

The Potency of Eco-Innovation, Natural Resource and Financial Development on Ecological Footprint: A Quantile-ARDL Based Evidence from China

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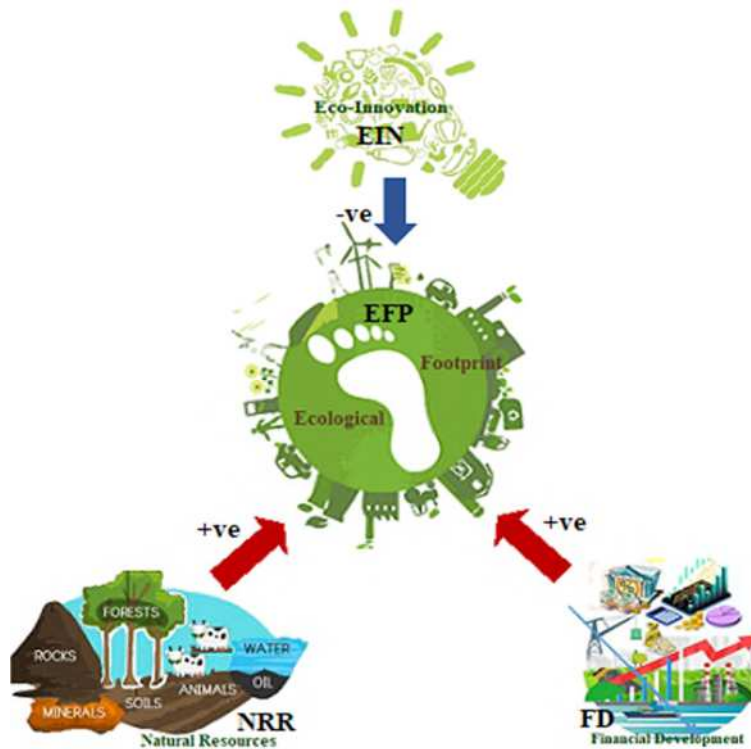
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34 **The potency of eco-innovation, natural resource and financial development on ecological**
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36
37 **Abstract**

38 Given the alarming deterioration of the environment, the present analysis investigates the role of
39 eco-innovation, natural resources and financial development in influencing the environmental
40 degradation of China. Applying the novel method of Quantile-ARDL, the current research is
41 beneficial in portraying the dependence patterns of the variables with special emphasis on the
42 nexus of eco-innovation and ecological footprint across numerous quantiles of the distribution
43 which has not been examined so far in the literature. The empirical findings reveal that in the long
44 run, eco-innovation reduces the level of ecological deterioration in China across all quantiles. On
45 the other hand, the results suggest that the increase in credit to the private sector and natural
46 resource rents augment environmental degradation. The outcomes imply that the over-dependence
47 on natural resources and financial development can worsen the goals of sustainable development
48 in China if the strategies of conservation and management are ignored. Moreover, witnessing the
49 favourable role of eco-innovation, competent policies and regulations can be made towards
50 sustainable efficient technologies and eco-friendly energy sources to halt global warming.

51
52 **Graphical Abstract**



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58 **Keywords:** Ecological footprint; eco-innovation; natural resource; financial development;
59 Quantile-ARDL; China

61 1. Introduction

62 The emission of the green-house gases is cumulating every day, results in the rise of global
63 warming that ultimately responsible for environmental degradation on earth. Among the many
64 other global encounters, dreadful environment is one of the principal challenges of this century
65 (Li, 2016). In compliance to increasing global competition, many countries are urging to increase
66 their growth potential. However, increase in growth without focusing the standards of
67 sustainability could lead to poor quality of environment (Yasmeen et al., 2019), thus much
68 attention is needed to maintain the quality environment along with the significant economic
69 evolution. With the growing knowledge related to environmental threats and necessary actions
70 taken globally, such as Conference of the Parties (COP21), make it realize that excessive usage of
71 energy resources is intimidating our surroundings; hence assimilating the use of technology in
72 numerous growth dependent industries could become a useful tool to improve the quality of
73 environment through sustainable ecological innovations (henceforth, eco-innovation).

74 The idea of eco-innovation arises when strategic policies are fit with ecological assumptions.
75 It is characterized as the number of patents available in the economy particularly in the field of
76 environment. The idea of eco-advancement was first evolved by Fussler and James in 1996 and
77 characterized as the decrease of negative ecological effects while giving new products or services
78 critical for business growth (Hojnik & Ruzzier, 2016). Ecological innovation adds to the natural
79 supportability objectives through the acknowledgment of groundbreaking thoughts, processes,
80 inventions, and cycles (Rennings, 2000). Arundel and Kemp (2009) portray eco- innovation as
81 essentially improved business strategy that assists in diminishing ecological dangers,
82 contamination, and the negative impacts of resource utilization rather than conventional strategies
83 that do not consider the deteriorating impact of economic growth on the nature.

84 At present, economies around the globe are anxiously seeking ways to enhance their growth to
85 excel global competency. Among them, China is considered as the fastest growing economy,
86 accounting thirty percent of the total worldwide growth (Khan et al. 2020). Although, the process
87 of its economic and financial development promotes resource-intensive manufacturing owing to
88 high consumption thereby leaving severe inscriptions on atmosphere (Danish et al. 2019). The
89 country is deemed richest in terms of resources as it is a chief coal producer having big mining
90 industry and the largest user of fossil fuel (Ahmed et al. 2020). On the other hand, it is believed
91 that the increase extraction of natural resources can provide financial gains to the country but cause
92 environment hazard by decreasing bio capacity and enhancing natural burden of the region, in
93 terms of ecological footprint (Panayotou, 1993). Moreover, higher financial advancements
94 motivate foreign direct investments and financial intermediation making consumption easier and
95 prompts energy utilization, diminution of minerals, deforestation, depletion of resources, etc.,
96 thereby worsen the climate deterioration by increasing environmental imbalance in the form of
97 higher ecological footprints (Suki et al. 2020).

98 This study attempts to sightsee the influence of eco-innovation, financial development and
99 natural resources on ecological footprint in case of China. Given the augmented emphasis of
100 literature in analyzing the environmental quality of emerging nations, the current study focuses on

101 the Chinese economy. China has been successful in maintaining the crucial global position being
102 the fastest growing nation and largest manufacturing and exporting economy. However, despite
103 the substantial growth driven policies, the country is considered as the prime carbon emitter in the
104 World (Guo et al. 2019). Given its huge population, China also tops in terms of total ecological
105 footprint around the World. Hence, the current study is significant to analyze the environmental
106 degradation in China since the country is augmenting its focus on innovation and Made-in-China
107 narrative, for which economic and environmental sustainability play significant part. Also, it gives
108 more attention to strong and stable financial structure in order to strengthen the association
109 between real and financial sector of the economy which is prime to resource management and
110 technology innovation.

