

Does financial development reinforce ecological footprint in singapore? Evidence from ardl and bayesian analysis

Bui Hoang Ngoc

Ho Chi Minh City Open University

asharawan (✉ asharawan786@hotmail.com)

University of Azad Jammu and Kashmir <https://orcid.org/0000-0002-9922-7795>

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1 **DOES FINANCIAL DEVELOPMENT REINFORCE ECOLOGICAL FOOTPRINT IN**
2 **SINGAPORE? EVIDENCE FROM ARDL AND BAYESIAN ANALYSIS**

3 **Bui Hoang Ngoc ^a & Ashar Awan ^{b,*}**

4 ^a Ho Chi Minh City Open University, Ho Chi Minh City, Vietnam

5 ^b NisanTasi University, Graduate school, Turkey

6 ^b University of Azad Jammu and Kashmir, Muzaffarabad, AJK, Pakistan

7 * Corresponding author

8 **Bui Hoang Ngoc**

9 Finance, Economics and Management Research Group, Ho Chi Minh City Open University,

10 35-37 Ho Hao Hon Street, District 1, Ho Chi Minh City, Vietnam;

11 Email: ngoc.bh@ou.edu.vn

12 ORCID ID: 0000-0002-1892-985X

13 **Ashar Awan**

14 NisanTasi University, Graduate school, Turkey

15 University of Azad Jammu and Kashmir, Pakistan

16 Email: asharawan786@hotmail.com

17

Abstract

Singapore has been ranked in the most dynamic financial market and the highest ecological deficit country, indicating that the trade-off hypothesis may exist. The main goal of the present study is to probe the impact of financial development, economic growth, and human capital on ecological footprint in Singapore from 1980 to 2016. The outcomes obtained from the Autoregressive Distributed Lag (ARDL) method have failed to provide a clear impact of financial sector development on ecological footprint. However, the Bayesian analysis reveals that both financial development and economic growth have a harmful influence on EF, while the impact of human capital is beneficial. A theoretical conclusion derived is that monetary expansion policies should be associated with improving human capital to achieve the United Nations SDGs in the context of Singapore. The findings of the study are of particular interest to policymakers for developing sound policy decisions for sustainable economic progress which is not at the cost of environment.

Keywords: Financial development; Ecological footprint; ARDL; Bayesian analysis; Singapore.

JEL Classification Code: B26, Q01, Q28.

35 **1. Introduction**

36 Climatic change is one of the biggest issues of the 21st century and a topic of overwhelming interest
37 among the research community, policymakers, and professionals in international organizations
38 working for sustainable development (Bayar & Maxim, 2020). Therefore, environmental protection
39 is top priority of countries, particularly signatories of the Paris Agreement (Saud et al., 2020;
40 Usman, Kousar, & Makhdum, 2020). The biggest threat to the environment is the emission of
41 Green House Gases (GHG) which causes an increase in the temperature of the earth and
42 consequently disturbing ecological balance (Ahmed et al., 2019; Baloch et al., 2019). Although,
43 CO₂ emission is the major contributor of GHG and cause of climate change (Bilgili et al., 2021),
44 however, anthropogenic actions particularly, fossil fuel based energy consumption, water waste
45 management and production of fertilizers are causing damage to the ecosystem too. Earlier
46 literature used the ecological footprints as a proxy of pressure on nature by human activities (Al-
47 Mulali & Ozturk, 2015). The ecological footprint is a comprehensive measure of pressure on the
48 ecosystem due to various human activities (Ahmed et al., 2019).

49 While on the other side, keeping a equilibrium between economic growth and environmental
50 damage is a key issue that policymakers are facing. Therefore, it is emphasized to keep the
51 environmental consequences of growth policies (Adedoyin et al., 2021; Destek et al., 2018; Ozturk
52 et al., 2016). In addition, factors that are essential for economic activities may or may not damage
53 the eco-system. While human capital is regarded as an environment-friendly determinant of
54 economic growth (Ahmed & Le, 2021; Danish et al., 2019), financial sector development may
55 (Saud et al., 2020) or may not (Shahbaz et al., 2013; Uddin et al., 2017) be harmful; there is a
56 negative relation between financial sector development and pollution (Hashmi & Alam, 2019;
57 Meirun et al., 2021).

58 An efficient financial system contributes to economic growth, it enables people to buy houses,
59 home appliances and automobiles, however, all this puts pressure on nature by increasing energy
60 demand (Baloch et al., 2019). Furthermore, a financial development boosts investment in new
61 plants and factories and consequently causes more water waste and pollution (Danish et al., 2018).
62 However, financial development is also credited to reduce pollution by boosting investment in
63 research and development of green technologies and energy-efficient machines (Shahbaz et al.,
64 2016). Financial development serves as a strong policy tool for the government to control pressure
65 on the environment. Government can use their influence on financial institutions on their credit
66 provision to less polluting production activities. The financial sector can contribute to
67 environmental protection by discouraging loans to those investment initiatives that produce massive

68 pollution. In addition, investment in a green environmental project, loans to socially responsible
69 firms, and credit to eco-friendly projects curb environmental degradation (Saud et al., 2020).

70 Indisputably, financial sector development – the crucial factor for economic development – is
71 associated with ecological quality through technique, scale and composition effects (Saud et al.,
72 2020). However, findings of previous literature about the influence of financial sector development
73 on the natural environment are mixed. On the one hand, literature reported the negative effect of
74 financial development on EF; for instance, in a panel of 27 countries (Uddin et al., 2017), for China
75 (Destek & Sarkodie, 2019), for Malaysia (Furuoka, 2015); for Nigeria (Omoke et al., 2020). On the
76 other hand other studies reported a positive influence of financial development on EF for instance,
77 (Usman, Kousar, & Makhdum, 2020) for a panel of 20 highest economies, (Mrabet & Alsamara,
78 2017) for Qatar and (Godil et al., 2020) for Turkey.

79 Interestingly, Singapore is a country with rapid economic growth, industrialization and structural
80 change experience (Katircioğlu, 2014; Tan et al., 2014). Though the country is small in size 721
81 km², however, it faces serious challenges of environmental pollution due to its dense population
82 (Han, 2017). While the country has achieved enormous economic growth targets, the government of
83 Singapore is much concerned about negative externalities such as risk to the environment (Ridzuan
84 et al., 2017). Though the literature has been indicating the factors that put pressure on the
85 environment in Singapore, and the government has been introducing regulations, the voices are
86 raised to curb environmental pollution in a way that does not compromise on the economic growth
87 of Singapore (Meirun et al., 2021).

88 Earlier literature on Singapore, examining the influence of financial sector development on EF is
89 mixed. Some studies have found that financial development has increased EF (Destek & Sarkodie,
90 2019) while the others show an opposite result (Naqvi et al., 2020; Saud et al., 2020). These
91 contradictable findings have made a big question about the impact of financial development on EF
92 in Singapore. Therefore, the purpose of this study is to further investigate the impact of financial
93 development, economic growth, and human capital on EF in Singapore. However, our research
94 differs from some previous studies in many ways. Firstly, previous studies have attempted to reveal
95 the relationship between financial development and EF (Destek & Sarkodie, 2019; Naqvi et al.,
96 2020; Saud et al., 2020), but no studies included economic growth and human capital as control
97 variables. Secondly, the impact of financial development on EF in Singapore has seemingly been
98 ambiguous or even contradictable (Destek & Sarkodie, 2019) found positive and; (M. T. I. Khan et
99 al., 2019) insignificant and (Naqvi et al., 2020; Saud et al., 2020) negative. This ambiguity might
100 be the result of the adoption of a frequentist inference, where parameters are unknown but fixed.
101 Therefore, the present study applied two statistical inference types: frequentist inference and

102 Bayesian inference, to provide probability interpretations of uncertainty and various effects of
103 financial development, economic growth, and human capital on EF.

104 The remainder of the study is presented as follows. The next section shows discussions related to
105 relevant literature about the nexus between economic growth, financial development, pollution and
106 human capital. This section is followed by Data and methodology section; after which, the results
107 and findings are discussed in the “Empirical results and discussion” section. Lastly, the conclusion
108 and policy recommendations based on empirical findings are presented in the “Conclusion and
109 recommendations” section of the present study.

