

Energy Consumption, Economic Growth And Environmental Sustainability Challenges For Belt And Road Countries: A Fresh Insight From “Chinese Going Global Strategy”

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1 **Energy consumption, Economic growth and Environmental Sustainability**
2 **Challenges for Belt and Road Countries: A Fresh Insight from “Chinese**
3 **Going Global Strategy”**
4

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24 **Abstract**

25 The present study investigated impact of energy and economy related variables on CO₂
26 emissions in 49 countries of belt and road initiative from 1995-2018. The robust type of cross-
27 section dependence and heterogeneity methods were adopted to analyze data set of countries.
28 Energy consumption, foreign direct investment, medium and high-tech industry, and GDP has
29 been found highly unfavorable for the ecological health (CO₂ emissions) in 49 nations on BRI
30 panel. However, renewable energy consumption has been found in positive correlation with
31 environmental quality (CO₂). Financial development indicator has no significant impact on CO₂
32 emissions in present study. The present outcomes clearly claim strong relationship of economic
33 growth and energy with increased CO₂ emissions in 49 nations. Therefore, it is important for
34 policy makers, experts and governments to incentivize and appreciate portfolio investors for
35 sustainable green investments to transform the economic growth into a sustainable and energy
36 efficient development.

37

38 **Keywords:** Energy consumption; Economy; Environment; Belt and Road Initiative; Sustainable
39 development; Carbon Dioxide

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45 **Declaration**

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1. Introduction

In the current era of development and modernization, climate change is the biggest threat particularly to the human beings and earth ecosystem in total. Global emissions of Greenhouse gases (GHGs), Carbon dioxide, and rise in atmospheric temperature are being considered as the core reasons of the global climate change (IPCC 2014). The 21st Conference of Parties (December 2015) held in Paris reached at an agreement called “Paris agreement” which emphasis on the limiting of global warming to well below 2 °C and working on developing strategies for long term reduction of greenhouse gases to achieve long term goals of Paris agreement (UNFCCC. 2018). The international trade has implications on the environment, which ultimately plays its role in global climate change (Cai et al. 2018).

The “Belt and Road (BRI)” is an initiative taken by Chinese government to develop international cooperation and economic strategy (Chen 2016, Rauf et al. 2020). This initiative has main goals to cover international trade, infrastructural and financial connectivity among partner countries, policies integration and coordination, sharing technologies for the development and economic advancement of partner countries around the globe (Finance 2021, Intelligence 2017). BRI has potential to develop a unified world trade partnership along with a strong geopolitical coalition, which will bring a common future for all partner nations (Ho 2017). The projects under BRI will have strong impact on the economic development of the partner economies (Yii et al. 2018), through trade extension, access to advanced markets, shared skills, technologies and manpower, and inflow of funds towards the under developed, developing and emerging countries (Economy 2017). The Ministry of Ecology and Environment of China issued guidelines in 2017 for promoting “Green Belt and Road”. Later on, mentioned Chinese ministry initiated “BRI International Green Development Coalition” focusing on green initiative (finance,

89 transport, innovation, urbanization and standards). They are mainly addressing the five goals of
90 BRI initiative with the green development concept (Finance 2021). Along with advancement in
91 infrastructure, economic and trade cooperation with developing and developed countries (Du
92 &Zhang 2018), the climate change related issues and energy cooperation are major concerns of
93 China while expanding BRI projects range (Zhang et al. 2017b). European Union (EU) and
94 China in this regard showed strong commitment on clean energy (Zhang 2021) and climate
95 change(Torney &Gippner 2018) through intensifying economic, political and technical
96 cooperation (Liu &Hao 2018).

97 The energy growth has strong correlation with financial development (here in the sense
98 of economic growth) and environmental change which can be found in literature. As, Grossman
99 and Krueger (Grossman &Krueger 1995) testified the three stages of Environmental Kuznet
100 curve (EKC) (Kuznets 1955) where first phase focusses on evolution of economy along with
101 policy formulations while neglecting the ecological impacts of the development. Second phase
102 shows the intensified emissions of CO₂ due to economic evolution while third phase is about
103 realization of damage and adaptation of environment friendly policies and technologies to
104 minimize the environmental impact. Many researchers around the globe tested EKC hypothesis
105 and found a strong liaison between economic development and emissions of CO₂. Toman
106 andJemelkova (2003) tested 25 OECD countries, Musolesi et al. (2010) tested 106 countries,
107 Jaunky (2011) tested 36 high income countries; Apergis andOzturk (2015) tested 14 Asian
108 countries. All aforementioned studies approved EKC hypothesis and found that emissions of
109 CO₂ have strong long-term relationship with economic growth. Balsalobre-Lorente et al. (2018)
110 also investigated the economic advancement relationship with emissions of CO₂ that was N-
111 shaped nexus between them in 5 European countries.

