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## **Does biomass energy drive environmental sustainability? An SDG perspective for top five biomass consuming countries**

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1 **Does biomass energy drive environmental sustainability? An SDG**  
2 **perspective for top five biomass consuming countries**

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24 **Abstract**

25 Efficient use of biomass energy is integral to achieving many of the Sustainable Development Goals  
26 (SDGs). Their contributions, trade-off patterns, and implementation vary geographically, requiring in-  
27 depth analysis to sustainably manage its impact. Here, we analyzed the contribution of biomass energy  
28 intensity and efficiency on sustainable development across the top five biomass energy-consuming  
29 countries—Brazil, China, Germany, India, and the US. We compared the impact of biomass energy  
30 consumption, economic development, urbanization, and trade openness on carbon dioxide emissions  
31 and ecological footprint. Using annual frequency data from 1970 to 2016, we utilized continuously-  
32 updated fully-modified, and continuously-updated bias-corrected panel estimation techniques that  
33 allow controlling of cross-section dependence among sampled countries. Our empirical analysis shows  
34 income level escalates ecological footprint and emissions by 0.05-0.21%. Similarly, urban sprawl  
35 increases long-term emissions and ecological footprint by 0.07-0.17%. Biomass energy consumption  
36 increases ecological footprint by 0.18-0.90% but declines emissions by 0.02-0.09%. However, trade  
37 openness reduces both ecological footprint and CO<sub>2</sub> emissions by 0.34-0.55%. Our results reveal that  
38 income level stimulates biomass consumption in early stages of growth, but declines in technologically  
39 oriented industrial-based economy, however, outgrows in service-inspired economy. This highlights  
40 that biomass extraction in developed countries can surpass regenerative capability, necessitating  
41 sustainable domestic material consumption management.

42

43 **Keywords:** Biomass energy; domestic material consumption; economic growth; trade openness;  
44 urbanization

45

## 46 **1. Introduction**

47 Natural resource extraction and material flow remain the heartbeat of production-based economies.  
48 However, the nature of extraction, production, and consumption determine its impact on  
49 environmental sustainability. Thus, accounting for domestic material consumption is a useful tool in  
50 assessing material footprint and natural resource security [1]. Domestic material consumption typically  
51 encompasses biomass, fossil fuels, metal ores, and nonmetal ores. Though fossil fuel sources are finite  
52 whereas renewable energy resources are infinite but remain the global economic powerhouse —  
53 driving the world’s economic growth through production and consumption [2]. Meanwhile, the  
54 current and potential future fluctuations in energy security and climate change would require the  
55 adoption of clean and renewable energies to safeguard the environment and livelihoods [3]. Thus,  
56 renewable energy development, use, and economic growth are some of the pressing tri-variate nexuses  
57 in the climate change discourse and sustainable development agendas [4]. The heterogeneity in the  
58 socio-economic and geographical dimensions in the development and use of renewable energy in an  
59 integrated system of future energy supply is poorly understood [5]. These disparities have incited a  
60 renewed opportunity for studying the contribution of renewable energy to the sustainable  
61 development agenda in an energy-growth economy.

62 Biomass energy is “any source of heat energy produced from non-fossil biological materials”  
63 [6]. By 2016, biomass energy accounted for 5%, 4%, 11%, 31%, and 21% of the total energy use in  
64 the USA, China, Germany, Brazil, and India, respectively [7]. Bioenergy source is chiefly biofuels,  
65 wood and wood-derived biomass, and municipal waste. It is projected that the global biomass  
66 potential of energy crops would range from 11 EJ (Exajoule) in the sustainable land use scenario in  
67 2020 to 96 EJ in the business-as-usual scenario in 2050. These projections are equivalent to about 2  
68 to 19% of the primary energy demand in 2010 [ $\sim$ 500 EJ] [8]. Despite the potential of bioenergy to  
69 replace traditional fossils, it's generally considered more eco-friendly [9, 10], however, land area

70 requirements for energy crops limit their production. In competing and displacing agricultural and  
71 marginal lands [11, 12], increased biomass energy production and consumption could double the price  
72 of food commodities on the global market [13, 14].

73 In contrast, biomass energy consumption is reported to enhance economic growth and  
74 environmental degradation. From an economic perspective, biomass energy consumption is stronger  
75 for economic development in developing countries compared to developed countries. A short- and  
76 long-run causality analysis indicated that biomass energy supports the growth of countries in economic  
77 transitions [15]. On the other hand, biomass energy use can slow down economic development  
78 depending on the source, nature of the renewable energy, and technology requirements [5, 16]. These  
79 studies resonate further with the idea that optimizing the benefits of wood biomass as a renewable  
80 source of energy could likely reduce its adverse socio-environmental effects. Although partially  
81 significant linkages are observed between GDP and biomass, the inclusion and use of energy-efficient  
82 technologies to reduce the prevailing high energy intensity of output in developing countries including  
83 Nigeria, Burkina Faso, the Gambia, Mali, and Togo [17]. Shocks in the food production system could  
84 alter biomass energy consumption patterns, requiring modernized biomass energy to support and  
85 improve long-term energy use efficiency in developed countries [16].