110

111 **2. Literature review**

112 **2.1. Financial development and EF**

113 The previous studies showed that there were two effects (technological effect and structural
114 effect) of financial development on the environment, which means financial development may be
115 beneficial or harmful to EF (Du et al., 2012; Saud et al., 2020). On the positive side, financial
116 development boosts a country’s financial structure, brings about financial channels, and attracts
117 FDI, which in turn brings green-environment technology and fosters R&D activities. As a result,
118 global commercials, renewable energy, and technology advancements take place (Ahmed & Le,
119 2021; Hsueh et al., 2013). Thus, financial development can improve environmental quality and
120 decrease EF. On the other side, financial development can cause some scale effects of economy
121 growth process. This point of view suggests that financial development could increase pollution
122 which is caused by a high level of production of the economic-liberalization and higher energy
123 consumption (Pazienza, 2015; Saud et al., 2020). According to (Ha et al., 2020), it is impossible to
124 do economic and household activities without having a harmful influence on the natural ecosystem
125 or environment.

126 Surprisingly, some empirical evidence supports both of these views based on the various
127 development policies in each country and region. For instance, (Uddin et al., 2017) has applied the
128 FMOLS and DOLS methods on the panel data of 27 leading world EF contributors from 1991 to
129 2012, and found that financial development has improved environmental quality by decreasing EF.
130 Similarly, (Ahmed et al., 2019) has researched the connection between financial development and
131 EF in Malaysia from 1971 to 2014. By adopting the Bayer-Hanck cointegration test and ARDL
132 method, the outcomes have verified that financial development mitigates EF. (Omoke et al., 2020)
133 has discovered the negative relationship between financial development and EF in Nigeria from
134 1971-2014. However, some studies have revealed a contrary result regarding the relationship
135 between financial development and EF. (A. Khan et al., 2019) used five Belt and Road initiative

136 (BRI) regions as a research context for the association between financial development and EF. They
137 used the augmented mean group (AMG), and the common correlated effect mean group (CCEMG)
138 approaches, then they found that EF has been fostered by financial development.

139 In the same line, Usman et al. (2020) examined the 20 highest economies from 1995 to 2017.
140 The results showed that financial development deteriorates environmental quality by increasing EF.
141 Godil et al. (2020) has also revealed a similar finding while testing the financial development – EF
142 nexus in Turkey between 1986 and 2018. A comprehensive literature review is presented in Table
143 1.

144

145

[Insert Table 1 here]

146

2.2. Economic growth, human capital, and EF

147 There are evidences that financial development has a mutual relationship with economic
148 growth and human capital (Hsueh et al., 2013). Economic growth (Ahmed, Zafar, et al., 2020; Alola
149 et al., 2019; Usman, Kousar, Yaseen, et al., 2020) and human capital (Ahmed, Asghar, et al., 2020;
150 Ahmed, Zafar, et al., 2020; Pata & Caglar, 2021)) also have influences on EF. Accordingly, the
151 investigation of the interaction between financial development and EF cannot provide a clear
152 understanding without integrating human capital and economic growth.

153 Regarding the influence of economic growth on EF, most previous studies have
154 demonstrated a trade-off between economic growth and EF since the rapid economic development
155 has generated an unprecedented rise in energy demand, especially non-renewable energy (S.
156 Nathaniel & Khan, 2020; Udemba, 2020; Zafar et al., 2019). Furthermore, economic growth could
157 facilitate urban migration and urbanization (Ahmad et al., 2019; S. Nathaniel et al., 2020; Wu et al.,
158 2019), which means that it could certainly bring more pressures to urban infrastructure and
159 ecological assets (Wu et al., 2019). However, there are also some studies suggesting that economic
160 growth would improve EF in Africa and Europe (Usman, Kousar, Yaseen, et al., 2020) or Pakistan
161 (Hassan et al., 2019).

162 Earlier literature also verifies that human capital mitigates environmental degradation,
163 including EF (S. Nathaniel et al., 2020; S. Nathaniel & Khan, 2020; Pata & Caglar, 2021). Some
164 scholars argue that human capital plays a significant role in fostering the adaption of technology
165 change, so it could probably make sustainable growth (Ackah & Kizys, 2015; Consoli et al., 2016).
166 Moreover, human capital generates concerns about environmental problems (Adil, 2018; Asongu,
167 2018; Reynolds et al., 2010; Ulucak & Li, 2020). Nevertheless, some studies, such as Croes et al.
168 (2021) and Ahmed et al. (2021), postulated that the beneficial outcomes of economic growth are
169 insufficiently invested in human capital. As the result, human capital is not giving a significant

170 effect on sustainable development (Dietz et al., 2007). In a recent study, Kassouri and Altintas
171 (2020) indicated that human capital increases EF in MENA countries.

172 Besides, the impact of financial development on EF in Singapore is likely ambiguous or
173 even contradictable (positive; Destek & Sarkodie (2019); insignificant; Khan et al (2019); negative;
174 Naqvi et al., (2020); Saud et al., (2020). In addition, Singapore has been ranked in the most
175 dynamic financial markets in the world, but Singapore has also been listed in the highest ecological
176 deficit countries. It implies that the trade-off hypothesis between financial development and
177 ecological assets may be valid. Therefore, further investigation is necessary to provide probabilistic
178 interpretations of model uncertainty and various influences of financial development, economic
179 growth, and human capital on EF in Singapore.

180

181 **3. Data and methodology**

182 **3.1. Research model and data sources**

183 The role of financial development in economic growth is enormous, but it is undeniable to say
184 that it could be harmful to natural ecosystems. Mutually, a rise in financial development brings
185 about an increase in economic growth, which is directly proportional to EF. Besides, the study
186 incorporated the human capital per person index (labelled, HC) as the control variable. According to
187 Neumayer (2012), it is recognized that the concern on environmental quality in a high human
188 capital country is better than in a low human capital country. Therefore, to assess the impact of
189 financial development, economic growth, and human capital on EF in the case of Singapore, the
190 study has followed the previous works of Baloch et al. (2019); Godil et al. (2020); Pata and Yilanci
191 (2020); Saud et al. (2020); Chen et al. (2019); Usman et al. (2020); Zhao et al. (2019) to propose an
192 initial model, detailed as:

$$193 \quad EF_t = \beta_0 + \beta_1 \cdot \ln FD_t + \beta_2 \cdot \ln GDP_t + \beta_3 \cdot HC_t + u_t \quad (\text{Equation.1})$$

194 where, $\beta_1, \beta_2, \beta_3$ are the long-run coefficients, while t is the time (from 1980 to 2016), and u
195 is the error term. The EF variable is the ecological footprint index (units: gha per capita) collected
196 from the Global Footprint Network. The FD variable is the financial development index (units:
197 point) obtained from the International Monetary Fund (IMF).

198 In the study, we have used the financial development index as a proxy of financial
199 development because it is a financial inclusion index, which is calculated based on the depth,
200 access, and efficiency of financial institutions and financial markets of a country. The GDP variable
201 is the income per capita (at a fixed price of 2010, units: U.S. dollar) abstracted by the World Bank,
202 while the HC variable is the human capital per person index (units: point), quoted by the Federal

203 Reserve Bank of St. Louis. A scale of zero to ten is applied for the human capital per person index,
 204 where zero is the lowest educated economy and 10 is the highest educated economy. In this work,
 205 two variables (FD and GDP) are used by following the logarithm to clarify smooth data, while the
 206 EF and HC variables have original data in use. The descriptive statistics of all variables are
 207 demonstrated in Table 2.

208 **[Insert Table 2 here]**

209 According to Table 2 the mean of Singapore's ecological footprint was 5.843 gha per capita,
 210 which is more than 2.8 gha per capita, the global average. Singapore has been listed in a group of
 211 countries where the ecological deficit is severe. Recently, Singapore has implemented many
 212 positive steps to reduce the ecological deficit situation and has made significant advancements in
 213 renewable energy technology. However, based on the International Energy Agency (IEA, 2018)
 214 data, Singapore has been ranked as 27th out of 142 countries in terms of emissions per capita.
 215 Likewise, the mean of the lnGDP variable was 10.287, while the maximum value of the HC
 216 variable was 3.809. These data showed that Singapore had been a developed education system and
 217 high-income country.

218 **3.2. Methodology**

219 The estimated coefficients of Eq.(1) have only provided the long-run effects of
 220 financial development, economic growth, and human capital on ecological footprint. To analyze the
 221 short-run impacts, the study has applied the autoregressive distributed lag (ARDL) model,
 222 introduced by. The ARDL model has some advantages, such as: firstly, the estimated coefficients
 223 are unbiased and reliable in the case of small sample size; secondly, it could be applied in all three
 224 cases whether the variables are stationary at I(0), I(1), or a mixture of both; thirdly, it provides both
 225 the short and long-run estimated coefficients; and fourthly, it could be used in two cases, where the
 226 cointegration among all variables exists or not (Nkoro & Uko, 2016).