111 The current study is novel in many ways. First, it seeks to expand EKC theory by incorporating
112 the fundamental aspect of eco-innovation. To the best of our knowledge, the present study is
113 pioneer in analyzing eco-innovation in EKC framework of China, along with the critical variables
114 of natural resource and financial development that have also shown disagreements in the literature
115 regarding their link with the environment (Ahmed et al. 2020). Second, despite the traditional
116 measure of environmental quality in the form of carbon-di-oxide emission (CO₂), the current study
117 has opted the ecological footprint to reflect environmental deterioration in China. The measure of
118 CO₂ has been traditionally utilized in various earlier literatures (Lantz & Feng, 2006; Lindmark,
119 2002; Tisdell, 2001) as it holds the prime segment of green-house gases. However, it only reflects
120 the air quality of the environment. Hence, understanding the importance of the need of an
121 aggregate indicator Solarin and Bello (2018) emphasizes on the alternate of CO₂ emission as a
122 proxy of ecological degradation. In this regard, ecological footprint is characterized as an inclusive
123 measure of environmental quality for considering the environmental impact of economies in terms
124 of soil, air and water (Sharif et al. 2019). Third, observing the emphasis of the literature in adopting
125 advanced econometric methods, the current study has adopted the empirics of Quantile ARDL to
126 perform empirical estimations.

127 Lastly, suitable economic techniques, such as the Bayer-Hanck cointegration test, bootstrap
128 causality, and the ARDL test following the pathway proposed by Kripfganz and Schneider (2018),
129 are employed. These recent econometric techniques offer numerous advantages and are
130 appropriate for reliable estimates. The semi parametric technique Quantile ARDL takes high place
131 among many of the recent econometric and time series techniques. The relaxation of statistical
132 distribution assumptions grabs attention of researchers to consider QARDL for drawing various
133 conclusions related to several stages of ecological footprint. Unlike traditional econometric
134 methodologies, Quantile ARDL elaborates clearer picture of dependent variable by considering
135 numerous quantiles rather than its average relationship with covariates. Understanding the
136 importance of association of environmental degradation with eco-innovation and natural
137 resources, in terms of long-run and short-run dynamics, Quantile ARDL provides the best
138 parameter estimation method for several quantiles of ecological footprint. In order to access the
139 behavior of fat-tail dependent variable by taking its conditional distribution and explore its
140 association with explanatory variables by capturing its asymmetric effects, Quantile ARDL
141 technique is best suited. Lahiani (2018) asserts the significance of Quantile ARDL in the
142 framework of co integration as it empowers the parameters of co integrated vector to vary across
143 the quantiles of ecological footprint arises from structural changes in the time series variable.
144 Furthermore in making policy strategies this technique provides us an opportunity to observe the

145 behavior of innovation at low as well as high quantiles in detail. Due to such benefits QARDL
146 considers one of the sophisticated methodologies to explore the association among the variables
147 in the frame work of co integration in depth.

148 **2. Review of Literature**

149 An ample literature has been well recognized to detect the impact of technological innovation,
150 financial development and natural resources on environmental degradation. Numerous research
151 employed carbon dioxide (CO_2) emissions as a measure of ecological degradation and produce
152 various outcomes leading to valuable policy inferences (need three references of such studies,
153 preferably before 2015). However, a recent shift of literary research has been identified in studying
154 ecological footprint as an indicator of environmental degradation (need three references ideally of
155 last three years).

156 In order to explore the influence of technological on carbon dioxide (CO_2) emissions, Chen
157 and Lee (2020) utilized spatial econometrics model named as STIRPAT (Stochastic Impact by
158 Regressions on population, Affluence and technology) for balanced panel data of 96 countries
159 spanning from 1996-2018. For technological innovation several variables have studied such as
160 innovation total population, GDP, Research and Development expenditures, industrial value
161 added, forest area, urban population, energy use, globalization index, economic, political and
162 social globalization. The study classifies the 96 nations into high, middle and low income
163 economies. The comprehensive analysis exposes that spatial models plays an important role in
164 accessing such kind of relationships. The openness and globalization index plays vital role in
165 decreasing carbon dioxide (CO_2) emissions and raise technological innovations in developed
166 nations.

167 Keeping the panel data of seventeen oil refining companies from all over the globe, Fethi and
168 Rahuma (2020) explored the influence of ecological innovation to reduce carbon dioxide (CO_2)
169 emissions from their operation process. In the context of oil refining companies, the Research &
170 development and training & investment variables have taken as the proxy of economic innovation
171 for all economies. The panel cointegration test and ARDL approach has utilized to achieve the
172 required goals of the study. The study inclusively concludes that ecological innovation is the
173 determinant of reducing carbon dioxide (CO_2) emissions and operation process of the oil refining
174 companies.

175 Taking the panel of top 20 oil refining economies Fethi and Rahuma (2018) inspects the role
176 of ecological innovation on the reduction of carbon dioxide (CO_2) emissions in the scenario of
177 environmental Kuznets curve covering from 2007 till 2016. The selection of specific 20 countries
178 is due to the pledge signed to reduce greenhouse gases down to the prescribed level. Research and
179 development of investment countries was used as a proxy of eco-innovation. The study concludes
180 that all countries under study plays positive role to maintain environment standards and also
181 helping in reducing carbon dioxide (CO_2) emissions.

182 The comparative study between symmetric and asymmetric ARDL by Ahmed et al. (2021)
183 also explored the link between economic globalization, economic growth, financial development
184 and ecological footprint. The aim of the study is to highlight the asymmetric characteristics among

185 the said variables for Japan. Keeping the importance of Population density and energy
186 consumption the study also includes the variables in their econometric model. The study favor the
187 asymmetric ARDL estimates as it captures the nonlinear dynamics. Results of symmetric ARDL
188 clinches that economic globalization and financial development motivate the ecological footprint
189 while asymmetric ARDL indicates that positive and negative variations in the globalization may
190 decrease the footprint in Japan. Overall both the models support the environmental Kuznets curve
191 hypothesis.

192 Understanding the importance of energy consumption, financial development, economic
193 growth and ecological foot print, Destek and Sarkodie (2019) scrutinized the rationality of
194 environmental Kuznets curve. The enormous yearly data of 11 newly industrialized economies
195 spanning from 1977 to 2013 was taken for empirical investigation. It was found that energy
196 consumption in industrialized countries and ecological footprints are positively related with each
197 other. The group panel estimation method supports the inverted U-shaped EKS hypothesis along
198 with the existence of bi-directional causal relationship between GDP and ecological footprint in
199 the countries under study.