86 In addition to biomass use, emissions, and environmental quality, discussions in extant  
87 literature include trade openness and urbanization [18, 19, 1]. The openness of trade could have a  
88 positive or negative on environmental performance depending on the economic status of nations and  
89 methodologies employed. Trade openness can augment the production capacities of high exporting  
90 countries and hence their impact on agricultural and marginal lands, forests, and global commodity  
91 markets. Thus, technological spillover effects of trade openness occur through export activities which  
92 reduce the EF in the long run [20]. For instance, trade openness was found to intensify ecological  
93 degradation in the Middle East and North African nations between 1996–2012 and in 93 countries

94 between 1980-2008 globally [21, 22]. Trade openness was found to substantially reduce the ecological  
95 degradation of 24 OECD countries between 1980 to 2014 using panel methodologies [23]. The  
96 increasing urban sprawl means a rise in the demand for resources which would require more  
97 development of new areas for housing, social amenities, commercial and other urban land uses [24].  
98 Yet empirical studies have reported tentative results. Thus, urbanization exacerbates environmental  
99 degradation through its positive effect on the ecological footprints of lower-middle-, upper-middle-  
100 and high-income countries, including changes in urban domestic sewage, industrial effluent, and solid  
101 waste [22].

102 The motivation of this paper is to investigate the combined impacts of biomass energy  
103 consumption and economic growth on environmental quality using ecological footprint and carbon  
104 emissions. We test the hypothesis that biomass energy utilization does not affect wealth. This paper  
105 augments the existing consensus on biomass-environmental quality relationships and their potential  
106 impacts on sustainable development goals. Thus, assessing the impacts of biomass energy on carbon  
107 emissions and environmental performance in high consuming nations (Brazil, China, Germany, India,  
108 and the US) are crucial to informing policies on the development of efficient renewable energy  
109 technologies, which reduces the energy footprint of these nations while enhancing development. The  
110 innovation of this study is the inclusion of interaction between economic growth and bioenergy  
111 consumption indicators to account for the combined impact on environmental quality and the SDGs  
112 (sustainable development goals). To the best of our knowledge, no study has informed environmental  
113 policies from this perspective. Besides, we ascertain the connection between environmental impacts  
114 of biomass consumption and country-specific stages in development, viz. structural transformation  
115 processes. As a result, we examine the possibility of a parabolic relationship between country-specific  
116 income levels and biomass energy use.

117

118 The remaining sections of the study are organized as follows: Section 2 “*Materials & Method*” outlines  
119 the empirical strategy used for model estimation; Section 3 presents the results of the parameter  
120 estimation; Section 4 presents a discussion of the results while Section 5 summarizes the study  
121 findings.

122

## 123 **2. Materials & Methods**

### 124 **2.1. Data**

125 Based on our main hypothesis and review of the previous studies, we construct two different models  
126 to categorize the environmental degradation as carbon dioxide emission and ecological footprint by  
127 incorporating the impact of economic growth, biomass energy consumption, urbanization, and trade  
128 openness. The first empirical model that constructed to observe the impact of biomass usage on  
129 carbon emissions is as follows:

$$130 \quad CO_2 = f(GDP, BIO, URB, TR) \quad (1)$$

131 whereas the second empirical model which is constructed to check the impact of biomass usage on  
132 ecological footprint is expressed:

$$133 \quad EF = f(GDP, BIO, URB, TR) \quad (2)$$

134 where CO<sub>2</sub> is carbon emissions to represent the first degradation indicator and measured in per capita  
135 carbon emissions in tons, EF is ecological footprint as a proxy for second degradation indicator and  
136 measured in per capita ecological footprint and measured in gha, GDP refers to economic growth and  
137 measured as per capita real gross domestic product in 2010 constant US dollars, BIO is per capita  
138 biomass energy and measured as biomass extraction in tons, URB is urbanization level and measured  
139 in urban population % share in total population, and TR is trade openness and measured as total trade

140 (sum of export and import) % share in gross domestic product. The model specification of Equation  
141 (2) using ecological footprint is a more comprehensive indicator of environmental degradation. The  
142 ecological footprint indicator includes different sub-dimensions including cropland, grazing land,  
143 fishing grounds, and forest land. Therefore, analysis within the scope of Model 2 with biomass energy  
144 consumption is essential in assessing specific targets of the Sustainable development goals—including  
145 responsible consumption and production (SDG 12), Life below Water (SDG 14), and Life on Land  
146 (SDG 15).

147 All variables are used for the empirical analyses were in natural logarithmic form for the annual data.  
148 The temporal series of our data were limited to the period 1970-2016 because of data availability for  
149 Brazil, China, Germany, India, and the US. The dataset for CO<sub>2</sub> emissions was retrieved from  
150 OurWorldInData of Ritchie and Roser [25], whereas data for EF was obtained from Global Footprint  
151 Network. The dataset for BIO was retrieved from the Global Material Flows Database. Data for URB  
152 and TR were downloaded from the World Development Indicators of the World Bank. For empirical  
153 analysis, we utilized the Gaussian software.

154

## 155 **2.2. Empirical Strategy**

### 156 **2.2.1. Preliminary Tests**

157 For the estimates to be reliable and consistent for policy suggestions, it is crucial to select appropriate  
158 estimators for the model and to perform some pre-tests. In panel data analysis, the first of these pre-  
159 tests is the cross-sectional dependency test, which examines the shock permeability between cross-  
160 sections (countries in our case). In this line, we used the CD test with the null hypothesis of no cross-  
161 sectional dependence developed by Pesaran [26] to test the cross-sectional dependency. The next  
162 important issue is to examine the stationarity process of the variables. For the stationary test, it is



163 necessary to decide the suitable unit root test based on the result of the cross-section dependence  
 164 tests. Therefore, under a null hypothesis of unit root, the CIPS panel unit root test Pesaran [27] was  
 165 performed. After observing the stationary properties of variables, we employ the ECM-based panel  
 166 cointegration method [28] with the null hypothesis of no cointegration. This cointegration test also  
 167 allows cross-section dependence among observed countries. The other reason for choosing this  
 168 cointegration test is that using this methodology is one of the most suitable tests for our empirical  
 169 model because Westerlund [28] argues that the error-correction-based test shows better accuracy than  
 170 residual-based cointegration test in a situation where the explanatory variables are weakly exogenous.

