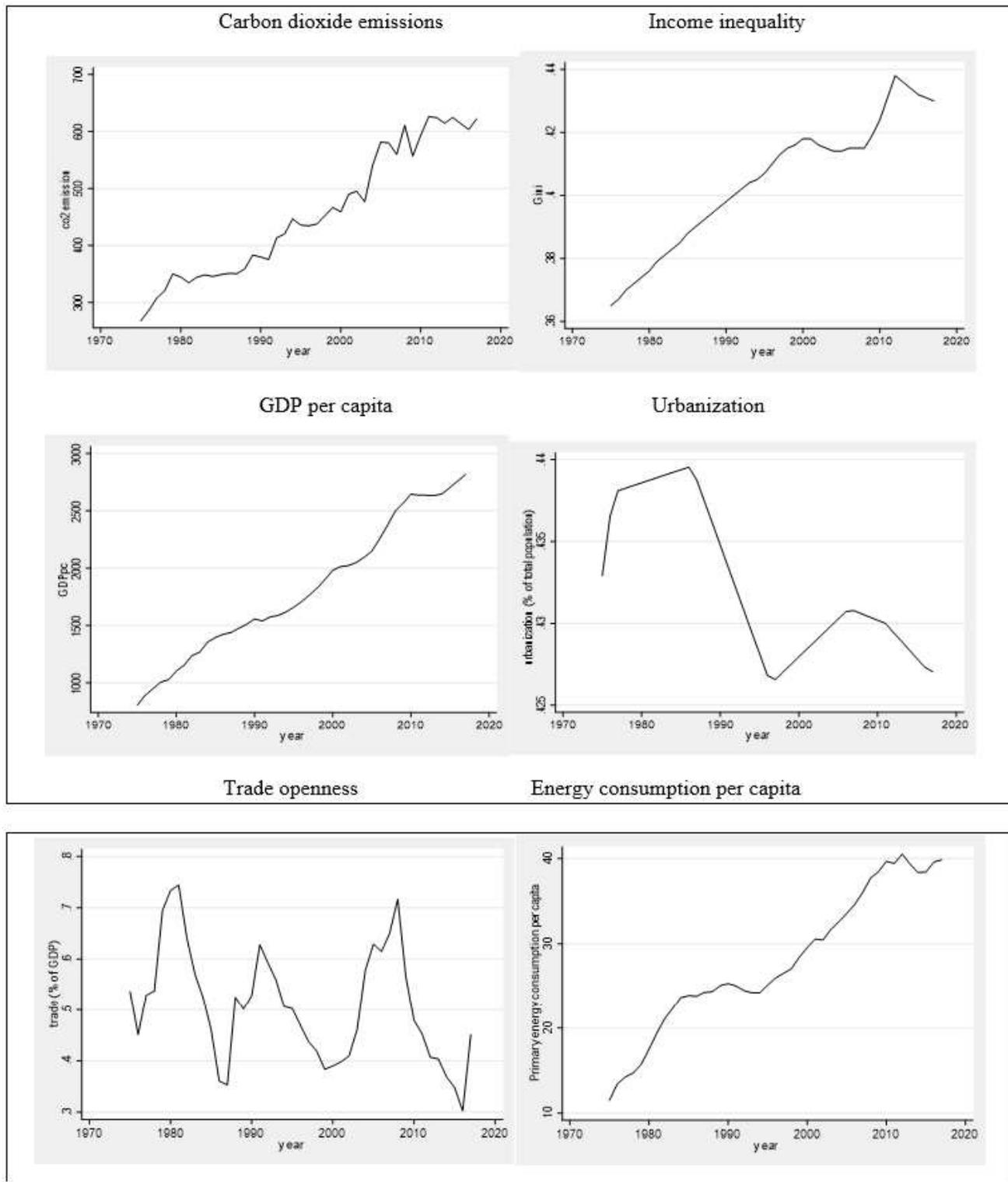


Figure 1

Time path evolution of the series.