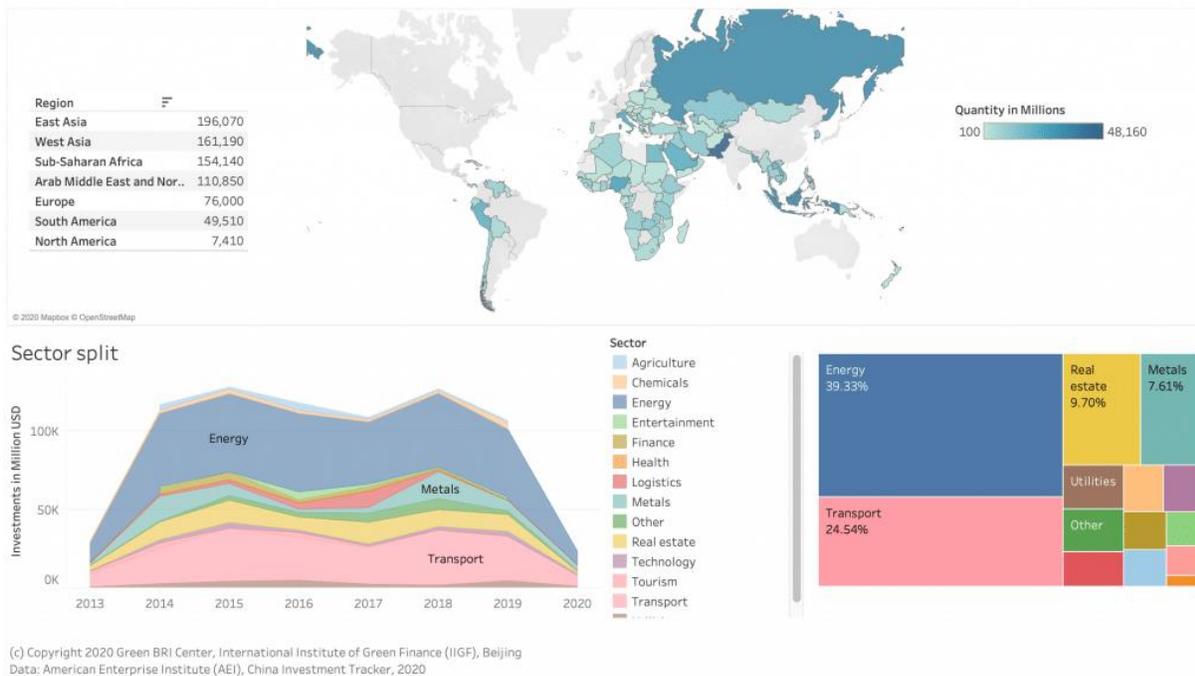
112 Ayeche et al. (2016) investigated 40 European economies and found a strong linkage
113 between GDP, financial development, trade openness and emissions of CO₂. In case of China,
114 Xu and Lin (2016a, b) found that rapid industrialization and economic development are the key
115 factors behind CO₂ emissions in China. But Dombrowski (2017) warned that BRI projects will
116 transfer the CO₂ emitting industries and businesses to other BRI partner countries, relieving
117 China in a better state without having severe environmental impacts on it. The Statistical review
118 by British Petroleum (Petroleum 2017) showed that BRI partner countries are contributing
119 around 61.4% of CO₂ emissions among which 80% is energy consumption based emissions, that
120 leading to global ecological deterioration impacts. Rauf et al. (2018a) also mentioned that BRI
121 host countries might face severe natural resources deterioration effects along with stern impacts
122 on their culture and ecology. Figure 1 shows the investments of China in BRI partner nations and
123 39.3% of those investments belongs to the energy sector.

124 The BRI projects can have impacts of accelerated global warming due to increased
125 infrastructural, energy generation and trade activities (Fan et al. 2017, Zhao et al. 2016). On the
126 other hand, it could also be an opportunity to minimize the global CO₂ emissions and to improve
127 the quality of environment while implementing BRI in the partner countries (Zhang et al. 2017a).
128 It is possible especially through empowering developing countries having abundant renewable
129 resources but do not have sufficient technological and financial equipment to utilize those
130 resources (Schwerhoff & Sy 2017). The BRI could also serve as a platform for the partner
131 countries to take joint action against CO₂ emissions (Zhang et al. 2018). There has been highly
132 sophisticated research been done on the economic, trade and international cooperation impacts of
133 BRI on partner countries in particular and on the world in general. But it is a dire need to do the
134 research on the long-term climate change related impacts of BRI in the partner countries and also

135 on the global warming. Therefore, the present study objectives are; to investigate the impacts of
 136 energy consumption, economic growth and other developmental parameters on the emissions of
 137 CO₂ in BRI partner countries; and to evaluate the linkages between economic growth,
 138 environmental sustainability and energy growth factors in BRI countries.

139 **Figure 1:** Investments of China in BRI countries from 2013-H12020 (million USD) ¹

140 2. Materials and Methods



141 2.1. Data and Variables

142 This study contemplates BRI-associated nations in the terrestrial locations of Europe,
 143 East Asia, Pacific, Central Asia, South Asia, Middle East and North Africa. There were 49

¹ Source: <https://green-bri.org/wp-content/uploads/2020/09/Investments-in-the-Belt-and-Road-Initiative-BRI-2020-1024x614.png>

144 countries selected from those regions based on the data available (Please see list of countries in
145 Table 1A).

146 The present study used CO₂ as dependent variable and other variables mentioned in Table
147 1 considered independent variables. The dataset has been log-transformed for the purpose of
148 standardization. This standardization will minimize the robustness from data and will minimize
149 the enlargement of coefficients, multi-correlations and autocorrelations related problems. The
150 description and sources of all the variables have been stated in Table 1.

151 **Table.1** Data Description and Symbolization

Variables	Description
Energy consumption (EC)	Litres to kilograms energy usage per capita
Carbon emission (CO ₂)	Metric tones of CO ₂ atmospheric release per capita
Foreign Direct Investment (FDI)	Percent of GDP (total inflow of foreign direct investment)
Trade friendliness/openness (TOP)	Trade (% of GDP)
Financial development (FD)	Percent of GDP (private sector domestic credit)
Industry (medium and high-tech) (MHI)	MHI as percent of value added manufacturing
Renewable energy consumption (REC)	REC as percent of total final energy consumption
Gross domestic product (GDP)	GDP per capita with a constant o 2010 US dollars

152 ²

153 2.2. Steps for Econometric analyses

² Source: <https://data.worldbank.org/>

154 To test the hypothesis for the underlying variables, i.e., CO₂, ECON, FD, GDP, FDI,
 155 MHI, TOP, and REC a primary equation is formed (Equation-1). Based on several recent studies
 156 e.g. (Al-Mulali et al. 2015, Behera & Dash 2017, Doğan et al. 2021, Gulistan et al. 2020, Haseeb
 157 & Azam 2020, Khan et al. 2020, Khan et al. 2019, Rauf et al. 2020, Saboori & Sulaiman 2013) the
 158 following relationship among variables under investigation in this study has been developed;

$$159 \quad CO_2 = f(EPC, FD, GDP, FDI, MHI, TOP, REC,) \quad (1)$$

160 Here in Eq. 1, Carbon dioxide as mentioned earlier is the dependent variable and
 161 relationship will be estimated as “CO₂ is equal to the function of independent variables”. The
 162 econometric analysis for this study comprises the following steps. (i) Analysis of the descriptive
 163 statistics such as correlation analysis for the selected variables for this study. (ii) Testing the
 164 dependence (cross-sectional) of the countries data to confirm that the estimate drawn from the
 165 dataset is reliable. Afterwards, co-integration checkup based on the results can verify long-run
 166 integrated forms amongst the variables (Al-Mulali et al. 2013). (iii) Fully Modified OLS
 167 (FMOLS) and Dynamic OLS (DOLS) models for equilibrium relationships. Finally, (iv) The
 168 Panel Heterogeneous Granger causality test will be used to detect interconnectivity as used by
 169 Rauf et al. (2018a).