171

### 172 **2.2.2. Panel Cointegrated Regressions**

173 To validate cross-sectional dependent cointegration among variables, the coefficient of cointegrated  
 174 regressor is used to search for an estimation technique that allows cross-sectional dependence. Thus,  
 175 we conduct CUP-FM (continuously-updated and fully-modified) and CUP-BC (continuously-updated  
 176 and bias-corrected) estimators developed by Bai et al. [29]. These estimators augment the basic panel  
 177 regression model and assume cross-sectional dependence and error term ( $\varepsilon_{it}$ ) e.g. Bai and Kao [30]  
 178 as follows:

$$179 \quad y_{it} = a_i + \beta x_{it} + \varepsilon_{it} \tag{3}$$

$$180 \quad \varepsilon_{it} = \lambda_i' F_t + \mu_{it} \tag{4}$$

181 where  $F_t$ ,  $\lambda_i'$  and  $\mu_{it}$  indicate the vector of common factors, corresponding factor loadings, and the  
 182 idiosyncratic component of the error term, respectively. The computation process of CUP-FM is  
 183 based on repeatedly estimating coefficients and long-run co-variance matrix until reaching the  
 184 convergence as follows:

$$\hat{\beta}_{Cup} = \left[ \sum_{i=1}^N \left( \sum_{t=1}^T \hat{y}_{it}^+(\hat{\beta}_{Cup})(X_{it} - \bar{X}_i)' - T \left( (\lambda_i'(\hat{\beta}_{Cup}) \hat{\Delta}_{F\epsilon i}^+(\hat{\beta}_{Cup}) + \hat{\Delta}_{\mu\epsilon i}^+(\hat{\beta}_{Cup})) \right) \right) \right] \times$$

$$\left[ \sum_{i=1}^N \sum_{t=1}^T (x_{i,t} - \bar{X}_i)(x_{i,t} - \bar{X}_i)' \right]^{-1} \quad (5)$$

where  $\hat{y}_{it}^+ = y_{it} - (\hat{\lambda}_i' \hat{\Omega}_{F\epsilon i} + \hat{\Omega}_{\mu\epsilon i}) \hat{\Omega}_{\epsilon i}^{-1} \Delta X_{it}$ ,  $\hat{\Omega}_{F\epsilon i}$  and  $\hat{\Omega}_{\mu\epsilon i}$  are estimated long-run co-variance matrices and  $\hat{\Delta}_{F\epsilon i}^+$  and  $\hat{\Omega}_{\mu\epsilon i}$  are estimated one-sided long-run co-variance.

There are also some reasons for using the CUP-FM and CUP-BC estimators in this study. First, similar to our preferred cointegration test, these estimators are also consistent tests in the case of exogenous explanatory variables. Also, these estimators can be used for variables that are integrated of different orders. Moreover, since the CUP-FM estimator is a test developed based on the fully modified OLS estimator which uses the Bartlett-Kernel procedure, especially it can also be used in possible autocorrelation and heteroskedasticity situations (Kiefer and Vogelsang [31]; Khan and Ulucak [32]. Finally, both estimators are robust in the case of endogeneity [29].

### 3. Empirical Results

#### 3.1. Descriptive analysis

Descriptive statistical analysis is critical to understanding the characteristics of data series. The maximum per capita carbon dioxide emission, ecological footprint, and GDP were 22.123 tons, 11.097 gha, and US\$ 52,534.370, which is equivalent to the environmental degradation and economic status of the US in 1973, 1973, and 2016, respectively (Table 1). The share of trade in economic development was highest in Germany among other four countries (86.514 % of GDP in 2012). Per capita biomass energy usage was highest (BIO = 12.633 tons) in 2016 for Brazil. CO<sub>2</sub>, EF, GDP, and URB exhibit platykurtic distribution whereas BIO and TR exhibit leptokurtic distribution. All variables except URB are positively skewed; BIO has the highest skewness. The Jargue-Bera statistic shows that all the

207 variables are not normally distributed — requiring logarithmic transformation during the empirical  
208 analysis to provide a more stable data variance.

209 **[INSERT TABLE 1 HERE]**

210

### 211 **3.2. Conditional panel-based tests**

212 First, we used Pesaran’s CD test to control for the presence of shock-dependency among observed  
213 countries (Table 2). The tests strongly rejected the null hypothesis of no cross-sectional dependence  
214 (CSD) for all variables, indicating the importance of CSD due to globalization in our country-based  
215 panel data analysis. It was therefore imperative to account for these CD shocks in the panel  
216 methodologies.

217 Next, we examined the unit root process of the variables at level and first-difference to determine the  
218 order of integration (Table 2). The null hypothesis of a unit root process cannot be rejected, indicating  
219 all the variables are non-stationary at level form. However, variables are deemed stationary in first  
220 differenced forms. These findings were a point of reference for cointegrating the relationship between  
221 variables for both models.