200 On the other hand, Ahmed et al. (2020) studied the influence of natural resources, economic
201 growth and technological innovations on ecological foot prints. Taking the panel data of 22
202 emerging countries from 1984 to 2016, the required objective has been achieved by authors under
203 the framework of panel cointegration and causality tests. The fallouts of the study indicate that
204 natural resources, GDP, technological innovation and ecological footprint possess long run
205 relationship with each other. The CS-ARDL test urges that economic growth and ecological foot
206 prints hold environmental Kuznets curve hypothesis. Moreover technological innovations reveal
207 negative relationship with ecological footprint thus output growth, natural resources and
208 technological innovations leave marks on ecological footprint.

209 Hassan et al. (2019) also inspects the link among natural resources, output growth and
210 ecological footprint along with the control variables human capital, bio-capacity and urbanization
211 of Pakistan for the yearly data over the period 1970-2014. The objective of the study is to examine
212 the cogency of environmental Kuznets curve hypothesis in Pakistan. The ARDL approach reveals
213 that economic growth is playing positive role and favors EKC hypothesis while human capital and
214 bio-capacity do not improving the ecological footprint progression in the scenario of Pakistan.
215 Furthermore the bidirectional causality is evident by employing ARDL granger causality test
216 between ecological footprint and bio capacity.

217 Emphasizing on the role of financial development, Baloch et al. (2018) collect the data of 59
218 belt and road Countries (BRI) and attempts to investigate the impact of financial development on
219 ecological footprint over the period of 1990-2016. Along with financial development, the authors
220 also take in account the role of GDP, energy, FDI and urbanization as explanatory variables. As
221 per suggestion of World Bank, the study utilized domestic credit provided by the private sector,
222 domestic credit provided by the financial sector, and domestic credit provided by the banking
223 sector as the measure of financial development which provides deep insight for BRI nations. The
224 estimation result reveals that financial development and economic growth have positive influence
225 while foreign direct investment leads in an increment of ecological footprint.

226 Recently, Aydin and Turan (2020) consider the BRICS nation to sightsee the impact of
227 financial openness, trade openness, and energy intensity on ecological footprint for the data set
228 comprising from 1996-2016 under the shade of environmental Kuznets curve hypothesis. Authors
229 employed three different models for the assessment of EKC one with financial openness, second
230 with trade openness and third is interaction model with both variables. The study finds
231 cointegration between the variables by utilizing the Westerlund's (2008) panel cointegration test.
232 Overall the study clinches that energy intensity contributes to raise the pollution in all BRICS
233 countries except Russia. For China and India trade openness has dropping effect on environment
234 pollution. The results also specify that the model with financial openness and the model
235 incorporated with trade openness favors EKC hypothesis for India and South Africa respectively.

236 Wang et al. (2020) explores the role of ecological innovation and export diversification on the
237 level of carbon dioxide (CO_2) emissions. The study covers the panel data of G-7 countries
238 straddling from 1990 to 2017. Taking cross sectional dependence and slope heterogeneity into
239 account the order of integration varies for the selected variables thus CS-ARDL model finds
240 appropriate for estimation. The estimation result exposes that ecological innovation is negatively
241 related while export diversification is positively linked with carbon dioxide (CO_2) emissions thus
242 ecological innovation helps to improve the standard of environment by decreasing the level of
243 carbon dioxide (CO_2) emissions for G-7 economies.

244 Utilizing the Quantile ARDL model Godil et al. (2020) studies the asymmetric impact of
245 tourism, financial development and globalization on ecological footprint for Turkey in terms of
246 environmental Kuznets curve hypothesis. The nonlinear and asymmetric behavior of variables has
247 been evident through Wald test of parameter dependency. The estimates are observed at different
248 quantiles for accessing the short and long run association among the said variables. Overall the
249 explanatory variables are showing substantial effect on ecological footprint as well as the validity
250 of environmental Kuznets curve hypothesis was also observed in Turkey.

251 The relationship between natural resources, human capital, urbanization, economic growth,
252 and ecological footprint for the emerging economy China has been studied by Ahmed et al. (2020).
253 The large span of data over the period of 1970 to 2016 was collected and ARDL model was
254 employed to explore the short and long run dynamics. The environmental Kuznets curve
255 hypothesis is no more valid according to the long run coefficients while ecological footprint
256 exhibits the positive relationship with economic growth. The short run estimates disclose that
257 economic growth and urbanization in China increases ecological footprint whereas human capital
258 is found to be negatively insignificant.

259 Considering the ASEAN countries Kongbuamai et al. (2019) analyzed the influence of
260 economic growth, energy consumption, tourism, and natural resources on the ecological footprint.
261 The Driscoll-Kraay panel regression has been deployed over the data from 1996 to 2016. The
262 estimation results asserts that tourism brings ASEAN countries towards good environment
263 conditions along with the validity of inverted U-shaped EKC performance between GDP and
264 ecological footprint.

265 The above review reveals that several studies took carbon dioxide (CO_2) emissions as an
266 environmental degradation and numerous bases of the ecological footprint have explored

267 comprising technological innovation, human capital, foreign direct investment, natural resources
 268 and financial development but none of the studies has considered the influence of ecological
 269 innovation on the ecological footprint. To bridge the gap in the literature, this lacking inspires us
 270 to estimate the effect of ecological innovation on the ecological footprint in the scenario of China.

271 3. Methodology

272 In order to examine the association of eco-innovation, natural resource rent and financial
 273 development with ecological footprint over a range of quantiles, the application of QARDL is the
 274 suitable approach for providing an inclusive dependence structure among the variables in different
 275 levels of the criterion variables in both short-run and long-run. Therefore, to investigate the short-
 276 run and long-run association of the studied variables under the EKC framework, we have applied
 277 the QARDL framework and Wald test to inspect the significance and constancy of the parameters
 278 across the series of quantiles.