227 Therefore, (Eq.1) is written by the ARDL(p,q) model, as follows:

$$\begin{aligned}
 \Delta EF_t &= \alpha_0 + \beta_1 \cdot EF_{t-1} + \beta_2 \cdot \ln FD_{t-1} + \beta_3 \cdot \ln GDP_{t-1} + \beta_4 \cdot HC_{t-1} + \\
 &+ \sum_{k=1}^{p-1} \alpha_{1k} \cdot \Delta EF_{t-k} + \sum_{k=0}^q \alpha_{2k} \cdot \Delta \ln FD_{t-k} + \sum_{k=0}^q \alpha_{3k} \cdot \Delta \ln GDP_{t-k} + \sum_{k=0}^q \alpha_{4k} \cdot \Delta HC_{t-k} + \varepsilon_t
 \end{aligned}$$

228
 229 (Equation. 2)

230 where: Δ is the first difference

231 $\beta_1, \beta_2, \beta_3, \beta_4$ are coefficients of the long-run impacts

232 $\alpha_1, \alpha_2, \alpha_3, \alpha_4$ are coefficients of the short-run impacts

233 ε_t is the error

234 To achieve the research objectives, our approach is summarized in a four-step process. *First*,
 235 the ordinary least square (OLS) method is applied to Eq.(2) to estimate the long-run coefficients.
 236 *Second*, the cointegration test is used to verify the integration among variables, in which the null
 237 hypothesis is stated: $(H_0 : \beta_1 = \beta_2 = \beta_3 = \beta_4 = 0)$ whilst the alternative hypothesis is written
 238 $(H_1 : \beta_1 \neq \beta_2 \neq \beta_3 \neq \beta_4 \neq 0)$. If the F-statistic value lies below the F-critical values, the null
 239 hypothesis is accepted accordingly. That means there is no cointegration among variables in the
 240 long run. On the other hand, if the F-statistic value exceeds the F-critical value (the upper Bound
 241 value, I(1)), the null hypothesis is denied, and the Eq.(2) would be re-parameterized as an error
 242 correction model (ECM) as:

$$243 \quad \Delta EF_t = \alpha_0 + \lambda \cdot ECM_{t-1} + \sum_{k=1}^{p-1} \alpha_{1k} \cdot \Delta EF_{t-k} + \sum_{k=0}^q \alpha_{2k} \cdot \Delta \ln FD_{t-k} + \sum_{k=0}^q \alpha_{3k} \cdot \Delta \ln GDP_{t-k} + \sum_{k=0}^q \alpha_{4k} \cdot \Delta HC_{t-k} + \varepsilon_t$$

244 (Equation.3)

245 where: p, q are the lag order of each variable collected from the stationary test result. In
 246 Eq.(3), if the estimated coefficient of λ is negative, and belongs $[0;1]$ and significant, which means
 247 that the EF variable can itself re-adjust to long-run equilibrium point after short-run shocks caused
 248 by financial development, economic growth, or human capital.

249 *Third*, the major problem with the frequentist inference (e.g., ARDL model) is that estimated
 250 coefficients are unknown but fixed. More importantly, it is impossible to assess the link between
 251 two variables if the estimated coefficient is insignificant. To solve this case, the Bayesian inference
 252 approach through the integrated Markov chain Monte Carlo sample is applied to provide
 253 probabilistic interpretations of model uncertainty and to differ effects of financial development,
 254 economic growth, and human capital on EF. And in the last process, the modified Wald test
 255 introduced by Toda and Yamamoto (1995) is used to reveal the causal relationship between a pair
 256 of variables in our proposed model. The Toda and Yamamoto procedure is based on the vector
 257 autoregressive model (VAR). The causal relationship between the EF variable and $\ln FD$ variable is
 258 illustrated as follows:

$$259 \quad EF_t = \alpha_0 + \sum_{i=1}^h \alpha_{1i} \cdot EF_{t-i} + \sum_{j=h+1}^p \alpha_{2i} \cdot EF_{t-j} + \sum_{i=1}^h \delta_{1i} \cdot \ln FD_{t-i} + \sum_{j=h+1}^p \delta_{2i} \cdot \ln FD_{t-j} + \mu_{1t}$$

260 (Equation. 4)

$$261 \quad \ln FD_t = \beta_0 + \sum_{i=1}^h \beta_{1i} \cdot \ln FD_{t-i} + \sum_{j=h+1}^p \beta_{2i} \cdot \ln FD_{t-j} + \sum_{i=1}^h \alpha_{1i} \cdot EF_{t-i} + \sum_{j=h+1}^p \alpha_{2i} \cdot EF_{t-j} + \mu_{2t}$$

262 (Equation. 5)

263 where: h is the optimal lag order of the VAR model, and $p = (h+d_{\max})$, with d_{\max} is the
264 maximum lag order. According to Toda and Yamamoto (1995), the maximum of d_{\max} is 2, because
265 if $d_{\max} > 2$, the tests based on F-statistic are not reliable (Davoud et al., 2013; Nkoro & Uko, 2016)

266

267 In Eq.(4), if an estimated coefficient of δ_l is significant, there will be a uni-directional
268 causality running from financial development to ecological footprint. The advantage of the

269 Toda and Yamamoto (1995) procedure is beneficial in minimizing the risk of wrong
270 determination of each variable's lag order and being able to be applied to the variable that is
271 stationary at $I(0)$ or $I(1)$, or a combination of both cointegration and no-cointegration (Sankaran et
272 al., 2019; Toda & Yamamoto, 1995), or either of them.

273 **4. Empirical result and discussion**

274 **4.1. Empirical results**

275 *Unit-root test*

276 Nelson and Plosser (1982) suggested that it is necessary to check the stationary of time-series
277 variable because most of the economic variable is non-stationary. Therefore, to avoid the empirical
278 results being spurious, the three tests are employed, including the ADF test (Dickey & Fuller,
279 1981), the PP test (Phillips & Perron, 1988), and the GLS-ADF test proposed by Elliott et al. (1992)
280 to confirm the stationary of each variable. Compared to the ADF test, the PP test is advantageous in
281 accounting for the potential serial correlation and heteroskedasticity in the residuals. Similarly, the
282 GLS-ADF test has an advantage in allowing the series to be stationary around a linear time trend; or
283 it is to allow the series to be stationary around a possible nonzero mean with no time trend. Results
284 of the stationary tests are presented in Table 3.

285 The stationary test presented in Table 3 provides evidence to reject the null hypothesis of a
286 random walk with drift. More particularly, the $\ln FD$ variable remains stationary at $I(0)$, while three
287 variables (EF, $\ln GDP$ and HC) remain unchanging at $I(1)$ obtaining from three tests. No variable is
288 stationary at $I(2)$. In a brief conclusion, the condition in applying the ARDL model is satisfied
289 accordingly (Nkoro & Uko, 2016; Pesaran et al., 2001). Hence, the cointegration test could proceed
290 for further analysis.

291 **[Insert Table 3 here]**

292 *Cointegration test*

293 The stationary test provided an inconsistent result of lag order between variables (mixture of
294 $I(0)$ and $I(1)$). In the next step, it is necessary to check the long-run association of all variables.

295 Hence, a new cointegration test, Bound-testing, has been employed to verify the cointegration in
296 financial development, economic growth, human capital, and ecological footprint in the case of
297 Singapore. The Bound-testing technique was proposed by Pesaran et al. (2001) with the null
298 hypothesis quoted that H_0 : No cointegration, against the alternative hypothesis H_1 : there is
299 cointegration between examined variables. The result of the Bound-testing is given in Table 4.
300 Accordingly, the F-statistic value (= 4.602) exceeds the F-critical value (= 4.35) at a significant
301 level of 5%. Likewise, the t-critical value (= -3.46) is higher than the t-statistic value (= -3.671) at a
302 significant level of 10%. These results give evidence to reject the null hypothesis, which means that
303 a long-run association between variables in our proposed model exists. Thus, Eq.(2) must be
304 estimated by using the ECM model.

305 **[Insert Table 4 here]**

306 *The short and long-run impacts by the error correction model*

307 Another advantage of the ARDL model is that it could auto-select the optimal lag of each
308 variable. Based on the Akaike Information Criterion (AIC), Schwarz Bayesian information criterion
309 (SBIC), and Hannan-Quinn Information Criterion (HQIC), the empirical result in Table 5 showed
310 that the optimal lag of EF and lnGDP variable is one, and lnFD variable is zero. Simultaneously, the
311 obtained result also indicates that the volatility of current-period human capital is associated with
312 two-period previous human capital.

313 **[Insert Table 5 here]**

314 After confirmation of cointegration, the error correction model is applied to study the impact
315 of financial development, economic growth, and human capital on EF both in the short and long-
316 run. The optimal lag of each variable is set to select the final ARDL specification. Applying to the
317 data of Singapore, the best ARDL model is the ARDL(1,1,0,0). The coefficients of the short and
318 long-run impacts are shown in Table 6.