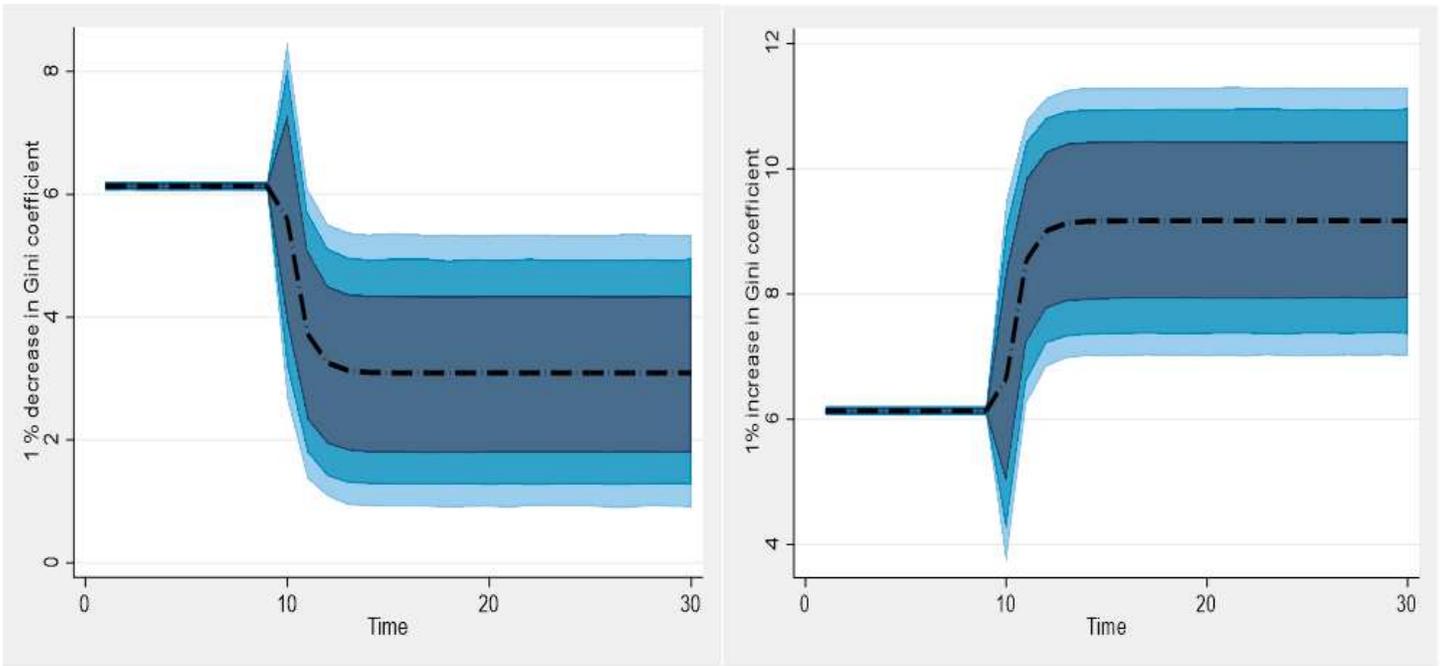


Figure 2

Effects of 1 % decrease and a 1% increase in predicted Gini coefficient on co2 in Egypt. The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) 75%, 90%, and 95% confidence intervals.