170 Equation 1 is an appropriate representation of the base model. The base model is
 171 rewritten with the natural log form of data and equation 2 is formed.

$$172 \quad CO_{2i,t} = \alpha + \beta_1 \ln ECON_{i,t} + \beta_2 \ln FD_{i,t} + \beta_3 \ln GDP_{i,t} + \beta_4 \ln FDI_{i,t} + \beta_5 \ln MHI_{i,t} + \beta_6 \ln TOP_{i,t} \\
 173 \quad + \beta_7 \ln REC_{i,t} + \varepsilon_{i,t} \quad (2)$$

174 Whereas, i = number of countries, t = time, \ln = natural logarithm, α = intercept, β =
 175 slope to the parameters, and $\epsilon_{i,t}$ = error terms for the equation.

176 **2.2.1. Dependency test (Cross-sectional)**

177 Testing cross-sectional dependency of variables is critical to form any econometric
 178 model. Cross-sectional dependence (CD) is an important test in large data econometric modeling.
 179 It should be checked by investigators before evaluating investigation of any group. The existence
 180 or not of such a violation will correct the auxiliary path, which must be followed later. If the
 181 information in the dataset has a cross-sectional dependency, the other phases of the analysis
 182 should retain tests that are consistent with the cross-sectional dependency.

183 In structured variables, the cross dependence based on residuals can be examined by LM test
 184 (Breusch & Pagan 1980), scaled LM test (bias-corrected) (Baltagi et al. 2012) and CD test
 185 (Pesaran 2004). “No” cross-sectional dependence among that residual based dataset is the Null
 186 hypothesis.

187 Therefore, both the CD and LM test are structured in the following ways:

$$188 \quad LM = \sqrt{\frac{2T}{N(N-1)}} \left(\sum_{i=1}^{N-1} \sum_{j=i+1}^N \hat{\rho}_{ij} \right) \frac{(T-k)\hat{\rho}_{ij}^2 - E(T-k)\hat{\rho}_{ij}^2}{Var(T-k)\hat{\rho}_{ij}^2} \quad (3)$$

$$189 \quad CD = \sqrt{\frac{2T}{N(N-1)}} \left(\sum_{i=1}^{N-1} \sum_{j=i+1}^N \hat{\rho}_{ij} \right) \sim N(0,1) i, j = 1, 2, 3 \dots 65 \dots N \quad (4)$$

190 Whereas, $\hat{\rho}_{ij}^2$ is residuals correlation, which was valued by using Ordinary Least Square
 191 equation. The results of the above given equations are given below in Table 2 with 1% level of
 192 significance for Null hypothesis (H^0).

193 **Table.2** Results of (CD) Test

Test	Statistic
LM of Breusch-Pagan	11392.35***
LM of Pesaran scaled	210.6577***
Bias-corrected scaled LM	209.6777***
CD by using Pesaran	73.54330***

194 ¹Note: “***” represent 1% level of significance.

195 The cross dependence of the dynamic panels for residuals of dataset has been
 196 investigated by using CD tests of Frees (Frees 1995, 2004) Friedman (1937) and Pesaran (2004).
 197 The short term and large cross-sectional residual dependence for given dataset (time and number
 198 of countries) gave a clear understanding about the relationship between the countries over the
 199 given time period. The results (Table 3) of these tests rejected Null hypothesis of cross-section
 200 independence.

201 **Table.3** Results of (CD) test of the residuals

Test	Statistic
Pesaran CD test	73.54330***
Friedman test	265.9161***
Frees test	7.504881***

202 ¹Note: “***” represent 1% level of significance.

203 2.2.2. Unit Root Tests

204 The unit root tests are of two types. First type considers the self-determining power of
 205 CD of target countries. The 2nd type/generation test allows the CD of the countries. The current
 206 study used both types of unit root tests to provide strong justification of stability in the results.
 207 The current panel's data acknowledges that there are longer-term events, which could increase
 208 the degree of independence (d.f) and exacerbate the multidimensional crisis to assess the
 209 equation of OLS. Therefore, panel data can withstand more compelling scientific techniques and
 210 asymptomatic statistics, which follow a general distribution rather than a noise distribution.

211 Choi(Choi 2006) devised the panel unit root test with opposite assumption/hypothesis
 212 like Hadri (2000). Levin et al. (2002) used restricting type of panel unit root test for samples of
 213 the finite properties while Im et al. (2003) also advised heterogenous panel unit root test.

214 Therefore, this study applies the LLC, IPS, and ADF Fisher Chi-square test for testing the
 215 unit root (Table 10) to hold the order of cointegration among the variables under this study. In
 216 this connection, the panel unit root test of IPS is depicted with the following equation:

$$217 \quad \Delta y_{i,t} = \alpha_i + \beta_i y_{i,t-1} \sum_{j=1}^{p_i} \rho_{ij} \Delta y_{i,t-j} + \varepsilon_{i,t} \quad i = 1, \dots \quad t = 1, \dots T \quad (5)$$

218 Equation 5 above depicts $y_{i,t}$ as the dataset containing i countries for t time but the lag
 219 operators are denoted with “ Δ ”. Here, $\varepsilon_{i,t}$ stands for the error term for the normally distributed
 220 sample BRI countries.

221 The results (Table 2 and Table 3) show the cross dependence in the dataset. Therefore,
 222 we need to apply 2nd type of CD tests to justify the hitch of cross-dependence. Pesaran(Pesaran

2007) defined the process of cross-sectional Im, Pesaran, and Shin (CIPS) and cross-sectional augmented Dickey-Fuller (CADF). The country to county cross-sectional dependence, reliability and steadfastness will be the outcomes of these two methods with their natural heterogeneity.

Therefore, the test may further be built as follows:

$$\Delta y_{i,t} = c_i + \alpha_i y_{i,t-1} + \beta_i \bar{y}_{t-1} + \sum_{j=0}^p \gamma_{ij} \Delta \bar{y}_{i,t-j} + \sum_{j=1}^p \delta_{ij} \Delta \bar{y}_{i,t-j} + \eta_{i,t} \quad i = 1, \dots, n \quad (6)$$

Whereas; c_i = constant, \bar{y} = mean of cross-section at “ t ” period, and p = lag operator.