319 **[Insert Table 6 here]**

320 The empirical result showed in Table 6 indicates that the coefficient of $\text{CoinEq}(-1)$ is
321 negative and significant level at 1% ($\lambda = -0.5647$, p-value = 0.001). This finding supports that
322 ecological footprint can re-adjust itself to the long-run equilibrium point after the short-run is
323 suddenly affected by financial development, or economic growth, or human capital. Two years is
324 the time needed for an adjustment ($= 1/|\lambda|$). Table 6 also indicates that economic growth has a
325 positive and significant impact on EF in the short and long-run. More specifically, an 1% rise in
326 economic growth leads to a 4.39 gha per capita increase in the short-run ecological footprint and a
327 7.78 gha per capita increase in the long-run ecological footprint. Hence, our analysis reveals that the

328 long-run impact of economic growth on EF is greater than the short-run impact. Similarly, the
329 estimated coefficient of the HC variable is -2.10 in the short-run and is -3.72 in the long-run,
330 respectively. These results imply that EF is affected by human capital in the case of Singapore.
331 However, contrary to the influence of economic growth, an increase in the human capital per person
332 induces a fall in ecological footprint. We believe that these fascinating findings could provide
333 Singapore policymakers more insights into sustainable development strategies. More details about
334 these findings will be discussed in the next section.

335 Nevertheless, the main aim of our research is to study the influence of financial
336 development on EF in Singapore, which means that whether there is a trade-off between economic
337 growth and environmental destruction. Unfortunately, the empirical outcome reveals that financial
338 development has a positive impact on EF, but not yet significant. In the view of frequentist
339 inference, the obtained outcome from the ECM approach has failed to demonstrate the influence of
340 financial development on EF in the case of Singapore. Besides, the major diagnostic tests were used
341 to confirm the above conclusions, including the heteroskedasticity test, the autocorrelation test, the
342 distribution of residuals, and the functional form test. The results of diagnostic tests are given in the
343 lowest in Table 6. Accordingly, the four tests have a p-value that is higher than 0.05, and it is
344 evident to reject the null hypothesis. Besides, the work has conducted the cumulative sum of
345 recursive residuals (CUSUM) and the cumulative sum of squares of recursive residuals
346 (CUSUMSQ) test to check the stability of the long-run coefficients and the short-run dynamics.
347 Both CUSUM and CUSUMSQ lie within the Bound-critical value at the level of 5% significance
348 (see Figure 1a, 1b). When all diagnostic tests are satisfied, it is possible to conclude that our
349 proposed model is stable and the obtained coefficients by the ECM approach are reliable.

350 **[Insert Figure 1a, and Figure 1b here]**

351 As the result, the failure in defining the effect of financial development on EF leads to
352 difficulty makes some difficulties in suggesting efficient environmental protection policies.
353 Therefore, the Bayesian analysis for the generalized linear model (GLM) was employed to re-
354 examine the influence of financial development on EF in the context of Singapore.

355 *The empirical results by Bayesian inference*

356 Contrary to frequentist inference (i.e., where observed are assumed to be random and
357 estimation parameters are unknown but have fixed quantities), the Bayesian inference assumes that
358 the observed data is fixed, and estimation parameters are random (Bernardo & Smith, 1994). The
359 Bayesian analysis is based on the Bayes's rule and the posterior distribution results from using the
360 prior information about model parameters with evidence from the observed data. An advantage of

361 Bayesian analysis is that the Bayesian paradigm allows us to prove some probability statements, as
362 a variable is likely or unlikely to impact on another, or the true value of a parameter falls into a
363 certain interval with a pre-specified probability (Bernardo & Smith, 1994; S.K.Thompson, 2012) .

364 The specification of the Bayesian GLM regression is expressed as:

$$365 \quad y_t : N(\beta^T X_t, \delta^2 I) \quad \text{(Equation. 6)}$$

366 where, y_t is the ecological footprint drawn from normal Gaussian distribution, and X_t is the
367 matrix of the independent variables. β^T denotes the transposed weight matrix, while δ^2 is the
368 variance and I is the identity matrix, to give the model a multi-dimensional formulation. Generally,
369 the prior distribution is defined as pre-existing information about model parameters and is often
370 derived from theoretical or expert knowledge. Fortunately, Lemoine (2019) suggested that the
371 estimated coefficients obtained from the OLS approach could be used in the case of weakly
372 informative priors in Bayesian analysis. By this brilliant suggestion, the study adopts the estimated
373 coefficient of variables from the OLS estimator to set the initial information of the Bayesian GLM
374 model with assuming a normal distribution.

375 In the Bayesian GLM model, the posterior distribution of the estimated model parameters is
376 generated from a probability distribution based on the prior information, observed data, and the
377 outputs. The posterior distribution can be calculated as the following equation:

$$378 \quad P(\beta | y_t, X_t) = \frac{P(y_t | \beta, X_t) * P(\beta | X_t)}{P(y_t | X_t)} \quad \text{(Equation.7)}$$

379 where, $P(y_t | \beta, X_t)$ is the likelihood of the data, and $P(\beta | X_t)$ denotes the prior probability
380 information of the model parameters, while $P(y_t | X_t)$ represents the normalization constant. In this
381 work, the adaptive random-walk Metropolis-Hastings algorithm was used to avoid the spurious
382 convergence and provide probabilistic interpretations of model uncertainty and varying effects of
383 financial development, economic growth, and human capital on EF. The result of Bayesian analysis
384 is presented in Table 7.

385 **[Insert Table 7 here]**

386 The empirical outcome given in Table 7 reveals a positive effect of financial development and
387 economic growth on EF, while the impact of human capital is negative. These results are in line
388 with the coefficients obtained by the ECM model. More importantly, the linkage between financial
389 development and ecological footprint has been established. Additionally, the acceptance rate of 0.37
390 is larger than the optimal acceptance rate of 0.234 (Gelman et al., 1997), whereas standard deviation
391 values of the parameters are small, and the Monte Carlo chain standard errors (MCSE) are close to
392 one decimal. Besides, the CUSUM plots of the parameters are jagged, not smooth, and crossing the

393 X-axis (see Figure 2). These obtained outcomes provide evidence to accept the Metropolis-Hastings
394 algorithm and confirm the high accuracy of the parameter estimates (Cowles & Carlin, 1996). Thus,
395 Bayesian inference is valid.

396 **[Insert Figure 2 here]**

397 Contrary to frequentist inference, in Bayesian inference, 95% credible intervals indicate
398 which range the true value of a certain parameter belongs to. For example, the mean value of the
399 financial development variable (lnFD) lies in an interval between -0.1329 and 0.3213 with a 95%
400 probability. As expected, given probability, we may state that financial development has a strongly
401 positive effect on ecological footprint with a 79% probability (Block et al., 2011). The outcome is
402 justified by the historical trend in money supply strategies and the serious ecological deficit
403 situation in Singapore.

404 *The Granger causality test*

405 The ECM or GLM approach does not guide us about the causal relationship among the
406 variables. In the final step, the Toda and Yamamoto (1995) procedure is applied to check the
407 Granger causality between numbers of pairs of variables. The equation is used to test has been
408 shown in Section 3, and the empirical result is given in Table 8 and Figure 3. According to Table 8,
409 there is bi-directional causality between either economic growth (lnGDP) or human capital (HC)
410 and ecological footprint (EF). At the same time, Figure 3 also indicated that (there if) uni-
411 directional causality running lnGDP or EF or HC to financial development (lnFD), and running
412 from HC to lnGDP.

413 **[Insert Table 8 here]**

414 **[Insert Figure 3 here]**

415 **4.2. Discussion**

416 In frequentist inference, the obtained results demonstrate that the impact of financial
417 development exerts an ambiguous effect in both the short-run and long-run, which is similar to the
418 finding of (M. T. I. Khan et al., 2019). However, thanks to the Bayesian inference, the obtained
419 outcomes have found a positive strongly influence of financial development on EF. With a 79%
420 probability, the study may conclude that an expansion financial development policy may be harmful
421 to environmental quality in the context of Singapore. The finding is inconsistent with Saud et al
422 (2020) and Naqvi et al (2020), who validate the negative impact of financial development on EF,
423 while it is in line with Destek and Sarkodie (2019), who find that financial development increases
424 EF in Singapore. More importantly, by applying Bayesian analysis, this work overcame the
425 ambiguity found in a previous study (M. T. I. Khan et al., 2019). Thus, the study strongly confirmed

426 that financial development plays a positive role in ecological footprint in Singapore. More broadly,
427 the finding is consistent with many studies in various contexts, such as the research of (A. Khan et
428 al., 2019) in the context of five Belt and Road initiative (BRI) regions, or (Godil et al., 2020) in
429 Turkey. Nevertheless, it should be noted that some studies show contradicted results, such as
430 (Uddin et al., 2017) on 27 leading world EF contributors, (Ahmed et al., 2019) on Malaysia, or
431 (Omoke et al., 2020) on Nigeria (see also Table 1). In the context of Singapore, the study believes
432 that financial development causes the scale effect by fostering economic liberalization and attract
433 foreign direct investment (Pazienza, 2015; Saud et al., 2020). It means that an increase in financial
434 development in Singapore leads to a rise in economic growth, high manufacture and energy
435 consumption, especially non-renewable energy consumption due to economic-liberalization (Destek
436 & Sarkodie, 2019). This argument is consistent with some recent studies that ASEAN countries,
437 including Singapore where non-renewableenergy has been largely used in many industries
438 (Kongbuamai et al., 2020; S. Nathaniel & Khan, 2020). Consequently, the financial development is
439 both side-effects on) the increase of CO₂ emissions, air and soil pollution, and on ultimate raise
440 level of EF (Pazienza, 2015; Saud et al., 2020). As a result, the Singapore Government should
441 considerably improve current policies regarding to financial development.