222

223 **[INSERT TABLE 2 HERE]**

224

225 The results of ECM-based cointegration test for the existence of a long-run relationship between  
226 variables for each model are shown in Table 3. For the CO<sub>2</sub> model, the test statistics for  $G\alpha$  and  $P\alpha$   
227 were -10.873 ( $p < 0.01$ ) and -6.711 ( $p < 0.05$ ) respectively, rejecting the null hypothesis of no  
228 cointegration. Similarly, the null of no cointegration is rejected for the EF model given  $G\tau$  (-6.83,

229  $p < 0.05$ ),  $G\alpha$  (-15.06,  $p < 0.01$ ), and  $P\alpha$  (-18.92,  $p < 0.01$ ). Therefore, we examined the long-run  
230 parameters of economic development, per capita biomass energy, trade, and urbanization.

231

232 **[INSERT TABLE 3 HERE]**

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234

### 235 **3.3. Drivers of Ecological Footprint and CO<sub>2</sub> emissions**

236 The results of the long-run impact of biomass consumption on environmental degradation indicators  
237 are summarized in Table 4. In the CO<sub>2</sub> emissions function (Table 4), we observe increasing effects of  
238 income level escalates atmospheric emissions by 0.08-0.21% —confirming the scale effect hypothesis.

239 As hypothesized, economic development significantly increases emissions in the top five biomass-  
240 consuming countries. While the long-term effect of urbanization on emissions is insignificant and

241 unnoticeable, the incorporation of interaction effect of income and biomass consumption stimulates  
242 urban sprawl to trigger CO<sub>2</sub> emissions by 0.09-0.13%. In contrast, 1% increase in trade openness

243 reduces carbon market failures, thus, reducing long-term emissions by 0.34-0.55% across sampled  
244 countries. Similarly, increasing consumption of biomass energy by 1% spur CO<sub>2</sub> emissions by 0.18-

245 0.90%. Biomass energy usage seems efficient on carbon mitigation, hence, increasing biomass energy  
246 consumption substantially reduces carbon dioxide emissions. The interaction between GDP and

247 biomass energy consumption reduces long-term emissions by 0.18-0.26%. These findings are  
248 consistent in both estimation strategies.

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**[INSERT TABLE 4 HERE]**

The empirical results in Table 5 show income level increases ecological footprint by 0.05-0.09% across sampled countries for both estimators. This perhaps occurs in linear economies with dependence on natural resource-extraction and limited circular economic structure. Similarly, increasing trade openness by 1% reduces ecological footprint by 0.27-0.72%, implying long-term ecological reserve. Surprisingly, unlike the CO<sub>2</sub> emissions model, increasing biomass energy consumption by 1 % harms environmental quality (ecological footprint in our case) by 0.18-0.90%. Besides, growth in urbanization increases ecological footprint of sampled countries by 0.07-0.17%. The interaction between GDP and biomass consumption increases ecological footprint by 1.09-1.11% in both estimators.

To corroborate the estimated panel models, we investigated the country-specific nexus between biomass energy consumption and income level using time series based on higher-order regression. While our panel models account for global common shocks and spillover effects, divergence in economic structure across sampled countries may hamper environmental convergence. In this regard, utilizing country-specific models is essential to account for country-specific dynamics. Based on top-bottom estimation approach, we used third-order polynomial of income level to account for complexities in biomass utilization. This scenario helps in assessing whether wealth influences the consumption of biomass across different income groups presented in Table 6. The resultant structural assessment and its predictive power are depicted in Figures 1-5. All the estimated models were statistically significant at  $p\text{-value} < 0.001$ . The goodness of fit test (R-square) reported 99% predictive power for China (Figure 2), 96% for Brazil (Figure 1), 68% for India (Figure 4), 61% for Germany (Figure 3), and 53% for the US (Figure 5). It can be observed in Figures 1-5 that while Brazil exhibits inverted-N-shaped relationship, China, India, Germany, and the US exhibit N-shaped relationship. The parameter estimation of Biomass-Wealth nexus in Table 6 reveals that growth in income declines biomass energy consumption at the initial stages of development in Brazil, but outgrows in industrial-

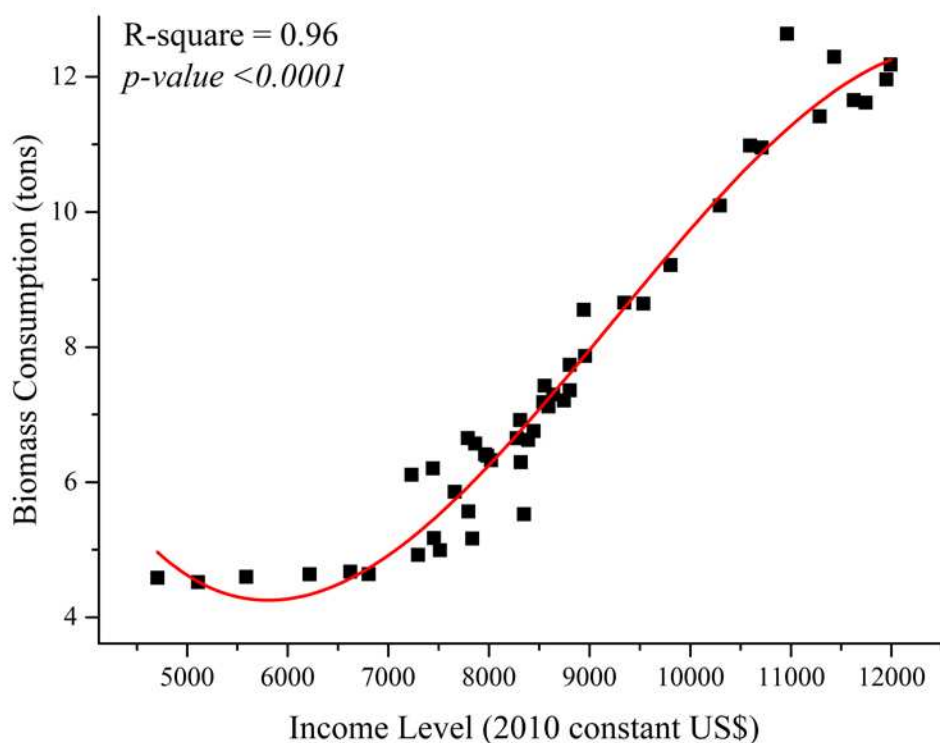
273 based economy and declines thereafter in a service-dominated economy as argued by Ref. [53]. In  
274 contrast, increasing levels of income spur biomass energy consumption at the initial developmental  
275 stage in China, India, Germany, and the US but declines in the technologically-driven industrial-based  
276 economy and outgrows afterward in a service-inspired economy. The residual plots to validate the  
277 higher-order regression estimates are presented in Appendices A-E, confirming the independence of  
278 the residuals and stability of the estimated parameters.