Supposing t_i (N, TM) same as the time ratio of α_i , the mean of t-ratios (time ratios) will be as follows;

$$CIPS(N, T_m) = \frac{\sum_{i=1}^N t_i(N, T_m)}{N} \quad (7)$$

Here, $t_i(N, T_m)$ is Augmented Dickey-Fuller (CADF) indicators for the i^{th} cross-sections.

2.2.3. Co-integration Tests

The results of both types of unit root tests approved the stability of dataset. Co-integration tests by Pedroni (1999), (Pedroni 2004) can further validate the level of co-integration. Robustness can be confirmed by using Westerlund co-integration test (Westerlund 2007) to get dependency of cross sections. The base of co-integration test is Engle-Granger (a typical unit root test) which has been further expanded by Westerlund et al. (2015). Also see (Al-Mulali et al. 2012, Ciarreta & Zarraga 2010 (Khan et al. 2017, Rauf et al. 2018a) for the purpose of determining long-run connectivity between candidate variables. Therefore, it has been verified that all given variables together formulated into first order (Equation 1).

242 Likewise, The Pedroni cointegration test augmented the following equation:

$$243 \quad CO2_{i,t} = \alpha + \delta_i t + \beta_1 \ln ECON_{i,t} + \beta_3 \ln FD_{i,t} + \beta_2 \ln GDP_{i,t} + \beta_4 \ln FDI_{i,t} + \beta_5 \ln MHI_{i,t} + \\ 244 \quad + \beta_5 \ln TOP_{i,t} + \beta_5 \ln REC_{i,t} + \varepsilon_{i,t} \quad (8)$$

$$245 \quad i = 1, \dots \quad t = 1, \dots T$$

246 Whereas, α_i is constant for each country, and $\delta_i t$ is the full panel deterministic trends of
247 the particular country. There were eleven statistical results of the Pedroni co-integration test
248 while investigating both hypotheses. β_1 is homogenous for null hypothesis and its heterogeneous
249 for alternative hypothesis. The co-integration among variables can be seen in Table 9. The
250 uniformity between the target variables has been found normally distributed, which verifies the
251 Pedroni co-integration test. This relationship could be written as following equation;

$$252 \quad \sqrt{\frac{N'_{N,T} - \mu\sqrt{N}}{\sqrt{V}}} \rightarrow N(0,1) \quad (9)$$

253 In equation 9, μ and V stand for the Monte Carlo oriented adjustment measures.

254 The first four results of Panels (v, rho, PP, and ADF statistics) in the Table 9 are within-
255 dimension statistics and latter three Groups (rho, PP and ADF) are between the dimension
256 statistics. Therefore, we have at least 4 statistics out of 7 fulfills the lowest criterion for long-run
257 linear co-integration approval within target variables.

258 After the approval of cross dependence between the target variables, the co-integration
259 test by Westerlund (2007) has the ability to produce stable and vigorous results for the approval
260 of the co-integration level. The results of (Westerlund 2007) are present in the form of two
261 groups/forms. The first group is called cluster based group (Gt and Ga), whereas, other group is

262 called panel statistics group (Pt and pa) (Rauf et al. 2018a, Saud et al. 2019) as shown in Table 8.
 263 The results show that co-integration exists among CO₂ and all other independent variables for all
 264 the 49 countries of the study regions.

265 **2.2.4. The Dynamic Panel Data Estimation**

266 The FMOLS valuation proposed by Pedroni (Pedroni 2001) and the DOLS valuation
 267 proposed by (Kao & Chiang 2001) and (Stock & Watson 1993) that have been used in this recent
 268 study to explore the long-run co-integration amongst variables.

269 Meanwhile, the following FMOLS and DOLS equations are presented to test the hypotheses:

$$270 \hat{\beta}_{NT} = \left[\frac{\sum_{i=1}^N \sum_{t=1}^T (x_{it} - \bar{x}_i)(y_{it} - \bar{y}_i) - T\hat{\gamma}_i}{\sum_{t=1}^T (x_{it} - \bar{x}_i)^2} \right] \quad (10)$$

$$271 \text{ Where } \hat{\gamma}_i = \hat{\Gamma}_{21i} + \hat{\Omega}_{21i}^0 - \frac{\hat{\Omega}_{21i}}{\hat{\Omega}_{21i}} (\hat{\Gamma}_{22i} + \hat{\Omega}_{22i}^2)$$

$$272 \text{ And } \hat{\Omega}_i = \hat{\Omega}_i^0 + \hat{\Gamma}_i + \hat{\Gamma}'_i$$

273 Whereas, $\hat{\Omega}_i$ = long-run matrix of stationarity, $\hat{\Omega}_{21i}^0$ = term to reject the covariance b/w
 274 errors terms of stationarity, and $\hat{\Gamma}_i$ = modified covariance between independent variables.

275 **2.2.5. Heterogeneous Panel Causality test**

276 At the last stage of econometric analysis, the panel Granger causality test has been used
 277 to find the instrumental correlation between target variables under investigation in the study.

278 **3. Results and Discussions**

279 The present study used a series of correlational and cross-sectional dependence tests to
 280 develop understanding about the effects of energy growth, financial development, GDP, medium
 281 and high-tech industries, trade openness and renewable energy consumption on the emissions of
 282 Carbon Dioxide (CO₂). The results presented in the form of series of Tables to clearly represent
 283 the outcomes of the present study for the regions and countries of the study. The empirical
 284 results obtained through current investigation can help the policy makers to achieve “Green BRI”
 285 goals in regional panels.

286 **3.1. Descriptive Statistics**

287 The summary of statistics has been presented in Table 4 comprising 49 countries and
 288 1274 observations dataset. The variables data has been standardized by using natural logarithm
 289 to avoid heteroscedasticity. The consumption of energy was a trending variable with the mean of
 290 1945.1240 and a standard deviation of 2698.9130. The mean emissions of CO₂ were lower
 291 (5.0492) in million tons but it was highly variable in the different countries of the different
 292 regions. This type of variation indicates the different level of advancement in individual
 293 countries. Similarly, mean GDP of the overall countries in the panel was higher (2698.9130)
 294 with lower standard deviation indicating the improved state of economy of the countries of the
 295 regions under investigation. The mean value of FD and MHI had similar outcomes (38.2041 and
 296 24.4740 respectively) indicating their strong dependence on each other. Similarly, TOP and FDI
 297 shows similar trends in the given time span.