442 Both the ARDL and Bayesian analyses confirm that economic growth has a positive impact
443 on EF (probability of 100%). The finding is consistent with the trade-off theory and the empirical
444 evidence between economic growth and EF since economic growth has increased energy (S.
445 Nathaniel & Khan, 2020; Udemba, 2020; Zafar et al., 2019), facilitation of urbanization (Ahmed et
446 al., 2019; S. Nathaniel et al., 2020; Wu et al., 2019), and pressure on infrastructure and ecological
447 assets(Sharma et al., 2020). Consequently, economic growth may break the balance in the
448 biodiversity and increases EF, supported by the hypothesis of scale effect. As mentioned above, the
449 increase in economic growth has generated a high level of non-renewable energy consumption in
450 ASEAN countries, including Singapore (Kongbuamai et al., 2020; S. Nathaniel & Khan, 2020). To
451 maintain environmental quality, therefore, more effective economic growth strategies should be
452 implemented.

453 Finally, human capital mitigates EF in both the short and long-run. The results are firmly
454 validated by both the ARDL and Bayesian analyses with the probability of 100%. The finding is
455 justifiable since some scholars (for instance Nathaniel, 2020; Nathaniel et al., 2021; Pata and
456 Caglar, 2021), who have discussed and empirically confirmed that human capital decreases EF. The
457 results show that an increase in human capital leads to a corresponding rise in technological change
458 adoption (Ackah & Kizys, 2015; Consoli et al., 2016) as well as the awareness towards
459 environmental quality problems(Adil, 2018; Asongu, 2018; Reynolds et al., 2010; Ulucak & Li,

460 2020); and thus, activities of environmental destruction should be reduced and sustainable
461 development could be more promoted in Singapore. The findings also demonstrate the consistency
462 with other contexts that emphasize the importance of human capital and have appropriate policies to
463 foster human capital, such as the United States (Zafar et al., 2019), G7 countries (Ahmed, Asghar,
464 et al., 2020; S. P. Nathaniel, 2020), Central and Eastern European Countries (Chen et al., 2019),
465 among the others. The results likely prove how human capital is important and beneficial in the
466 context of Singapore.

467 **5. Conclusion and policy implications**

468 **5.1. Conclusions**

469 We aim to uncover the true nature of the link between financial development and EF by
470 incorporating economic growth and human capital as control variables. To achieve this purpose, the
471 study used the ARDL approach to provide preliminary results, and then a Bayesian analysis was
472 employed to provide more insightful outcomes. As expected, the Bayesian analysis enables us to
473 firmly conclude that both financial development and economic growth have a positive impact on
474 EF, while human capital decreases EF. These findings indicated that the trade-off hypothesis
475 between financial development and ecological assets is valid. Hence, monetary expansion policies
476 should be associated with improving human capital to achieve sustainable development goals in the
477 context of Singapore.

478 The findings are undisputable in confirming that the current financial development policies of
479 Singapore may threaten environmental quality. They should be better assimilated and used more
480 effectively in policy instruments regarding financial development. As noted by the previous studies,
481 financial development may improve environmental quality by boosting research and development
482 and technological advancement (Ahmed et al., 2019). Thus, the Singapore Government should
483 emphasize the vital role of the financial sector in providing funds for an adaption to use eco-friendly
484 technologies in new ventures and also to in the existing businesses as a replacement of outdated
485 technologies (Ahmed et al., 2019; Usman, Kousar, & Makhdum, 2020). Furthermore, the study also
486 recommends that Singapore Government should facilitate the optimal utilization of energy by
487 boosting financial support for eco-friendly projects at a minimum interest rate (Usman, Kousar, &
488 Makhdum, 2020). Finally, additional funds should be allocated to support environmental policies as
489 well as to raise the environmental awareness of residents (Pata & Yilanci, 2020).

490 As economic growth increases EF in Singapore, the study suggests that the economic
491 development policies should be focused on the decrease of non-renewable energy (S. Nathaniel &
492 Khan, 2020; Udemba, 2020; Zafar et al., 2019) and maintenance of balance between urbanization

493 and the demand of ecological assets (Ahmed et al., 2019; S. Nathaniel et al., 2020; Wu et al., 2019).
494 Thus, fostering the use of renewable and decreasing the use of non-renewable energy is utterly
495 important. This can be implemented by using environmental taxes and subsidies for the removal of
496 pollution technology. To balance the urbanization and biodiversity relationship, the establishment
497 of smart cities seems to be the key solution (S. P. Nathaniel et al., 2021). Furthermore, the policies
498 of financial development and economic growth in the context of Singapore should be integrated
499 seamlessly to effectively foster environmental quality. More specifically, financial development
500 should allocate more resources to developing and applying cleaner and more eco-friendly
501 technologies. Novel technology is the key to remove “dirty” and obsolete technologies while
502 fostering advanced and smart technologies in manufacturing and living.

503 Finally, human capital is an important factor of EF fall in Singapore. Thus, Singapore
504 Government should generate a long and healthy living with a higher level of education and
505 welfare/good living standards for residents (Türe & Türe, 2021), since those are determinants of
506 sustainable growth and development (Ackah & Kizys, 2015; Consoli et al., 2016). Furthermore,
507 policies to improve environmental problems awareness are also significant in improving
508 environmental quality (Adil, 2018; Asongu, 2018; Reynolds et al., 2010; Ulucak & Li, 2020).
509 Similarly, the study should emphasize the urgency of the integration between financial
510 development, economic growth and human capital policies since previous studies (Ahmed et al.,
511 2021; Croes et al., 2021) have suggested insufficient resources resulting from financial development
512 and economic growth in human capital lead to the role of this variable fell in improving
513 biocapacity, and environmental quality, see also (Dietz et al., 2007). Furthermore, human capital
514 development can be considered a base for screening and selecting appropriate imported
515 technologies and R&D (S. Nathaniel et al., 2020; Pata & Caglar, 2021) . Therefore, the work
516 suggests that the Singapore Government should have a comprehensive vision about why financial
517 development, economic growth and human capital influence on ecological footprint, and provided
518 an integrated yet effective way to improve biocapacity, and environmental quality.

519

520

521 **References**

522

523 Ackah, I., & Kizys, R. (2015). Green growth in oil producing African countries: A panel data
524 analysis of renewable energy demand. *Renewable and Sustainable Energy Reviews*, 50, 1157–
525 1166.

526 Adedoyin, F. F., Nathaniel, S., & Adeleye, N. (2021). An investigation into the anthropogenic
527 nexus among consumption of energy, tourism, and economic growth: do economic policy
528 uncertainties matter? *Environmental Science and Pollution Research*, 28(3), 2835–2847.

529 Adil, A. A. (2018). Could Human Development Be the Key to Environmental Sustainability?
530 *Leonardo*, 51(2), 197–198.

531 Ahmad, M., Zhao, Z.-Y., & Li, H. (2019). Revealing stylized empirical interactions among
532 construction sector, urbanization, energy consumption, economic growth and CO2 emissions
533 in China. *Science of the Total Environment*, 657, 1085–1098.

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

537 Ahmed, Z., & Le, H. P. (2021). Linking Information Communication Technology, trade
538 globalization index, and CO 2 emissions: evidence from advanced panel techniques.
539 *Environmental Science and Pollution Research*, 28(7), 8770–8781.

540 Ahmed, Z., Wang, Z., Mahmood, F., Hafeez, M., & Ali, N. (2019). Does globalization increase the
541 ecological footprint? Empirical evidence from Malaysia. *Environmental Science and Pollution
542 Research*, 26(18), 18565–18582.