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**[INSERT TABLE 5 HERE]**

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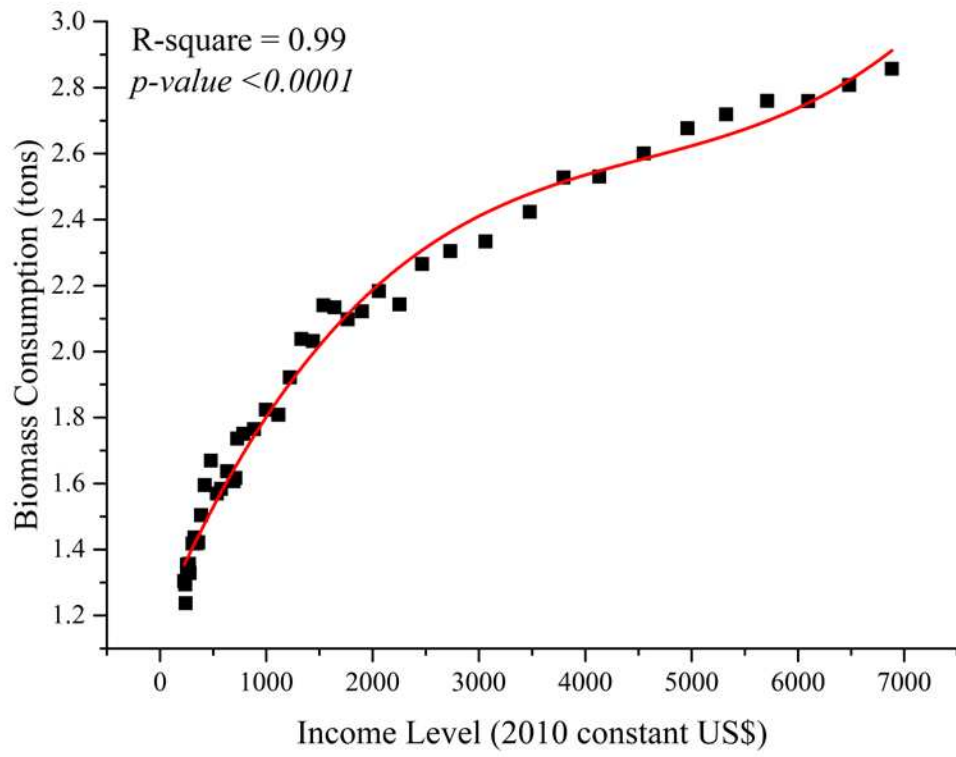


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Figure 1. Biomass Consumption—Wealth Nexus in Brazil

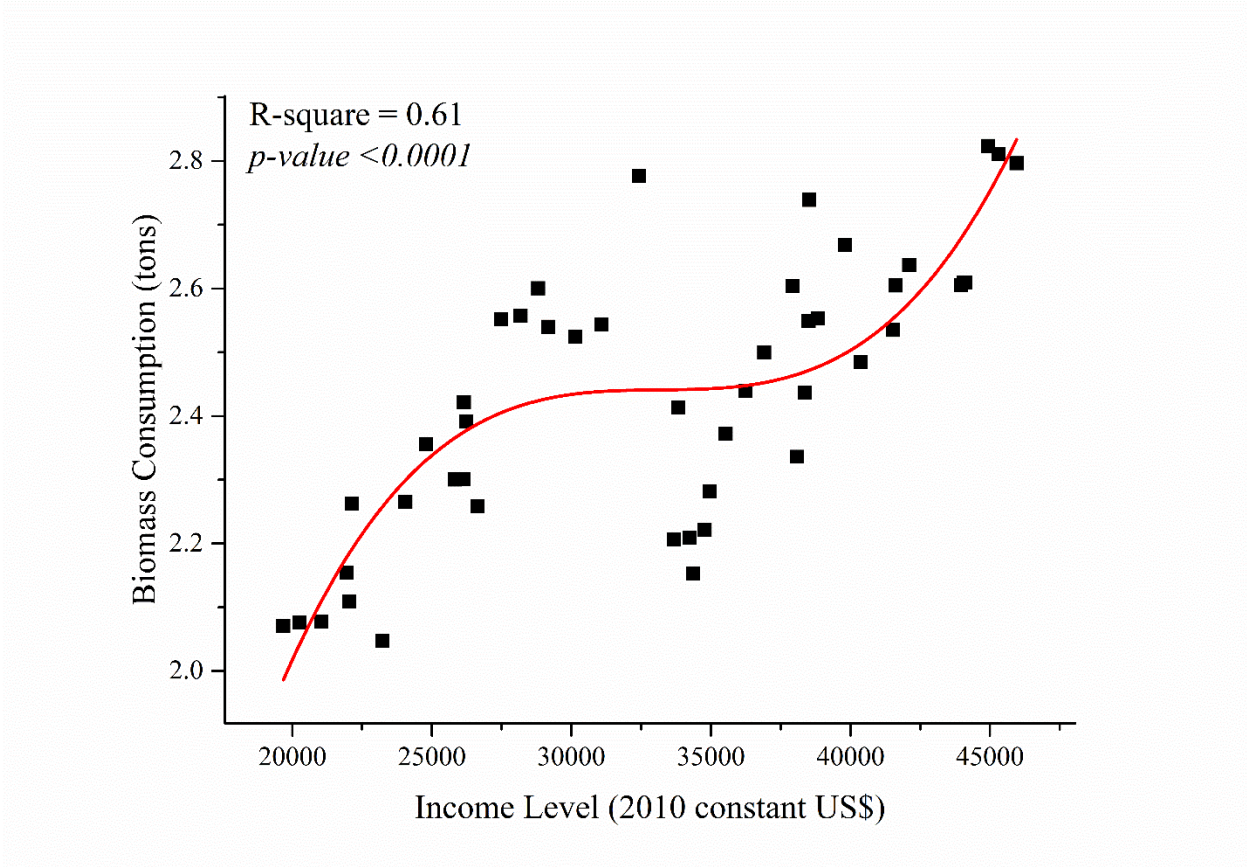
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Figure 2. Biomass Consumption—Wealth Nexus in China