298 **Table 4.** Descriptive Statistics

Variable→	CO ₂	EC	FD	FDI	GDP	MHI	REC	TOP
Mean	5.0492	1945.1240	38.2041	4.0261	2698.9130	24.4740	17.6670	88.0877

Median	2.9844	899.8239	33.4432	2.7452	4456.3600	24.0068	6.0378	81.1580
Maximum	35.9158	12406.7500	165.3904	54.2391	64864.7200	88.0370	92.3802	437.3267
Minimum	0.0000	0.0000	0.0000	-40.4143	0.0000	0.0000	0.0000	0.0000
Std. Dev.	6.4352	2698.9130	34.1423	5.5015	11334.0600	17.0908	23.1835	58.3168
Skewness	2.1786	2.2411	1.0170	2.3774	2.4375	0.6725	1.4498	2.0877
Kurtosis	7.9716	7.7552	3.8979	26.4151	9.4368	3.6670	4.2063	11.1954
Jarque-Bera	2319.8320	2266.6950	262.4066	30303.8600	3460.9280	119.6407	523.5765	4490.7950
Probability	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Sum	6432.74	2478088	48672.04	5129.211	11037057	31179.84	22507.74	112223.7
Observations	1274	1274	1274	1274	1274	1274	1274	1274

299

300 3.2. Correlation Analyses

301 The present study correlation analyses results show the highly significant positive
302 correlation between CO₂ and GDP (0.7153 ***) MHI (0.2564***), EC (0.8063***), FD
303 (0.08291***) and TOP (0.2080***) respectively. There was highly significant negative
304 correlation found between CO₂ and REC (-0.3956***) and as shown in Table 5. There was no
305 significant correlation found between CO₂ and financial development (FDI). The results of
306 strong association between GDP, MHI, and EC and CO₂ has also been observed by Rauf et al.
307 (2020). But current study result contradicts the weak correlation results of CO₂ and FD with
308 (Rauf et al. 2020). Therefore, it can be easily observed that CO₂ emissions are mainly connected
309 with the GDP of the countries, medium and high-tech industries, and energy consumption as
310 correlation statistics indicated clearly about it. These indicators are main drivers of the
311 atmospheric CO₂ emissions and controls the atmospheric conditions through increased
312 greenhouse gas emissions in the countries under current investigation. Therefore, the two-way

313 correlation estimated provides a good insight into series of datasets. However, it is important to
 314 further validate the results and develop cross sectional relationships to cross validate the
 315 established preposition.

316 **Table 5.** Correlation Statistics

Variab le	CO₂	EC	FD	FDI	GDP	MHI	REC	TOP
CO ₂	1.0000							
EC	0.8063** *	1.0000						
FD	0.08291* **	0.0577**	1.0000					
FDI	0.02907	0.0815** *	0.1795** *	1.0000				
GDP	0.7153** *	0.5745** *	0.2331** *	0.1343** *	1.0000			
MHI	0.2564** *	0.2280** *	0.3897**	0.0835**	0.4267** *	1.0000		
REC	- 0.3956** *	- 0.3394** *	- 0.2230** *	- 0.1180** *	- 0.3904** *	- 0.2234** *	1.0000	
TOP	0.2080** *	0.2283** *	0.3752** *	0.4265**	0.4050 ***	0.4427** *	- 0.2271** *	1.000 0

317 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.

318 **3.2.1. Unit Root tests and Slope Homogeneity**

319 This study used first-generation/type LLC, IPS and ADF (Table 6) and second-
 320 generation/type CIPS and CADF (Table 7) of unit root tests for data stationarity checking. The
 321 stationarity of individual variables examined and results indicated the difference between

322 variables in panels of study regions. Some of the unit roots tests discarded the null hypothesis at
 323 their level, the stationarity at 1st order has been supported by most of the tests. The results of CD
 324 test showed strong cross dependence between variables. Further, study examined the slope
 325 heterogeneity test for heterogenous panels in Table 8. The Pedroni and Kao based tests (1st type
 326 /generation co-integration tests) might face the issue of lower co-integration between the
 327 variables. To avoid this problem with Pedroni and Kao based co-integration tests, the
 328 (Westerlund 2007) has been used to estimate the level of co-integration between study variables,
 329 (see (Yasmeen et al. 2018)). The Pedroni co-integration test gave strong evidence of rejecting null
 330 hypothesis through 4 out of 7 values having higher significance level “p”. The results of
 331 Westerlund co-integration test under the cross-dependence situation proved to be the best choice
 332 as shown in Table 9 which validates the long-run co-integration between variables. The results of
 333 Pedroni (Table 10) and Kao (Table 11) shows the high level of co-integration among all
 334 variables in long-run. Further investigations using FMOLS and DOLS models will give insight
 335 into the long-run co-integration in full and regional panels.

336 **Table 6.** Panel Unit Root test results (LLC, IPS and ADF)

At level								
Methods	CO₂	EC	FD	FDI	GDP	MHI	REC	TOP
Levin, Lin & Chu t*	6.2575 8	9.0830 5	0.3153 7	- 7.1973 5***	2.7861 7	31.398 2	4.9201 6	- 20.249 6***
Im, Pesaran and Shin	6.1258 3	7.9944 0	1.9067 7	- 8.4359 6***	5.7933 8	5.9824 4	6.2350 7	- 13.017 0***
ADF - Fisher Chi-square	45.207 5	13.920 1	75.935 2	256.12 5***	65.704 6	48.270 1	51.045 6	174.24 9***

At first Difference

Methods	CO ₂	EC	FD	FDI	GDP	MHI	REC	TOP
Levin, Lin & Chu t*	-16.7716***	-18.4180***	-4.36735***	-16.7018***	-301.090***	186.936	-28.8923***	-117.817***
Im, Pesaran and Shin	-14.2569***	-13.9978***	-12.6790***	-21.9851***	-80.3471***	0.88611	-24.5213***	-54.9800***
ADF - Fisher Chi-square	383.619***	374.265***	362.666***	614.258***	395.857***	136.909***	651.300***	751.821***

337 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.