543 Ahmed, Z., Zafar, M. W., & Ali, S. (2020). Linking urbanization, human capital, and the ecological
544 footprint in G7 countries: an empirical analysis. *Sustainable Cities and Society*, 55, 102064.

545 Ahmed, Z., Zhang, B., & Cary, M. (2021). Linking economic globalization, economic growth,
546 financial development, and ecological footprint: Evidence from symmetric and asymmetric
547 ARDL. *Ecological Indicators*, 121, 107060.

548 Al-Mulali, U., & Ozturk, I. (2015). The effect of energy consumption, urbanization, trade openness,
549 industrial output, and the political stability on the environmental degradation in the MENA
550 (Middle East and North African) region. *Energy*, 84, 382–389.

551 Alola, A. A., Bekun, F. V., & Sarkodie, S. A. (2019). Dynamic impact of trade policy, economic
552 growth, fertility rate, renewable and non-renewable energy consumption on ecological
553 footprint in Europe. *Science of the Total Environment*, 685, 702–709.

- 554 Asongu, S. A. (2018). CO 2 emission thresholds for inclusive human development in sub-Saharan
555 Africa. *Environmental Science and Pollution Research*, 25(26), 26005–26019.
- 556 Baloch, M. A., Zhang, J., Iqbal, K., & Iqbal, Z. (2019). The effect of financial development on
557 ecological footprint in BRI countries: evidence from panel data estimation. *Environmental*
558 *Science and Pollution Research*, 26(6), 6199–6208.
- 559 Bayar, Y., & Maxim, A. (2020). Financial development and CO2 emissions in post-transition
560 European Union countries. *Sustainability*, 12(7), 2640.
- 561 Bernardo, J. M., & Smith, A. F. (1994). Bayesian theory: John Wiley and Sons. *New York*.
- 562 Bilgili, F., Kuşkaya, S., Khan, M., Awan, A., & Türker, O. (2021). The roles of economic growth
563 and health expenditure on CO 2 emissions in selected Asian countries: a quantile regression
564 model approach. *Environmental Science and Pollution Research*, 1–24.
- 565 Block, J. H., Jaskiewicz, P., & Miller, D. (2011). Ownership versus management effects on
566 performance in family and founder companies: A Bayesian reconciliation. *Journal of Family*
567 *Business Strategy*, 2(4), 232–245.
- 568 Chen, S., Saud, S., Saleem, N., & Bari, M. W. (2019). Nexus between financial development,
569 energy consumption, income level, and ecological footprint in CEE countries: do human
570 capital and biocapacity matter? *Environmental Science and Pollution Research*, 26(31),
571 31856–31872.
- 572 Consoli, D., Marin, G., Marzucchi, A., & Vona, F. (2016). Do green jobs differ from non-green jobs
573 in terms of skills and human capital? *Research Policy*, 45(5), 1046–1060.
- 574 Cowles, M. K., & Carlin, B. P. (1996). Markov chain Monte Carlo convergence diagnostics: a
575 comparative review. *Journal of the American Statistical Association*, 91(434), 883–904.
- 576 Croes, R., Ridderstaat, J., Bąk, M., & Zientara, P. (2021). Tourism specialization, economic growth,
577 human development and transition economies: The case of Poland. *Tourism Management*, 82,
578 104181.
- 579 Danish, Hassan, S. T., Baloch, M. A., Mahmood, N., & Zhang, J. (2019). Linking economic growth
580 and ecological footprint through human capital and biocapacity. *Sustainable Cities and*
581 *Society*, 47, 101516. <https://doi.org/https://doi.org/10.1016/j.scs.2019.101516>
- 582 Danish, Wang, B., & Wang, Z. (2018). Imported technology and CO2 emission in China:
583 Collecting evidence through bound testing and VECM approach. *Renewable and Sustainable*
584 *Energy Reviews*, 82, 4204–4214. <https://doi.org/https://doi.org/10.1016/j.rser.2017.11.002>
- 585 Davoud, M., Behrouz, S. A., Farshid, P., & Somayeh, J. (2013). Oil products consumption,
586 electricity consumption-economic growth nexus in the economy of Iran: A bounds test co-
587 integration approach. *International Journal of Academic Research in Business and Social*

- 588 *Sciences*, 3(1), 353–367.
- 589 Destek, M. A., & Sarkodie, S. A. (2019). Investigation of environmental Kuznets curve for
590 ecological footprint: the role of energy and financial development. *Science of the Total*
591 *Environment*, 650, 2483–2489.
- 592 Destek, M. A., Ulucak, R., & Dogan, E. (2018). Analyzing the environmental Kuznets curve for the
593 EU countries: the role of ecological footprint. *Environmental Science and Pollution Research*,
594 25(29), 29387–29396.
- 595 Dickey, D. A., & Fuller, W. A. (1981). Likelihood ratio statistics for autoregressive time series with
596 a unit root. *Econometrica: Journal of the Econometric Society*, 1057–1072.
- 597 Dietz, T., Rosa, E. A., & York, R. (2007). Driving the human ecological footprint. *Frontiers in*
598 *Ecology and the Environment*, 5(1), 13–18.
- 599 Du, L., Wei, C., & Cai, S. (2012). Economic development and carbon dioxide emissions in China:
600 Provincial panel data analysis. *China Economic Review*, 23(2), 371–384.
- 601 Elliott, G., Rothenberg, T. J., & Stock, J. H. (1992). *Efficient tests for an autoregressive unit root*.
602 National Bureau of Economic Research Cambridge, Mass., USA.
- 603 Furuoka, F. (2015). Financial development and energy consumption: Evidence from a
604 heterogeneous panel of Asian countries. *Renewable and Sustainable Energy Reviews*, 52, 430–
605 444.
- 606 Gelman, A., Gilks, W. R., & Roberts, G. O. (1997). Weak convergence and optimal scaling of
607 random walk Metropolis algorithms. *The Annals of Applied Probability*, 7(1), 110–120.
- 608 Godil, D. I., Sharif, A., Rafique, S., & Jermittiparsert, K. (2020). The asymmetric effect of
609 tourism, financial development, and globalization on ecological footprint in Turkey.
610 *Environmental Science and Pollution Research*, 27(32), 40109–40120.
- 611 Ha, N. M., Ngoc, B. H., & McAleer, M. (2020). Financial Integration, Energy Consumption And
612 Economic Growth In Vietnam. *Annals of Financial Economics*, 15(03), 2050010.
- 613 Han, H. (2017). Singapore, a garden city: Authoritarian environmentalism in a developmental state.
614 *The Journal of Environment & Development*, 26(1), 3–24.
- 615 Hashmi, R., & Alam, K. (2019). Dynamic relationship among environmental regulation, innovation,
616 CO2 emissions, population, and economic growth in OECD countries: A panel investigation.
617 *Journal of Cleaner Production*, 231, 1100–1109.
- 618 Hassan, S. T., Xia, E., Khan, N. H., & Shah, S. M. A. (2019). Economic growth, natural resources,
619 and ecological footprints: evidence from Pakistan. *Environmental Science and Pollution*
620 *Research*, 26(3), 2929–2938.
- 621 Hsueh, S.-J., Hu, Y.-H., & Tu, C.-H. (2013). Economic growth and financial development in Asian

622 countries: A bootstrap panel Granger causality analysis. *Economic Modelling*, 32, 294–301.

623 Kassouri, Y., & Altıntaş, H. (2020). Human well-being versus ecological footprint in MENA
624 countries: A trade-off? *Journal of Environmental Management*, 263, 110405.

625 Katircioğlu, S. T. (2014). Testing the tourism-induced EKC hypothesis: The case of Singapore.
626 *Economic Modelling*, 41, 383–391.
627 <https://doi.org/https://doi.org/10.1016/j.econmod.2014.05.028>

628 Khan, A., Chenggang, Y., Hussain, J., & Bano, S. (2019). Does energy consumption, financial
629 development, and investment contribute to ecological footprints in BRI regions?
630 *Environmental Science and Pollution Research*, 26(36), 36952–36966.

631 Khan, M. T. I., Yaseen, M. R., & Ali, Q. (2019). Nexus between financial development, tourism,
632 renewable energy, and greenhouse gas emission in high-income countries: a continent-wise
633 analysis. *Energy Economics*, 83, 293–310.

634 Kongbuamai, N., Bui, Q., Yousaf, H. M. A. U., & Liu, Y. (2020). The impact of tourism and
635 natural resources on the ecological footprint: a case study of ASEAN countries. *Environmental
636 Science and Pollution Research*, 27(16), 19251–19264.

637 Lemoine, N. P. (2019). Moving beyond noninformative priors: why and how to choose weakly
638 informative priors in Bayesian analyses. *Oikos*, 128(7), 912–928.