279 The model of QARDL initiated from the traditional ARDL framework, written in equation 1 as;

$$280 Y_t = \alpha + \sum_i^p \beta_1 Y_{t-i} + \sum_i^q \beta_2 X1_{t-i} + \sum_i^q \beta_3 X2_{t-i} + \sum_i^q \beta_4 X3_{t-i} + \sum_i^q \beta_5 X4_{t-i} + \\ 281 \sum_i^q \beta_6 X5_{t-i} - 1 + \epsilon_t \quad (1)$$

282 where, ϵ_t is the error term, explained as $Y_t - \Sigma[Y_t/\delta_{t-1}]$, with δ_{t-1} indicated the least σ – field
 283 produced by $\{Y_t, X1_t, X2_t, X3_t, X4_t, X5_t, Y_{t-1}, X1_{t-1}, X2_{t-1}, X3_{t-1}, X4_{t-1}, X5_{t-1}\}$, also, p & q identified
 284 the lag-orders measured via Schwarz information criterion and Y_t the studied criterion variables,
 285 i.e. ecological footprint, whereas, X1, X2, X3, X4 and X5 represent GDP, squared GDP, eco-
 286 innovation, natural resource rent and financial development respectively.

287 Linking from equation (1), the QARDL model is identified as;

$$288 Q_{Y_t} = \alpha(\tau) + \sum_i^p \beta_1(\tau) Y_{t-i} + \sum_i^q \beta_2(\tau) X1_{t-i} + \sum_i^q \beta_3(\tau) X2_{t-i} + \sum_i^q \beta_4(\tau) X3_{t-i} + \\ 289 \sum_i^q \beta_5(\tau) X4_{t-i} + \sum_i^q \beta_6(\tau) X5_{t-i} + \epsilon_t(\tau) \quad (2)$$

290 In which, $\epsilon_t(\tau) = Y_t - Q_{Y_t}(\tau/\epsilon_{t-1})$ (Kim and White, 2003) and $0 < \tau < 1$ is the quantile. For
 291 empirical purposes, the subsequent quantiles represent a total of nine quantiles represented from τ
 292 showing the quantiles of $\{0.05, 0.10, 0.20, 0.30, 0.40, 0.50, 0.60, 0.70, 0.80, 0.90$ and $0.95\}$.
 293 Furthermore, suspecting the probability of having serial correlation in the residual, the
 294 generalization of QARDL model is presented in Equation (3) as;

$$295 Q_{\Delta Y_t} = \alpha(\tau) + \rho Y_{t-i} + \beta_1 X1_{t-i} + \beta_2 X2_{t-i} + \beta_3 X3_{t-i} + \beta_4 X4_{t-i} + \beta_5 X5_{t-i} + \\ 296 \sum_{i=1}^p \varphi_i \Delta Y_{t-1} + \sum_{i=1}^q \omega_i \Delta X1_{t-1} + \sum_{i=1}^q \lambda_i \Delta X2_{t-1} + \sum_{i=1}^q \theta_i \Delta X3_{t-1} + \sum_{i=1}^q \mu_i \Delta X4_{t-1} + + \\ 297 \sum_{i=1}^q \gamma_i \Delta X5_{t-1} + \epsilon_t(\tau) \quad (3)$$

298 In addition, Equation (3) can be revised following (Cho et al., 2015) to provide ECM re-estimations
 299 of QARDL framework in model Equation (4) as;

$$\begin{aligned}
300 \quad Q_{\Delta Y_t} &= \alpha(\tau) + \rho(\tau)(Y_{t-i} - \omega_i(\tau)X1_{t-i} - \lambda_i(\tau)X2_{t-i} - \theta_i(\tau)X3_{t-i} - \mu_i(\tau)X4_{t-i} - \\
301 \quad \gamma_i(\tau)X5_{t-i}) &+ \sum_{i=1}^{p-1} \beta_1(\tau)\Delta Y_{t-i} + \sum_{i=0}^{q-1} \beta_2(\tau)\Delta X1_{t-i} + \sum_{i=0}^{q-1} \beta_3(\tau)\Delta X2_{t-i} + \\
302 \quad \sum_{i=0}^{q-1} \beta_4(\tau)\Delta X3_{t-i} &+ \sum_{i=0}^{q-1} \beta_5(\tau)\Delta X4_{t-i} + \sum_{i=0}^{q-1} \beta_6(\tau)\Delta X5_{t-i} + \epsilon_t(\tau)
\end{aligned} \tag{4}$$

303 Furthermore, utilizing the delta approximations, the additive short-term influence of
304 historic environmental degradation represented by eco-footprint on present level of eco-footprint
305 is calculated by $\beta_* = \sum_{i=1}^{p-1} \beta_1$, however, the additive short-term influence of concurrent and past
306 level of eco-innovation, natural resources rent and financial development on present level of
307 environmental degradation is measured as $\beta_* = \sum_{i=1}^{q-1} \beta_2$. Similarly, the long-term cointegrating
308 estimate β is measured using the delta approximation $\beta \frac{1}{4} - \gamma \rho$. Lastly, the speed of adjustment
309 estimates ρ in Eq. (4) is needed to be negative significant (Lahiani, 2018). Finally, in order to
310 estimate the long-term and short-term asymmetric effect of X1, X2, X3, X4 and X5 on Y_t , the
311 study applied the method of Wald to testify the null hypotheses of the long-term and short-term
312 asymmetry (Cho et al. 2015; Lahiani, 2018).

313 Results and Interpretations

314 To fulfil the required objective of the present study, preliminarily analysis of the data by
315 descriptive statistics and unit root testing methods has been employed followed by Q-ARDL
316 model. Table 1 presents the outcome of summary statistics for ecological footprint, economic
317 growth, natural resource rent, eco-innovation and financial development. The square of economic
318 growth has also considered observing the validity of EKC hypothesis in the scenario of China. The
319 average of outcome variable i.e. eco-footprint possess minimum value of 0.351 with the least
320 volatility of 0.011 among all the considered variables. Economic growth and financial
321 development holds large mean indicating that both variables posing almost same effect on China's
322 economic cycle. On average, the share of eco-innovation is more than natural resource rent while
323 the normality of all the variables has been rejected at 1% level of significance declaring the heavy
324 tails of the data (Sharif et al. 2019).

Table 1: Results of Descriptive Statistics

Variables	Mean	Min.	Max.	Std. Dev.	J-B Stats
EFP	0.351	0.201	1.123	0.011	16.010***
GDP	2.010	1.101	3.014	1.101	20.106***
GDP2	0.654	0.510	1.010	0.021	18.010***
ECO	1.121	1.010	2.011	1.010	22.101***
NRR	0.985	0.650	1.345	0.032	17.021***
FD	2.101	1.801	3.101	1.001	19.122***

Note: The asterisk ***, ** and * represent level of significance at 1%, 5% and 10% respectively.