338 **Table 7.** Results of CIPS and CADF (Panel unit root tests)

At Level

Methods	CO ₂	EC	FD	FDI	GDP	MHI	REC	TOP
CIPS	-2.679**	-6.093** *	-2.169	-3.825** *	-3.049** *	-2.202	-2.696**	-2.401
CADF	-2.256	-1.633	-3.033** *	-3.131** *	-2.539**	-2.179	-1.929	-1.923

1st Deference

Methods	CO ₂	EC	FD	FDI	GDP	MHI	REC	TOP
CIPS	-2.679** *	-6.320** *	-4.939** *	-5.793** *	-4.616** *	-4.459** *	-4.732** *	-4.419** *
CADF	-3.896** *	-3.638** *	-3.491** *	-4.580** *	-3.411** *	-3.352** *	-3.168** *	-3.373** *

339 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.

340 **Table 8.** Testing for slope heterogeneity

	Delta	P-value
	17.394***	0.000
Adj.	21.512***	0.000

341 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.342 **Table 9.** Westerlund test of Cointegration

Statistic	Value	Z-value	P-value
Gt	-13.126***	-49.216	0.0000
Ga	-19.064*	11.516	0.058
Pt	-56.338***	-4.123	0.0000
Pa	-24.097***	8.456	0.0000

343 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.344 **Table 10.** Pedroni test of Cointegration

Panel and Group statistics	Weighted			
	Statistic	Probability	Statistic	Probability.
v-Statistic (Panel)	2.5574***	0.0053	-0.0548	0.5219
rho-Statistic (Panel)	1.8471	0.9676	1.2252	0.8898
PP-Statistic (Panel)	-2.7401***	0.0031	-5.6625***	0.0000
ADF-Statistic (Panel)	-7.7309***	0.0000	-8.2642***	0.0000
rho-Statistic (Group)	4.2118***	1.0000		
PP-Statistic (Group)	-5.3655***	0.0000		
ADF-Statistic (Group)	-7.7040***	0.0000		

345 ¹ *** shows statistical significance at 1% significance level.

346 The Kao co-integration test employed to validate the results of Pedroni co-integration tests. The
 347 result of Kao test showed -13.18381*** (Table 11), which clearly indicates that the previously
 348 applied integration tests were efficient and Kao test results are authenticating those previously
 349 obtained outcomes of co-integration. The similar results found in the study by (Rauf et al. 2018b)
 350 where their results of Pedroni co-integration tests were validated by Kao co-integration test.

351 **Table 11.** Kao test of Cointegration

	<u>t-Statistic</u>	<u>Prob.</u>
ADF	-13.18381***	0.0000
Residual variance	0.126030	
HAC variance	0.111819	

352 ¹ *** shows statistical significance at 1% significance level.

353 **3.2.2. Dynamic Panel data models**

354 The estimations obtained from co-integration test validated the long term relationship among
 355 variables which gave strong reason to apply FMOLS and DOLS to get stable outcomes(Pedroni
 356 2001); (Pedroni 2004). Both DOLS and FMOLS has been used to establish the expected
 357 relationship between the regressor and the regressed. The results shown in Table 12 clearly
 358 reveals that EC, FDI, GDP, and MHI found unfavorably influencing the environmental quality
 359 through carbon dioxide emissions. These results can be validated by results obtained from (Rauf
 360 et al. 2020). Renewable energy consumption (REC), and trade openness (TOP) have favorable
 361 effect on the environment. (Rauf et al. 2018b) also found that trade openness do not have
 362 negative effects on the environment in BRI partner countries.

363 **Table 12.** Results for FMOLS and DOLS for countries under investigation

Regressor: CO₂ Emissions		
Panel-49 BRI countries		
Variables	Coefficients	
	FMOLS	DOLS
LNEC	0.173604***	0.282734***
LNFD	0.000992	0.031297
LNFDI	0.044221**	0.054350
LNGDP	0.438718***	0.467377***
LNMIHI	0.170090***	-0.155954
LNREC	-0.306305***	-0.482432***
LNTOP	-0.002527	0.015257
R-squared	0.869660	0.999602
Adjusted R-squared	0.862954	0.989165

364 ¹ *, **, *** shows statistical significance at the 10%, 5% and 1% respectively.

365 Precisely, it can be observed from the results of FMOLS that 1% increase in energy
366 consumption EC, FDI, GDP and MHI leads to the degradation of environment (CO₂ emissions)
367 having strong relationship of 0.173604***, 0.044221**, 0.438718***, and
368 0.170090*** respectively. The results of both models (DOLS and FMOLS) found to have been
369 similar in terms of developing relationship between regressor and the regressed ones. We can
370 summarize from the results of both models that the 49 countries studied need to understand that
371 economic growth, medium and high-tech industries and foreign direct investments should
372 transform their sources of energy generation from fossil fuel based to renewable energy
373 generation sources. These findings reveal that level of emissions of CO₂ in the BRI partner

374 countries is dependent on the consumption of energy. The increased magnitude of the energy
375 consumption will cause the increased impact on the environment. It further leads the researchers
376 to emphasis mainly on the innovation-based advancements. The adaptation of energy efficient
377 technologies can reduce the burden on energy generation sector (Choi et al. 2012), which will
378 ultimately have reduced impact on the ecological health of the countries. Therefore, it is strongly
379 recommended to promote renewable energy generation and sharing of environment friendly
380 technologies among the BRI partner countries for sustainable economic growth with minimum
381 impact on the ecological health of the partner countries. In this regard, the Green BRI initiative is
382 the good step to promote the environment friendly technologies and options for sustainable
383 energy, economy and environmental growth in BRI partner countries. The current study results
384 infers similar outcomes to (Apergis &Ozturk 2015, Arouri et al. 2012, Atici 2009, Bekhet
385 &Othman 2017, Hafeez et al. 2018, Jalil &Mahmud 2009, Khan et al. 2017, Nasir &Rehman
386 2011, Omri 2013, Rauf et al. 2018b, Xu &Lin 2016b). The option of carbon free technologies
387 such as nuclear, biomass based, wind turbines, hydropower, and solar energy and their associated
388 technologies can transform the growth patterns of BRI partner countries with improved
389 environmental quality. (Javid &Sharif 2016) found similar findings for Pakistan, (Zhang &Gao
390 2016) and (Xu &Lin 2016b) for China, (Kasman &Duman 2015) for European Union members,
391 ((Rauf et al. 2018a); (Rauf et al. 2020); (Rauf et al. 2018b) for BRI member countries.