639 Meirun, T., Mihardjo, L. W. W., Haseeb, M., Khan, S. A. R., & Jermsittiparsert, K. (2021). The
640 dynamics effect of green technology innovation on economic growth and CO 2 emission in
641 Singapore: New evidence from bootstrap ARDL approach. *Environmental Science and
642 Pollution Research*, 28(4), 4184–4194.

643 Mrabet, Z., & Alsamara, M. (2017). Testing the Kuznets Curve hypothesis for Qatar: A comparison
644 between carbon dioxide and ecological footprint. *Renewable and Sustainable Energy Reviews*,
645 70, 1366–1375.

646 Naqvi, S. A. A., Shah, S. A. R., & Mehdi, M. A. (2020). Revealing empirical association among
647 ecological footprints, renewable energy consumption, real income, and financial development:
648 a global perspective. *Environmental Science and Pollution Research*, 27(34), 42830–42849.

649 Nathaniel, S., Anyanwu, O., & Shah, M. (2020). Renewable energy, urbanization, and ecological
650 footprint in the Middle East and North Africa region. *Environmental Science and Pollution
651 Research*, 1–13.

652 Nathaniel, S., & Khan, S. A. R. (2020). The nexus between urbanization, renewable energy, trade,
653 and ecological footprint in ASEAN countries. *Journal of Cleaner Production*, 272, 122709.

654 Nathaniel, S. P. (2020). Biocapacity, human capital, and ecological footprint in G7 countries: the
655 moderating role of urbanization and necessary lessons for emerging economies. *Energy*,

- 656 *Ecology and Environment*, 1–16.
- 657 Nathaniel, S. P., Nwulu, N., & Bekun, F. (2021). Natural resource, globalization, urbanization,
658 human capital, and environmental degradation in Latin American and Caribbean countries.
659 *Environmental Science and Pollution Research*, 28(5), 6207–6221.
- 660 Nelson, C. R., & Plosser, C. R. (1982). Trends and random walks in macro-economic time series,
661 *Journal of Monetary Economics*. September, 10(2), 139–162.
- 662 Neumayer, E. (2012). Human development and sustainability. *Journal of Human Development and*
663 *Capabilities*, 13(4), 561–579.
- 664 Nkoro, E., & Uko, A. K. (2016). Autoregressive Distributed Lag (ARDL) cointegration technique:
665 application and interpretation. *Journal of Statistical and Econometric Methods*, 5(4), 63–91.
- 666 Omoke, P. C., Nwani, C., Effiong, E. L., Evbuomwan, O. O., & Emenekwe, C. C. (2020). The
667 impact of financial development on carbon, non-carbon, and total ecological footprint in
668 Nigeria: new evidence from asymmetric dynamic analysis. *Environmental Science and*
669 *Pollution Research*, 27(17), 21628–21646.
- 670 Ozturk, I., Al-Mulali, U., & Saboori, B. (2016). Investigating the environmental Kuznets curve
671 hypothesis: the role of tourism and ecological footprint. *Environmental Science and Pollution*
672 *Research*, 23(2), 1916–1928.
- 673 Pata, U. K., & Caglar, A. E. (2021). Investigating the EKC hypothesis with renewable energy
674 consumption, human capital, globalization and trade openness for China: Evidence from
675 augmented ARDL approach with a structural break. *Energy*, 216, 119220.
676 <https://doi.org/https://doi.org/10.1016/j.energy.2020.119220>
- 677 Pata, U. K., & Yilanci, V. (2020). Financial development, globalization and ecological footprint in
678 G7: further evidence from threshold cointegration and fractional frequency causality tests.
679 *Environmental and Ecological Statistics*, 27(4), 803–825.
- 680 Pazienza, P. (2015). The relationship between CO2 and Foreign Direct Investment in the agriculture
681 and fishing sector of OECD countries: Evidence and policy considerations. *Intelektinė*
682 *Ekonomika*, 9(1), 55–66.
- 683 Pesaran, M. H., Shin, Y., & Smith, R. J. (2001). Bounds testing approaches to the analysis of level
684 relationships. *Journal of Applied Econometrics*, 16(3), 289–326.
- 685 Phillips, P. C. B., & Perron, P. (1988). Testing for a unit root in time series regression. *Biometrika*,
686 75(2), 335–346.
- 687 Reynolds, T. W., Farley, J., & Huber, C. (2010). Investing in human and natural capital: An
688 alternative paradigm for sustainable development in Awassa, Ethiopia. *Ecological Economics*,
689 69(11), 2140–2150.

- 690 Ridzuan, A. R., Ismail, N. A., & Che Hamat, A. F. (2017). Does foreign direct investment
691 successfully lead to sustainable development in Singapore? *Economies*, 5(3), 29.
- 692 S.K.Thompson. (2012). *Sampling (3rd ed.)* (3rd ed.). Hoboken, N.J. : John Wiley & Sons. Inc.
- 693 Sankaran, A., Kumar, S., Arjun, K., & Das, M. (2019). Estimating the causal relationship between
694 electricity consumption and industrial output: ARDL bounds and Toda-Yamamoto approaches
695 for ten late industrialized countries. *Heliyon*, 5(6), e01904.
- 696 Saud, S., Chen, S., & Haseeb, A. (2020). The role of financial development and globalization in the
697 environment: accounting ecological footprint indicators for selected one-belt-one-road
698 initiative countries. *Journal of Cleaner Production*, 250, 119518.
- 699 Shahbaz, M., Jam, F. A., Bibi, S., & Loganathan, N. (2016). Multivariate Granger causality
700 between CO2 emissions, energy intensity and economic growth in Portugal: evidence from
701 cointegration and causality analysis. *Technological and Economic Development of Economy*,
702 22(1), 47–74.
- 703 Shahbaz, M., Solarin, S. A., Mahmood, H., & Arouri, M. (2013). Does financial development
704 reduce CO2 emissions in Malaysian economy? A time series analysis. *Economic Modelling*,
705 35, 145–152.
- 706 Sharma, R., Sinha, A., & Kautish, P. (2020). Examining the impacts of economic and demographic
707 aspects on the ecological footprint in South and Southeast Asian countries. *Environmental*
708 *Science and Pollution Research*, 27(29), 36970–36982.
- 709 Tan, F., Lean, H. H., & Khan, H. (2014). Growth and environmental quality in Singapore: Is there
710 any trade-off? *Ecological Indicators*, 47, 149–155.
711 <https://doi.org/https://doi.org/10.1016/j.ecolind.2014.04.035>
- 712 Toda, H. Y., & Yamamoto, T. (1995). Statistical inference in vector autoregressions with possibly
713 integrated processes. *Journal of Econometrics*, 66(1–2), 225–250.
- 714 Türe, C., & Türe, Y. (2021). A model for the sustainability assessment based on the human
715 development index in districts of Megacity Istanbul (Turkey). *Environment, Development and*
716 *Sustainability*, 23(3), 3623–3637.
- 717 Uddin, G. A., Salahuddin, M., Alam, K., & Gow, J. (2017). Ecological footprint and real income:
718 panel data evidence from the 27 highest emitting countries. *Ecological Indicators*, 77, 166–
719 175.
- 720 Udemba, E. N. (2020). A sustainable study of economic growth and development amidst ecological
721 footprint: New insight from Nigerian Perspective. *Science of the Total Environment*, 732,
722 139270.
- 723 Ulucak, R., & Li, N. (2020). The nexus between economic globalization and human development in

724 Asian countries: an empirical investigation. *Environmental Science and Pollution Research*,
725 27(3), 2622–2629.

726 Usman, M., Kousar, R., & Makhdum, M. S. A. (2020). The role of financial development, tourism,
727 and energy utilization in environmental deficit: evidence from 20 highest emitting economies.
728 *Environmental Science and Pollution Research*, 27(34), 42980–42995.

729 Usman, M., Kousar, R., Yaseen, M. R., & Makhdum, M. S. A. (2020). An empirical nexus between
730 economic growth, energy utilization, trade policy, and ecological footprint: a continent-wise
731 comparison in upper-middle-income countries. *Environmental Science and Pollution*
732 *Research*, 27(31), 38995–39018.

733 Wu, Y., Shen, L., Zhang, Y., Shuai, C., Yan, H., Lou, Y., & Ye, G. (2019). A new panel for
734 analyzing the impact factors on carbon emission: A regional perspective in China. *Ecological*
735 *Indicators*, 97, 260–268.

736 Zafar, M. W., Zaidi, S. A. H., Khan, N. R., Mirza, F. M., Hou, F., & Kirmani, S. A. A. (2019). The
737 impact of natural resources, human capital, and foreign direct investment on the ecological
738 footprint: the case of the United States. *Resources Policy*, 63, 101428.