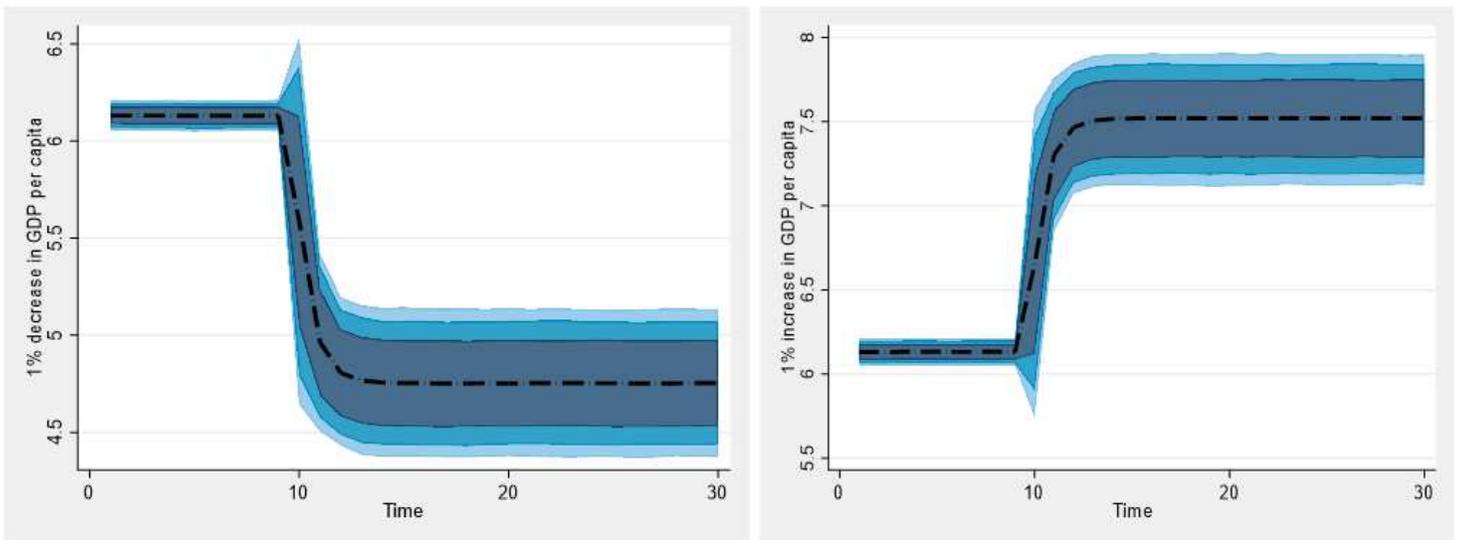


Figure 3

Effects of 1 % decrease and a 1% increase in predicted GDP per capita on co2 in Egypt. The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.

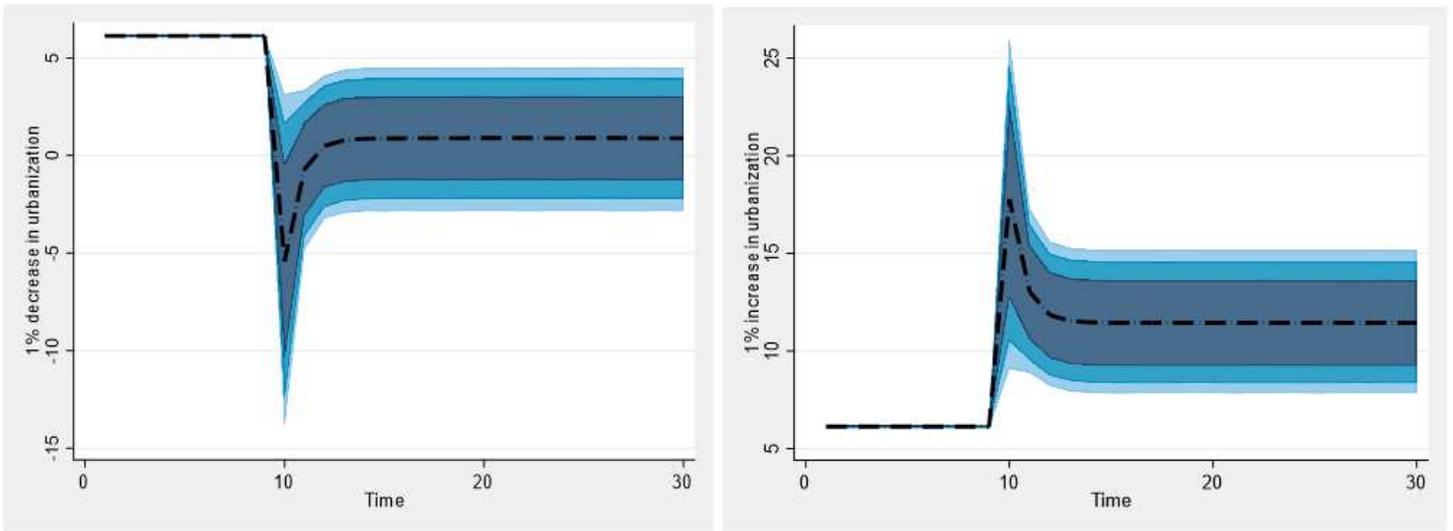


Figure 4

Effects of 1 % decrease and a 1% increase in predicted urbanization on co2 in Egypt. The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.