392 ***3.2.3. The Panel granger causality analysis***

393 The causality between CO₂ and independent variables has been investigated using Granger panel
394 causality analysis. (Dumitrescu &Hurlin 2012) developed the causality test, which also addresses
395 the problem of heterogeneity among variables. Therefore, the causality test by (Dumitrescu
396 &Hurlin 2012) has been used in the present study for the selected BRI countries. The divergent

397 results of causality test found for the 49 BRI partner countries in the present study. The results of
 398 Granger causality test are presented in Table 12. The results are showing quite clear causal
 399 relationship of CO₂ with other variables. Energy consumption has unidirectional (one way)
 400 relationship with CO₂ emissions, while financial development, GDP, MHI and REC showed
 401 bidirectional (feedback type) relationship with CO₂. Foreign direct investment (FDI), and TOP
 402 has inverse unidirectional relationship with CO₂. The resultant pathways of relationship of
 403 independent variable with environmental health will help the policymakers to let them develop
 404 sustainable and environment friendly strategies in BRI partner countries.

405 **Table 13.** Dumitrescu Hurlin Panel Causality analysis

<i>Null Hypothesis:</i>	<i>Zbar-Stat.</i>	<i>Prob.</i>	<i>Relationship directions</i>		
$LNEC \neq LNCO_2$	46.9575***	0.0000	<i>LNEC</i>	→	<i>LNCO₂</i>
$LNCO_2 \neq LNEC$	0.56576	0.5716			
$LNFD \neq LNCO_2$	40.5894***	0.0000	<i>LNFD</i>	↔	<i>LNCO₂</i>
$LNCO_2 \neq LNFD$	21.5626***	0.0000			
$LNFDI \neq LNCO_2$	-0.44992	0.6528	<i>LNFDI</i>	←	<i>LNCO₂</i>
$LNCO_2 \neq LNFDI$	4.27438***	0.0000			
$LNGDP \neq LNCO_2$	2.21961**	0.0264	<i>LNGDP</i>	↔	<i>LNCO₂</i>
$LNCO_2 \neq LNGDP$	8.40507***	0.0000			
$LNMIHI \neq LNCO_2$	-2.6427***	0.0082	<i>LNMIHI</i>	↔	<i>LNCO₂</i>
$LNCO_2 \neq LNMIHI$	181.943***	0.0000			
$LNTOP \neq LNCO_2$	1.28659	0.1982	<i>LNTOPG</i>	←	<i>LNCO₂</i>
$LNCO_2 \neq LNTOP$	23.939***	0.0000			
$LNREC \neq LNCO_2$	80.8119***	0.0000	<i>LNREC</i>	↔	<i>LNCO₂</i>
$LNCO_2 \neq LNREC$	2.2346**	0.0254			

407 This causal relationship between target variables clearly justifies the high rate of CO₂
 408 emissions connectedness with the economic advancement (Table 13). It conclusively defines the
 409 influence of economic growth on the environmental health in the BRI partner countries. Carrying
 410 on these kind of economic development activities may lead to the increased global warming
 411 along with unprecedented type of human health effects. The possible solution of such negative
 412 thrust back of economic growth is the adaption of technologically advanced and green solutions.
 413 These type of causal relationships are also found in studies by (Al-Mulali et al. 2015, Katircioglu
 414 2017); (Rauf et al. 2020, Saud et al. 2019)).

415 3.3. Checking regression robustness

416 The results of DOLS and FMOLS presented to see robustness (Table 11), this study
 417 applies the Dynamic Seemingly Unrelated Regression (DSUR). The purpose of applying DSUR
 418 is to check the robustness in the outcomes obtained from FMOLS and DOLS tests. The results
 419 presented in Table 14 shows that there was medium to high impact of indicators under
 420 investigation on emissions of Carbon dioxide ($R^2 = 0.60$) in the countries under investigation in
 421 the present study.

422 **Table 14.** Dynamic seemingly unrelated regression results

<u>Panel</u>	<u>Variable</u>	<u>Coefficient</u>	<u>z-Statistic</u>	<u>Prob.</u>
	LNEC	0.1559787***	24.67	0.0000
	LNFD	-0.0285969**	-2.51	0.0120
	LNFDI	0.0584332***	3.96	0.0000
	LNGDP	0.3897101***	21.37	0.0000
<u>49 BRI countries</u>	LNMIHI	0.1741924***	9.93	0.0000

LNREC	-0.243571***	-21.03	0.0000
LNTOP	0.0600654***	3.76	0.0000
CONSTANT	-3.481891***	-23.78	0.0000
R-squared	0.7559		
Chi Square	3944.86***		

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424 The results of DSUR endorse the results obtained by using FMOLS and DOLS, where
 425 EC, FDI, GDP and MHI are major drivers of the environmental degradation (Table 14). Trade
 426 openness and renewable energy consumption favored the environment as also found in FMOLS
 427 and DOLS test results. The indicator of FD has not been found with significant relationship with
 428 the emissions of CO₂, having similar results as previously used both models.

429 **4. Conclusions**

430 The present study conducted to examine the impacts of energy consumption, GDP,
 431 financial development, renewable energy consumption, foreign direct investment, and medium
 432 and high-tech industry on the emissions of Carbon dioxide (CO₂) in 49 nations on the panel of
 433 Belt and Road initiative. The duration of investigation expands from 1994 to 2019. The robust
 434 type of panel cross-section dependence and slope heterogeneity and other methods were adopted
 435 to analyze the dataset of BRI countries under investigation. The standardized (log-transformed)
 436 data has been used to employ slope heterogeneity, cross-sectional dependency of the panel data
 437 to confirm that the estimate drawn from the dataset is reliable.

438 Afterwards, panel tests (co-integration tests) used to verify the long-term integrated forms
 439 amongst the variables and, FMOLS and DOLS models used for long-term equilibrium
 440 relationship among the variables. Finally, The Panel Heterogeneous Granger causality test

441 applied to detect the interconnectivity between variables at causal bases (mainly CO₂ emissions
442 with other independent variables). The consumption of energy (EC) along with foreign direct
443 investment (FDI), medium and high-tech industry (MHI) and GDP has been found highly
444 unfavorable for the ecological health (CO₂ emissions) in 49 nations on BRI panel. However,
445 renewable energy consumption (REC) has been found a favorable impact on the environment
446 quality parameter (CO₂). There was no significant impact of financial development (FD)
447 indicator on CO₂ emissions has been observed in the present study.