325 The outcome of unit root tests is portrayed in Table 2. Similar to Godil et al. (2020) the
326 study deployed two unit root tests. It has been evident from the results of traditional ADF test and

327 ZA test incorporating structural breaks in the series that all the variables are integrated of order 1.
 328 The break years at first difference for eco-footprint are found to be 2001(Q2) indicating that China
 329 revised its environmental regulations to save the environment. Breaks year for eco-innovation at
 330 level and at first difference are quite close i.e. 2000(Q3) and 1999(Q1) which is the clue of
 331 involvement of this explanatory variable in China. Natural resource rent and financial
 332 development is seems to be much affected by few of the financial crisis happened all around the
 333 globe at several time periods.

Table-2: Results of Unit root test

Variables	ADF (Level)	ADF (Δ)	ZA (Level)	Break Year	ZA (Δ)	Break Year
EFP	-1.210	-3.015***	-1.425	2013(Q4)	-6.010***	2001(Q2)
GDP	-0.030	-5.010***	-0.654	2016(Q1)	-9.100***	2010(Q4)
ECO	-0.624	-3.009***	-0.741	2000(Q3)	-9.035***	1999(Q1)
NRR	-1.101	-5.011***	-1.357	2015(Q1)	-6.067***	2012(Q2)
FD	-0.451	-4.007***	-0.587	2014(Q3)	-8.020***	2005(Q1)

Note: The values in the table specify statistical values of the ADF and ZA test. The asterisk ***, ** and * represent level of significance at 1%, 5% and 10% respectively. Source: Author Estimation

334
 335 The evidence of highly skewed data points and confirmation of I (1) variables leads the
 336 analysis towards Q-ARDL framework. Table 3 reports the empirical findings of the model.
 337 Looking towards the error correction terms at each quantile it finds to be negative and highly
 338 significant with uniform speed of adjustment i.e. around 40% rate to maintain its equilibrium
 339 relationship with explanatory variables under studied. To check the validity of EKC hypothesis,
 340 keeps an eye on the quadratic term of economic growth under the caption of long run estimates
 341 the coefficient are found to be negative and significant especially at extreme quantiles from 0.6 to
 342 0.95. At 0.2, 0.3 0.4 and median the impact is although negative but it is insignificant, however
 343 the U-shaped inverted relationship among economic growth and environmental hazards in China
 344 are evident at higher quantiles. Thus in China economic growth leads eco-footprint towards
 345 downward trend in the early stage while reaching at maximum level improvement in economy
 346 started to harm environment. These results are in consistent with the Godil et al. (2020) as they
 347 found positive coefficients of squared economic growth across high quantiles, also not in line with
 348 Wang et al. (2013) as they conclude that EKC hypothesis is not valid for the countries holding
 349 more than 1 million of population while the results are similar to Dogan et al. (2019) and Destek
 350 and Sarkodie (2019).

351
 352 Furthermore betas of eco-innovation are found to be negative and highly significant at all
 353 quantiles indicating the negative association among eco-innovation and eco-footprint in China.
 354 This result is consistent with Ahmad et al. (2021) as eco-innovation helps in reducing the
 355 environmental degradation. In the light of these outcome it can be stated that enhancement in eco-
 356 innovation in China leads it toward healthy atmosphere as it possess same impact at all quantiles
 357 of eco-footprint. Eco-innovation is helping China to make their environment clean as health
 358 concerns are widely be connected with surroundings.

359 As far as natural resource rent is concerned. It is exerting positive and significant impact on
 360 eco-footprint in the long run, showing that natural resources are not helping China to reduce the

361 consumption of imports, usage of fossil fuels is rising and contaminated energy sources are widely
362 be utilized in the country. These results are dissimilar to the study of Zafar et al. (2019) in which
363 natural resource rent was found to be negative and significant for the scenario of the USA and
364 Hassan et al.(2018) for Pakistan as insignificant relationship has been observed among natural
365 resources and eco-footprint.

366
367 It is also noteworthy that financial development is seems to be positive and significant at all
368 quantiles showing upward trending relationship in the long run with environmental degradation.
369 This result is so far match with the study of Baloch et al. (2018) for BRI countries and mismatch
370 to Uddin et al. (2017) asserts that financial development helps to reduce climatic hazards. There
371 are several reasons behind the positive relationship of financial development with eco-footprint
372 for the case of China as FD encourages stakeholders to invest in infrastructure plans by providing
373 flexible loan schemes resulting in increasing land, water and air pollution. Moreover financial
374 development empowers general public to increase their standard of life by purchasing luxuries
375 through loans which ultimately harms the environment.

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377 Like FD Economic growth in China is also found to be positive and significant across all the
378 quantiles. It is well documented that economic progress is extensively based on industrialization
379 and to boost industrial sector conventional sources such as oil, gas and coal are using in China.
380 Furthermore huge investments and purchases in the country bring economic growth but decrease
381 the quality of environment.

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Table 3: Results of Quantile Autoregressive Distributed Lag (QARDL) for Ecological Footprint