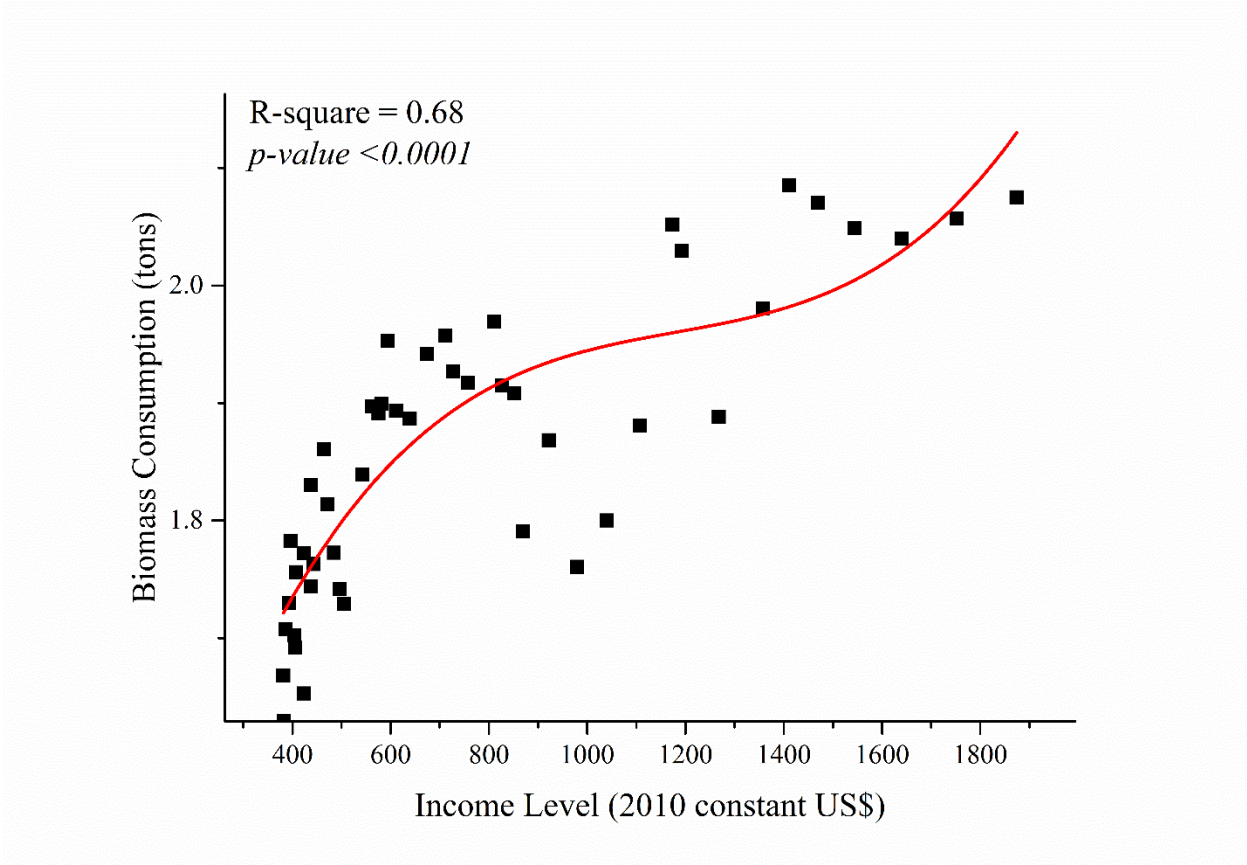


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Figure 3. Biomass Consumption—Wealth Nexus in Germany