448 More precisely, the energy consumption, medium and high-tech industries and GDP has
449 been the major variables found in all analytical outcomes, those having highly significant
450 impacts on the environmental quality/health (CO₂ emissions). However, the adaptation of
451 renewable energy sources has been found in significantly obliging impacts (favorable) in all
452 analytical outcomes with ecological health of the countries in BRI panel. The present outcomes
453 clearly claim the strong relationship of economic growth with increased CO₂ emissions in all 49
454 nations under investigation of Belt and Road initiative.

455 Therefore, it can be concluded that the huge investments of Chinese government under
456 BRI projects on energy sector (mainly based on fossil fuel-based energy generation) along with
457 industrial sector development are driving factors behind the environmental deterioration in those
458 countries. However, the impacts of BRI projects on environment can be minimized using
459 renewable energy generation sources especially those of carbon free energy generation
460 technologies. Further, the industrial pollution can also be minimized through regulating them
461 according to the environmental standards. The governments of BRI listed countries can
462 formulate sustainable options of green energy, green transport, green innovation and green
463 standards, which are in line with the initiative taken by Ministry of Ecology and Environment of

464 China. Present study can provide a strong justification of sustainable economic growth for the
465 policy makers of BRI partner countries while keeping in mind the environmental implications.
466 The transfer of technology between the partner countries can also help to transform the economic
467 growth into an energy efficient and sustainable development.

468 The estimates of recent study suggest some essential policy implications for lawmakers
469 and environmental experts. They must allocate economic resources based on the results of the
470 study to maximize productivity, but wisely. As a result, researchers will take short- and long-
471 term approaches to environmental issues, in particular the involvement of greenhouse gases
472 (GHGs) and the BRI economy's climate change sensitivity. This shows that continued economic
473 expansion is the key to improving the quality of the environment. Thus, for all regions tested to
474 reduce CO₂ emissions, more practical and stringent policies / strategies are needed from decision
475 makers and stakeholders. In addition, the various estimates of the current study are a useful tool
476 for developing renewable energy supply strategies to avoid the risk of (GHG) emissions not only
477 for the BRI partnered nations but it will be great gadget for larger countries of the world. It is
478 also important to anticipate demand and supply of energy to achieve the development of BRI
479 projects. In addition, improved GDP per capita (income) will allow the general public with the
480 provision of more dynamic and environment friendly services. Therefore, it is also important for
481 policy makers to incentivize and appreciate investors for green investment and inform them
482 about its benefits.

483 Additionally, researchers can modify variables that may produce points that can further
484 help to improve the understanding about the impacts of BRI projects investments on
485 Environment in general and on the regional climate in particular. In addition, we could measure
486 the relationship of energy and economic growth indicators with many other climate change and

487 environment related indicators, like natural disasters, global warming, oxides of nitrogen, oxides
 488 of sulfur, Carbon Monoxide, industrial pollution and health effects, in order to obtain an overall
 489 environmental impression.

490 **Appendix A**

491 **Table A1:** List of Selected BRI countries

S.No	Countries	S.No	Countries
1	Albania	26	North Macedonia
2	Armenia	27	Mongolia
3	Azerbaijan	28	Moldova
4	Bahrain	29	Maldives
5	Bangladesh	30	Malaysia
6	Belarus	31	Myanmar
7	Bosnia and Herzegovina	32	Nepal
8	Bulgaria	33	Oman
9	Cambodia	34	Pakistan
10	Colombia	35	Philippines
11	Croatia	36	Romania
12	China	37	Russian Federation
13	Czech Republic	38	Poland
14	Egypt, Arab Rep.	39	Saudi Arabia
15	Georgia	40	Singapore
16	Hungary	41	Slovak Republic
17	India	42	Sri Lanka
18	Indonesia	43	Thailand

19	Iran, Islamic Rep.	44	Turkey
20	Israel	45	Tajikistan
21	Jordan	46	Ukraine
22	Kazakhstan	47	Yemen, Rep.
23	Kyrgyz Republic	48	United Arab Emirates
24	Kuwait	49	Vietnam
25	Lebanon		

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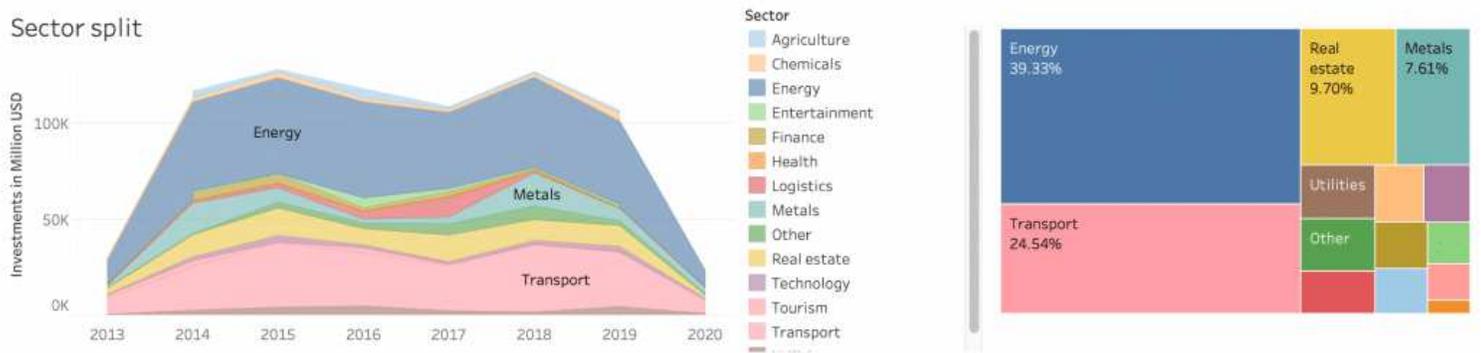
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Figures

Chinese investments in Belt and Road Initiative (BRI) countries 2013- H1 2020 (million USD)



Sector split



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Data: American Enterprise Institute (AEI), China Investment Tracker, 2020

Figure 1

Investments of China in BRI countries from 2013-H12020 (million USD) 1 Note: The designations employed and the presentation of the material on this map do not imply the expression of any opinion whatsoever on the part of Research Square concerning the legal status of any country, territory, city or area or of its authorities, or concerning the delimitation of its frontiers or boundaries. This map has been provided by the authors.