Quantiles	Constant	ECM	Long-Run Estimation					Short-Run Estimation					
	$\alpha_*(\tau)$	$\rho_*(\tau)$	$\beta_{GDP}(\tau)$	$\beta_{GDP2}(\tau)$	$\beta_{EIN}(\tau)$	$\beta_{NRR}(\tau)$	$\beta_{FD}(\tau)$	$\phi_1(\tau)$	$\omega_0(\tau)$	$\lambda_0(\tau)$	$\theta_0(\tau)$	$\mu_0(\tau)$	$\nu_0(\tau)$
0.05	0.101	-0.450***	0.215	-0.248**	-0.269*	0.108	0.359**	0.204	0.267*	-0.150**	-0.467	0.046	0.076
	(0.011)	(-3.001)	(0.010)	(-2.030)	(-1.649)	(1.609)	(2.098)	(1.151)	(1.930)	(-2.889)	(-1.609)	(0.064)	(0.064)
0.1	0.001	-0.439***	0.241	-0.151*	-0.477*	0.012	0.449**	0.102	0.179*	-0.191***	-0.483	0.037	0.101
	(0.020)	(-2.991)	(0.006)	(-1.961)	(-1.757)	(1.089)	(2.304)	(0.898)	(1.895)	(-3.019)	(-1.617)	(0.150)	(0.080)
0.2	0.010	-0.430**	0.257*	-0.147	-0.389*	0.103	0.360**	0.315	0.269*	-1.107	-0.437	0.033	0.071
	(0.032)	(-2.700)	(1.856)	(-0.079)	(-1.696)	(1.105)	(2.501)	(0.787)	(1.881)	(-0.689)	(-1.610)	(0.044)	(0.178)
0.3	0.106	-0.413*	0.248*	-0.157	-0.041**	0.208**	0.251**	0.347	0.258*	-0.143*	-0.359***	0.040	0.140
	(0.019)	(-1.864)	(1.780)	(-0.059)	(-1.991)	(2.010)	(2.535)	(1.175)	(1.849)	(-1.651)	(-2.998)	(0.059)	(0.085)
0.4	0.035	-0.433*	0.259**	-0.139	-0.039*	0.250**	0.039	0.138	0.371*	-0.048	-0.432**	0.069	0.029
	(0.011)	(-1.831)	(1.998)	(-0.081)	(-1.781)	(2.589)	(0.089)	(0.161)	(1.769)	(-1.129)	(-2.399)	(0.278)	(0.148)
0.5	0.018	-0.435**	0.249**	-0.131	-0.049**	0.119*	0.023	0.339	0.252**	-0.102**	-0.143	0.044	0.032
	(0.011)	(-2.725)	(2.234)	(-0.061)	(-2.003)	(1.817)	(0.078)	(1.095)	(2.091)	(-2.109)	(-1.561)	(0.257)	(0.167)
0.6	0.052	-0.421*	0.260**	-0.141**	-0.573**	0.290**	0.032	0.241	0.165*	-0.059	-0.031	0.041	-0.051
	(0.021)	(-1.860)	(2.001)	(-1.983)	(-2.701)	(1.968)	(0.062)	(0.096)	(1.666)	(-0.919)	(-0.999)	(0.051)	(-0.059)
0.7	0.109	-0.425**	0.252**	-0.163**	-0.681**	0.279**	0.027	0.223	0.172*	-0.131	-0.069	0.026	-0.038
	(0.041)	(-2.778)	(2.035)	(-2.010)	(-2.089)	(2.059)	(0.049)	(0.189)	(1.879)	(-0.979)	(-0.980)	(0.038)	(-0.089)
0.8	0.078	-0.427*	0.247**	-0.148***	-0.490***	0.282**	0.541***	0.133	0.137	-0.099**	-0.072	0.086	-0.031
	(0.015)	(-1.852)	(2.029)	(-2.991)	(-3.081)	(1.941)	(3.091)	(0.098)	(0.058)	(-1.982)	(-1.236)	(0.147)	(-0.041)
0.9	0.106	-0.429**	0.239**	-0.153***	-0.679***	0.209*	4.392***	0.142	0.106	-0.227*	-0.191	0.045	-0.057
	(0.055)	(-2.710)	(2.050)	(-3.010)	(-3.010)	(1.727)	(2.995)	(0.162)	(0.043)	(-1.714)	(-1.555)	(0.031)	(-0.032)
0.95	0.069	-0.447***	0.251**	-0.172***	-0.688***	0.341**	4.049***	0.337	0.238**	-0.153*	-0.169	0.037	-0.044
	(0.016)	(-3.010)	(2.038)	(-2.996)	(-2.979)	(1.989)	(3.005)	(0.149)	(2.035)	(-1.959)	(-0.971)	(1.021)	(-0.022)

Note: The t-statistics are between brackets. ***, ** and * indicate significance at the 1%, 5% and 10% levels, respectively. Source: Author Estimations

406 Table 4 reports the parameter constancy outcomes accessed through Wald test. The null hypothesis
 407 of linearity in error correction term is rejected at 5% level of significance in the scenario of China.
 408 Similarly the long-term integrating parameter β_{EIN} has been tested for the parameter constancy for
 409 all quantiles and it found to be significant. This result may assert that the long term relationship
 410 among eco-footprint and eco-innovation varies across all the quantiles. The parameter estimates
 411 of β_{GDP2} , β_{GDP} , β_{NRR} and β_{FD} contains mixed results as they found insignificant Wald statistics at
 412 various quantiles.

Table 4: Results of the Wald Test for the constancy of parameters

Variables	Wald-statistics
ρ	3.010*** (0.000)
β_{GDP}	4.012*** (0.000)
β_{GDP2}	1.991** (0.051)
β_{ECO}	2.791** (0.000)
β_{NRR}	3.101*** (0.000)
β_{FD}	2.980*** (0.000)
φ_1	3.099*** (0.000)
ω_0	2.890** (0.000)
λ_0	0.927 (0.789)
θ_0	3.101*** (0.000)
μ_0	1.591 (0.426)
γ_0	1.357 (0.321)

Note: The p-values are between square brackets. ***, ** and * indicate significance at the 1%, 5% and 10% levels, respectively.

413 As a final step of the procedure the Granger Causality test across the quantiles developed by
 414 Troster et al. (2018) has also employed (Displayed in Table 5). The purpose of the test is to observe
 415 the bi-directional causal relationship from the variations explanatory variables to the outcome
 416 variable. The results suggest the presence of bi-directional causality among the variables across all

417 the quantiles. Therefore it can be stated that eco-footprint possess bi-directional relationship with
 418 eco-innovation, economic growth, natural resource rent and financial development in China.

Table 5: Granger Causality in Quantile Test Results

Quantiles	ΔEFP_t to ΔGDP_t	ΔGDP_t to ΔEFP_t	ΔEFP_t to ΔEIN_t	ΔEIN_t to ΔEFP_t	ΔEFP_t to ΔNRR_t	ΔNRR_t to ΔEFP_t	ΔEFP_t to ΔFD_t	ΔFD_t to ΔEFP_t
[0.05-0.95]	0.000	0.000	0.000	0.000	0.359	0.000	0.000	0.000
0.05	0.000	0.000	0.006	0.000	0.005	0.002	0.000	0.009
0.10	0.000	0.000	0.010	0.000	0.000	0.000	0.000	0.000
0.20	0.007	0.000	0.001	0.000	0.051	0.008	0.000	0.000
0.30	0.001	0.000	0.005	0.000	0.164	0.000	0.001	0.000
0.40	0.000	0.000	0.011	0.000	0.597	0.000	0.000	0.000
0.50	0.000	0.000	0.015	0.007	0.002	0.045	0.000	0.000
0.60	0.000	0.000	0.012	0.000	0.000	0.000	0.000	0.001
0.70	0.000	0.000	0.009	0.000	0.039	0.000	0.000	0.000
0.80	0.000	0.000	0.013	0.000	0.698	0.000	0.000	0.000
0.90	0.000	0.000	0.004	0.000	0.211	0.001	0.003	0.000
0.95	0.000	0.004	0.000	0.000	0.342	0.000	0.000	0.006