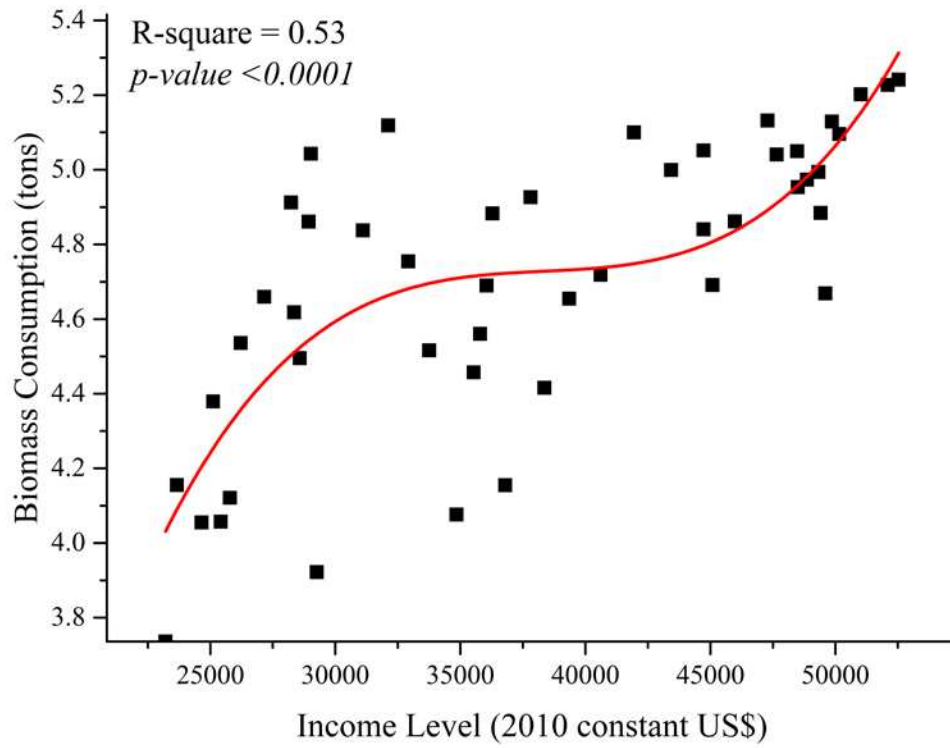




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Figure 4. Biomass Consumption—Wealth Nexus in India



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Figure 5. Biomass Consumption—Wealth Nexus in the US

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[INSERT TABLE 6 HERE]

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**4. Discussion**

This study compares the effect of biomass energy consumption on environmental degradation and quality. The finding that biomass energy consumption reduces carbon emissions in high biomass-consuming countries could indicate the importance of biomass as a useful tool to combat atmospheric pollution and subsequently climate change. These findings are indicative of the need for policymakers to increase the share of biomass energy in the total energy portfolio, which is integral for achieving the climate action objective of the sustainable development goal thirteen (SDG 13). The finding is consistent with similar studies by Bilgili et al. [33]; Danish and Wang [18]; Dogan and Inglesi-Lotz [34]; Sarkodie et al. [16]; Shahbaz et al. [35] and Destek and Aslan [36]. On the other hand, our finding reveals that increasing biomass energy consumption increases the ecological footprint of biomass-consuming countries. Environmental degradation metric considers more than atmospheric pollution. The results of biomass energy consumption—ecological footprint nexus reveal that biomass is not eco-friendly. Increasing biomass energy consumption directly reduces atmospheric pollution levels but leads to the deterioration of cropland, grazing land, fishing grounds, and especially, forest land. Even more unfavourable, the harmful effect of biomass energy on these ecological indicators is above the positive atmospheric impact. These findings align closely with sustainable development goals. This confirms that biomass energy consumption is an obstacle to achieving the objective of responsible consumption and production (SDG 12), Life below Water (SDG 14), and Life on Land (SDG 15).

Our study further raises an interesting question on how biomass energy consumption drives environmental sustainability while developing economically. The finding that interaction between biomass consumption and economic development is negative on carbon dioxide emissions suggests

327 that biomass energy could enhance the environmental sustainability parallel to the economic  
328 development trajectory of the US, China, Germany, India, and Brazil. Aligning biomass-based  
329 environmental sustainability to economic development is indicative that the development of biomass  
330 energy infrastructure and biomass energy consumption can proceed to support the environment while  
331 economic growth ensues [15]. Meanwhile, economic growth exacerbates the effects of biomass energy  
332 consumption on ecological footprints in the long run.

333  
334 The source and sink hypothesis in biomass production and consumption can explain why bioenergy  
335 consumption can increase the ecological footprint while reducing carbon dioxide emissions. It is well-  
336 known that burning fossil fuels and traditional biomass increases carbon dioxide emissions. When  
337 energy crops are fully grown, almost equivalent amounts of carbon dioxide are captured through  
338 photosynthesis. Biomass energy consumption reduces carbon dioxide emissions because the rate of  
339 renewal of plants as biomass energy resources may be higher than the rate of utilization. It is reported  
340 that biomass-derived from biological sources such as agricultural, wood, and animal husbandry  
341 residues can substantially reduce anthropogenic emissions and reduce the competition of land use  
342 [37]. The increasing effect of biomass energy consumption on ecological footprint can be attributed  
343 to the weighted share of biomass energy consumption from traditional sources (wood, animal waste,  
344 and traditional charcoal). The increase in modern biomass energy (liquid biofuels, bio-refineries, and  
345 biogas) consumption could account for the decreasing share for solid biomass in recent years, hence,  
346 their consumption declining carbon dioxide emissions [37]. However, the slow rate of conversion  
347 from traditional biomass consumption to modern resources is one of the most important reasons for  
348 the increase in ecological footprint, but not accelerating this transformation may also lead to  
349 atmospheric damage. Similarly, if the destruction of forests continues at this pace to produce energy  
350 crops that only increase the ecological footprint, the atmospherically positive picture may reverse due

351 to deforestation. Awareness of responsible land use should be increased to alleviate these adverse  
352 effects of bioenergy production and consumption.