739 Zhao, J., Zhao, Z., & Zhang, H. (2019). The impact of growth, energy and financial development on
740 environmental pollution in China: New evidence from a spatial econometric analysis. *Energy*
741 *Economics*, 104506.

742

743 Declarations

744 **5.2. Ethics approval and consent to participate**

745 No human data is involved in this research.

746 **5.3. Consent for publication**

747 It is not applicable to this study, as there is no human data involved in it.

748 **5.4. Availability of data and materials**

749 The datasets/codes used and/or analyzed during the current study are available from the
750 corresponding author on reasonable request.

751 **5.5. Competing interests**

752 The authors declare that they have no competing interests

753 **5.6. Funding**

754 There is not involvement of funding in this research.

755 **5.7. Authors' contributions**

756 **BHN** contributed in Writing – original draft, Conceptualization and Formal analysis and **AA**
757 performed Project administration, Writing – review & editing. All authors read and approved the
758 final manuscript.

759 **5.8. Acknowledgements**

760 The authors acknowledge the feedback of the two reviewers.

761 **5.9. Authors' information (optional)**

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764

Figures

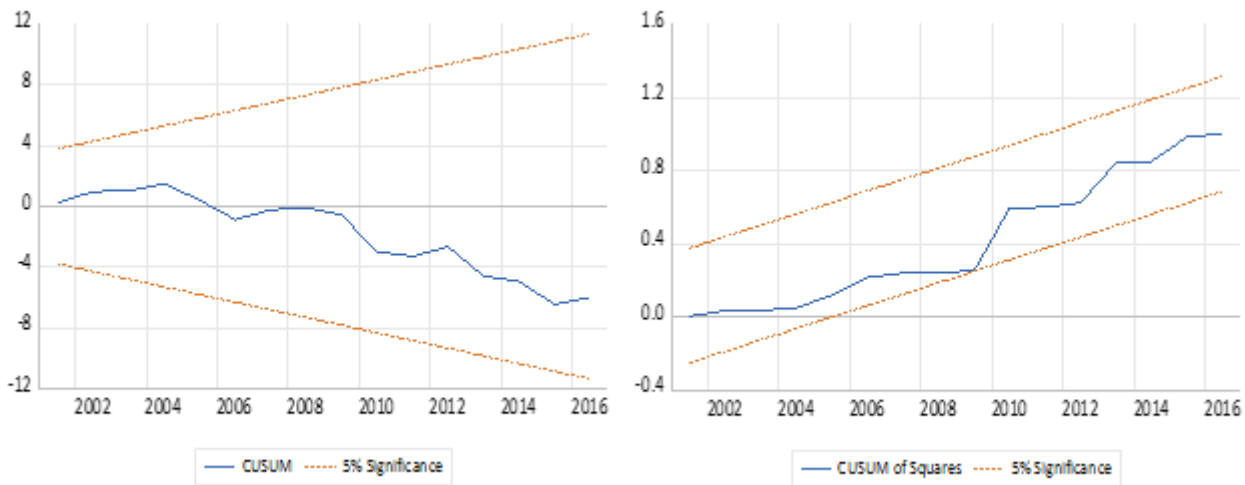


Figure 1

a. The CUSUM test and Figure b: The CUSUMSQ test

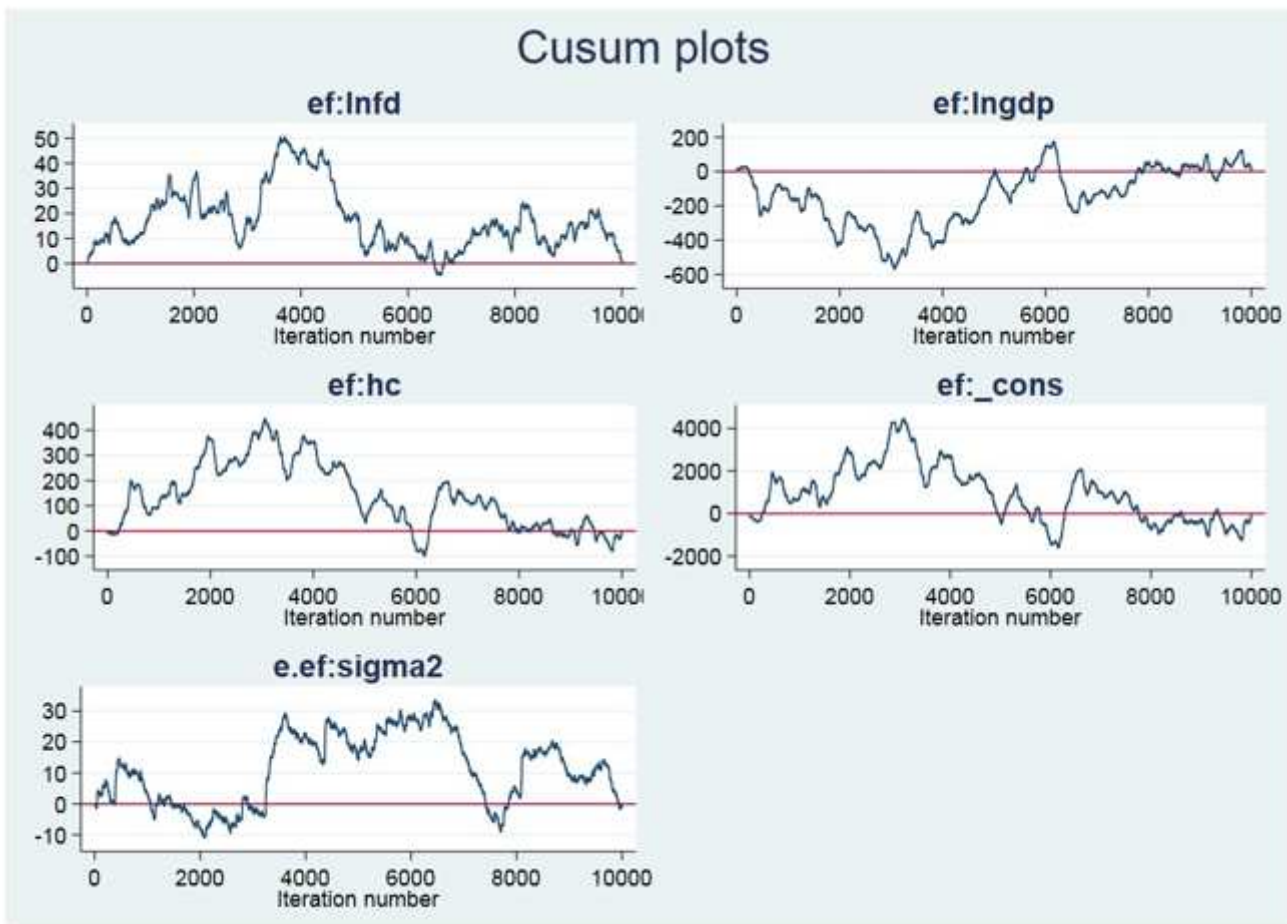


Figure 2

The CUSUM plot of the parameters

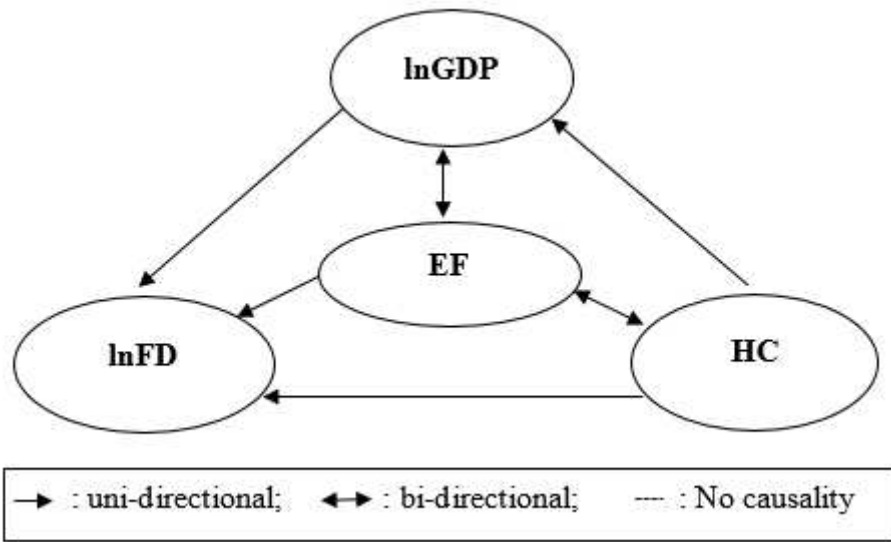


Figure 3

Plots of the causality test