Source: Authors Estimation

419

420 Conclusion and Policy Recommendation

421 The present examination analyzed the validity of EKC in China by analyzing the influence
 422 of growth, squared growth, eco-innovation, natural resource rent and financial development on
 423 environmental degradation of China by utilizing the measure of ecological footprint over the
 424 period of 1990 to 2017. The current study has applied the method of Q-ARDL in examining the
 425 association among the studied variables. The research is beneficial in portraying the dependence
 426 patterns of the variables with special emphasis on the relationship of eco-innovation with
 427 ecological footprint of China across numerous quantiles of the distribution which has not been
 428 examined so far in the literature. Also, the novel method of granger causality in quantiles is also
 429 applied in the current study to examine the causal connections among the studied variables. The
 430 causal outcomes found that all the variables under study preserve feedback association with each
 431 other except for natural resource rent. The study found that natural resource has a uni-directional
 432 causal connection with ecological footprint in China, where the direction of causality runs from
 433 NRR to EFP. Moreover, the results of Q-ARDL revealed that in short-run, GDP and GDP² are
 434 significant to impact environmental degradation in China. Also in long-run, the study validated

435 the presence of inverted U-shaped relationship among the variable exhibiting that economic
436 activity puts negative pressure on the environment in initial stages, however it transforms to
437 generate positive impact on the ecological condition of the country. Moreover, the outcomes of
438 the Q-ARDL reflected that the error correction is significant across all quantiles with the
439 anticipated negative sign confirming the existence of notable return to equilibrium association in
440 long-run among the studied variables and ecological footprint.

441 In long-run, the results exhibited that eco-innovation reduces the level of ecological
442 deterioration in China across all quantiles. These results are consistent with the findings of Ahmad
443 et al. (2021) in G-7 countries. Hence, considering the positive role of eco-innovation in reducing
444 the ecological deficit, further training and awareness campaigns should be initiated in cities, rural
445 and remote areas of the country to explain the significance of eco-innovation together with the
446 means to promote reduced resource utilization and sustainable functioning. Moreover, the
447 government policies should contain provisions for financing ecologically innovated projects,
448 particularly for creating suitable technologies that can guarantee desired balance between
449 increased economic developments and reduced ecological deterioration. Finally, higher level of
450 monetary resources and research on eco-innovation can be utilized to decline the potential issue of
451 energy security and obsolete technology convention in the country.

452 On the other hand, the result of financial development revealed that increase in credit to
453 private sector further enhances the environmental degradation, similar to the outcome of
454 Przychodzen and Przychodzen (2015) for the economies of Poland and Hungary. The results are
455 more evident in the high quantiles explaining that greater ecological deficit is more vulnerable to
456 increased financial development in China. These findings highlight that financial advancements in
457 China is non-sustainable indicating the ignorance of financial sector in preserving the standards of
458 sustainable development in its investments. In order to deal with the current situation, the
459 government needs to regulate the financial sector to consider ecological feasibility of investment
460 projects. Also the utilization of improved technology in rural areas facing the problem of declining
461 labor due to aging and less productivity (Wei, Zhu & Glomsrød, 2018), could also ensure
462 optimistic influence on the level of augmenting ecological deficit of the country.

463 Moreover, the outcome of natural resource found the positive association of natural
464 resource rent with ecological footprint in China consistent with the outcomes of Ahmed et al.,
465 (2020). This imply that increased utilization of natural resources puts negative pressure on
466 environment, not only by causing air and land pollution, but also by augmenting the ecological
467 deficit of the country. This necessitates that the prevailing regulations concerning minerals, soil
468 and emission in China requires modification and enforcement. Moreover, given the higher reliance
469 of coal in the power mix of the country, the shift towards natural gas has been recommended by
470 many studies to lessen the environmental burden (Chen et al., 2020). This may lead to further puts
471 negative pressure on ecological deficit by triggering impending water contamination, exhaustive
472 water utilization and enhanced emissions (Mallapragada et al. 2018) causing enlarged pressure on
473 ecological deficit through higher gas mining. As a more suitable solution, the emphasis of
474 government should be towards the adoption of green energy sources. In doing so, the present
475 policies that intend to attain 40% share of renewables in power sector by 2030 should be sustained.
476 However, this will require hastened development in solar and wind PV along with the utilization
477 of hydro-electricity obliging higher development in grid & transmission aptitude together with the

478 reforms in energy market. Also, the government can modify pricing strategy to decline the mining
479 of oil and coal and encourage the utilization of eco-friendly power sources which may lead to
480 resource restoration with lower natural resource depletion and the protection of bio-diversity with
481 lower ecological footprint.

482 **DECLERATIONS:**

483 **Ethical Approval**

484 Not applicable

485

486 **Consent to Participate**

487 Not applicable

488

489 **Consent to Publish**

490 The authors provide consent for the publication of manuscript submitted including all the details
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492

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494 Sahar Afshan Conceptualization, Writing Introduction, Methodology,
495 Conclusion.

496

497 Tanzeela Yaqoob Writing empirical interpretations and Literature review.

498

499 **Competing Interests**

500 The authors declare that they have no known competing financial interests or personal
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502

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505 **REFERENCES**

506 Ahmad, M., Jiang, P., Majeed, A., Umar, M., Khan, Z., & Muhammad, M. (2020). The dynamic
507 impact of natural resources, technological innovations and economic growth on ecological
508 footprint: An advanced panel data estimation. *Resources Policy*, 69, 101817

509 Ahmed, Z., Asghar, M. M., Malik, M. N., & Nawaz, K. (2020). Moving towards a sustainable
510 environment: the dynamic linkage between natural resources, human capital, urbanization,
511 economic growth, and ecological footprint in China. *Resources Policy*, 67, 101677.

512 Ahmad, M., Jiang, P., Murshed, M., Shehzad, K., Akram, R., Cui, L., & Khan, Z. (2021).
513 Modelling the dynamic linkages between eco-innovation, urbanization, economic growth
514 and ecological footprints for G7 countries: Does financial globalization
515 matter? *Sustainable Cities and Society*, 70, 102881.