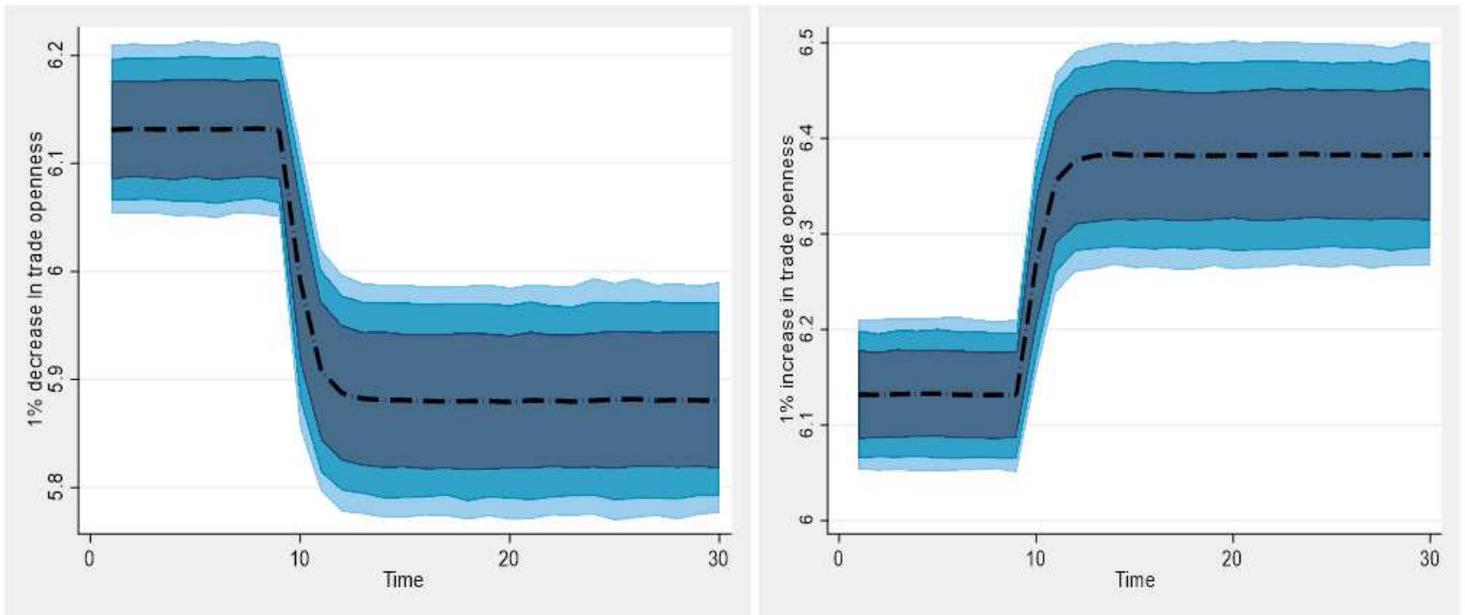


Figure 5

Effects of 1 % decrease and a 1% increase in predicted trade openness on co2 in Egypt. The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.

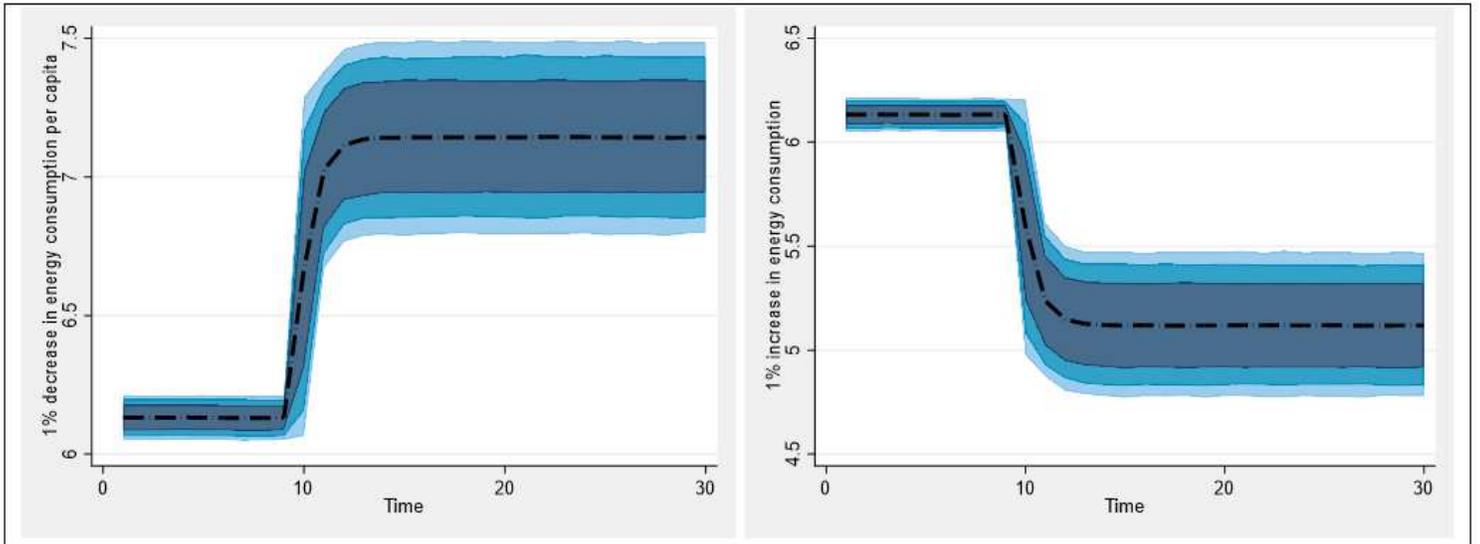


Figure 6

Effects of 1 % decrease and a 1% increase in predicted primary energy consumption per capita on co2 in Egypt. The average predicted values are indicated by the dashed line. The shaded area shows (from darkest to lightest) the 75%, 90%, and 95% confidence intervals.