- 516 Ahmed, Z., Zhang, B., & Cary, M. (2021). Linking economic globalization, economic growth,
517 financial development, and ecological footprint: Evidence from symmetric and asymmetric
518 ARDL. *Ecological Indicators*, 121, 107060
- 519 Ahmed, Z., Asghar, M. M., Malik, M. N., & Nawaz, K. (2020). Moving towards a sustainable
520 environment: The dynamic linkage between natural resources, human capital, urbanization,
521 economic growth, and ecological footprint in China. *Resources Policy*. 67, 101677.
- 522 Aydin, M., & Turan, Y, E. (2020). The influence of financial openness, trade openness, and energy
523 intensity on ecological footprint: revisiting the environmental Kuznets curve hypothesis
524 for BRICS countries. *Environmental Science and Pollution Research*. 27, 43233–43245
- 525 Baloch, M. A., Zhang, Iqbal, K, & Iqbal, Z. (2019). The effect of financial development on
526 ecological footprint in BRI countries: evidence from panel data estimation. *Environmental
527 Science and Pollution Research*, 26, 6199–6208
- 528 Chen, Y., Li, J., Lu, H., & Xia, J. (2020). Tradeoffs in water and carbon footprints of shale gas,
529 natural gas, and coal in China. *Fuel*, 263, 116778.
- 530 Chen, Y., & Lee, C, C. (2020). Does technological innovation reduce CO_2 emissions? Cross-
531 country evidence. *Journal of Cleaner Production*, 263, 121550,
- 532 Danish, Baloch, M. A., Mahmood, N., & Zhang, J. W. (2019). Effect of natural resources,
533 renewable energy and economic development on CO_2 emissions in BRICS countries.
534 *Science of the Total Environment*, 678, 632–638.
- 535 Destek, M. A., & Sarkodie, S. A. (2019). Investigation of environmental Kuznets curve for
536 ecological footprint: The role of energy and financial development. *Science of the Total
537 Environment*. 650, 2483–2489,
- 538 Dogan, E., Taspinar, N., & Gokmenoglu, K. K. (2019). Determinants of ecological footprint in
539 MINT countries. *Energy and Environment*, 30(6):1065–1086.
- 540 Fethi, S., & Rahuma, A. (2019). The role of eco-innovation on CO_2 emission reduction in an
541 extended version of the environmental Kuznets curve: evidence from the top 20 refined oil
542 exporting countries. *Environmental Science and Pollution Research*.
- 543 Fethi, S., & Rahuma, A. (2020). The impact of eco-innovation on CO_2 emission reductions:
544 Evidence from selected petroleum companies. *Structural Change and Economic
545 Dynamics*. 53. 108–115
- 546 Godil, D, A., Sharif, A., Rafique, S., & Jermisittiparsert, K. (2020).. The asymmetric effect of tourism,
547 financial development, and globalization on ecological footprint in Turkey. *Environmental
548 Science and Pollution Research*, 27, 40109–40120.

- 549 Hassan,S, T., Xia,E., Khan, N,H., & Shah, S, M,A.(2019). Economic growth, natural resources,
550 and ecological footprints: evidence from Pakistan. *Environmental Science and Pollution*
551 *Research*, 26, 2929–2938
- 552 Khan, M. A., Khan, M. A., Ali, K., Popp, J., & Oláh, J. (2020). Natural Resource Rent and Finance:
553 The Moderation Role of Institutions. *Sustainability*, 12(9), 3897.
- 554 Kongbuamai, N.,Bui,Q., Yousaf,H,M,A., & Liu,Y.(2020). The impact of tourism and natural
555 resources on the ecological footprint: a case study of ASEAN countries. *Environmental*
556 *Science and Pollution Research*, 27, 19251–19264.
- 557 Lantz, V., & Feng, Q. (2006). Assessing income, population, and technology impacts on CO2
558 emissions in Canada: where's the EKC?. *Ecological Economics*, 57(2), 229-238.
- 559 Li, A.H., (2016). Hopes of limiting global warming?. China and the Paris agreement on climate
560 change. *China Perspect.* 49–54
- 561 Lindmark, M. (2002). An EKC-pattern in historical perspective: carbon dioxide emissions,
562 technology, fuel prices and growth in Sweden 1870–1997. *Ecological economics*, 42(1-2),
563 333-347.
- 564 Mallapragada, D. S., Reyes-Bastida, E., Roberto, F., McElroy, E. M., Veskovic, D., & Laurenzi,
565 I. J. (2018). Life cycle greenhouse gas emissions and freshwater consumption of liquefied
566 Marcellus shale gas used for international power generation. *Journal of Cleaner*
567 *Production*, 205, 672-680.
- 568 Panayotou, T. (1993). Empirical tests and policy analysis of environmental degradation at different
569 stages of economic development. *ILO Working Papers*. Retrieved from
570 <http://ideas.repec.org/p/ilo/ilowps/292778.html>
- 571 Przychodzen, J., & Przychodzen, W. (2015). Relationships between eco-innovation and financial
572 performance—evidence from publicly traded companies in Poland and Hungary. *Journal of*
573 *Cleaner Production*, 90, 253-263.
- 574 Sharif, A., Afshan, S., & Qureshi, M. A. (2019). Idolization and ramification between globalization
575 and ecological footprints: evidence from quantile-on-quantile approach. *Environmental*
576 *Science and Pollution Research*, 26(11), 11191-11211.
- 577 Tisdell, C. (2001). Globalisation and sustainability: environmental Kuznets curve and the
578 WTO. *Ecological Economics*, 39(2), 185-196.
- 579 Troster, V., Shahbaz, M., Uddin, G.S. (2018). Renewable energy, oil prices, and economic
580 activity: a Granger-causality in quantiles analysis. *Energy Economics*, 70, 440–452.

- 581 Uddin, G, A., Salahuddin, M., Alam, K., & Gow, J. (2017). Ecological footprint and real income:
582 panel data evidence from the 27 highest emitting countries. *Ecological Indicators*, 77, 166–
583 175.
- 584 Wang,L., Chang,H,L.,Rizvi,S,K,A.,Sari,A. (2020) Are eco-innovation and export diversification
585 mutually exclusive to control carbon emissions in G-7 countries?. *Journal of*
586 *Environmental Management*, 270, 110829.
- 587 Wang,Y., Kang,L., Wu,X., & Xiao,Y.(2013). Estimating the environmental Kuznets curve for
588 Ecological footprint at the global level: A spatial econometric approach. *Ecological*
589 *Indicators*, 34, 15-21.
- 590 Wei, T., Zhu, Q., & Glomsrød, S. (2018). How Will Demographic Characteristics of the Labor
591 Force Matter for the Global Economy and Carbon Dioxide Emissions? *Ecological*
592 *Economics*, 147, 197-207.
- 593 Yasmeen, H., Wang, Y., Zameer, H., & Solangi, Y.A., (2019b). Does oil price volatility influence
594 real sector growth? Empirical evidence from Pakistan. *Energy Reports*, 5, 688–703.
- 595 Zafar,M,W.,, Zaidi,S,A,H.,, Khan,N,R., Mirza,F,M., Hou,F., & Kirmani,S,A,A. (2019). The
596 impact of natural resources, human capital, and foreign direct investment on the ecological
597 footprint: The case of the United States. *Resources Policy*, 63, 101428.
- 598