353 The finding that economic development increases carbon dioxide emissions and decreases ecological  
354 quality is consistent with similar studies [38; 39; 40; 41; 42; 43; 44]. The increasing effects of income  
355 level on atmospheric emissions confirm the scale effect hypothesis. The scale effect postulates  
356 economic development driven by environmental degradation, viz. natural resource exploitation, waste  
357 generation, and emissions [54]. While developed countries may limit environmental pollution through  
358 innovation and technological advancement, emissions could still be imported into wealthy countries.  
359 The transboundary effect of emissions through spillover effects of goods and services from  
360 developing countries (i.e., China, and India) could trigger a rise in emissions. The production structure  
361 of the US, China, Germany, India, and Brazil are mainly dependent on fossil-fuel energy sources. The  
362 conversion to clean energy resources is not sufficiently achieved in such production structures. Hence,  
363 increasing the share of clean energy sources in production activities will eliminate the negative impact  
364 of economic development on environmental quality.

365 Besides, our finding that trade openness reduces environmental degradation is consistent with the  
366 studies of Dogan and Turkekul [45]; Zhang et al. [46]; Gozgor [47]; Shahbaz et al. [48] and Destek  
367 and Sinha [23]. This finding is possibly sourced from our sampled country group consisting of middle-  
368 income and high-income countries. Increasing trade openness reduces both ecological and emission  
369 levels, hence, improving ecological reserves and environmental quality. It is well-known that more  
370 high-income countries have implemented pollution-reducing trade measures compared to developing  
371 countries with lax trade regulations. Thus, trade openness removes market barriers, hence, increases  
372 patronization of green trade and innovation that may serve as abatement technologies with long-term  
373 emission-reduction effects. Besides, trade openness improves natural resource market competition  
374 and drives green technology and innovations that find artificial alternatives to natural resources—

375 which could limit anthropogenic emissions. The finding that urbanization increases ecological  
376 footprint is consistent with Sarkodie et al. [49], pointing out that urbanization is particularly harmful  
377 to agricultural lands and water resources of observed countries.

378 On the nexus between biomass energy consumption and income level, our study confirmed that while  
379 biomass utilization decreases with increasing income level in Brazil, strong evidence that wealth  
380 increases biomass energy consumption is confirmed in the US, Germany, China, and India. This  
381 means that modern biomass resource consumption, as a supply chain of increased service is triggered  
382 by population demand in wealthy countries. The use of traditional biomass for cooking and heating  
383 purposes is reported to be rampant in developing countries with high multi-dimensional poverty [50].  
384 It is reported that over 38% of the World's population from poor countries depends on traditional  
385 solid biomass [51]. Biomass resource consumption is mediated by resource extraction either through  
386 legal or illegal logging. It is reported that illegal logging of forest products—a source of biomass often  
387 occurs in developing countries and is driven by market pressure. For example, the market demand for  
388 the endangered rosewood species is reported to have triggered illegal logging, which in effect hampers  
389 ecosystem biodiversity [52]. Thus, export-driven biomass resource extraction from developing  
390 countries may explain the consistent use of modern biomass in wealthy countries.

391

## 392 **5. Conclusion**

393 This study explored the impact of biomass energy consumption on both carbon emissions and  
394 ecological footprint by incorporating economic growth, trade, and urbanization in top five biomass-  
395 consuming countries (Brazil, China, Germany, India, and the US). First, we tested the hypothesis that  
396 biomass utilization does not affect emissions and ecological footprint. Second, we hypothesized  
397 wealth does not underpin biomass energy consumption. To observe how biomass energy affects  
398 environmental degradation indicators, we used annual data from 1970-2016 and panel data techniques

399 that control for cross-sectional dependence across the sampled countries. Our empirical analysis  
400 demonstrated that increasing income level escalates emissions by 0.08-0.21%. While the effect of  
401 urban sprawl on CO<sub>2</sub> emissions was insignificant, the inclusion of interaction effect of income and  
402 biomass energy consumption causes urban sprawl to increase long-term CO<sub>2</sub> emissions by 0.09-0.13%.  
403 However, trade openness reduces CO<sub>2</sub> emissions by 0.34-0.55%. Likewise, increasing energy  
404 consumption stimulates CO<sub>2</sub> emissions by 0.18-0.90%.

405 The key empirical findings showed increasing biomass energy consumption is conducive to expanding  
406 the ecological footprint. In contrast, our study found biomass energy consumption as efficient tool  
407 for carbon mitigating policies. While economic growth and urbanization were found to deteriorate the  
408 environment, the mitigation effect of trade openness improves environmental quality. We observe  
409 that a shift in biomass consumption patterns has additional benefits of moderating the impact of  
410 economic development on environmental sustainability. In the context of policy implication, our  
411 results show that focusing only on one goal in the implementation of policies to achieve sustainable  
412 development targets may be an obstacle in attaining other targets. As observed, the leading countries  
413 in biomass consumption have implemented biomass policies with a focus on reducing atmospheric  
414 pollution, however, biomass consumption-led environmental damage has been neglected. Therefore,  
415 accelerating and managing the transformation from the use of traditional biomass to modern biomass  
416 could improve the green effects of biomass consumption, hence, reducing environmental  
417 deterioration. This transformation could improve energy efficiency attributable to biomass energy  
418 production, thus, allowing for more possibility to renew forest lands. Besides, a possible introduction  
419 of policies and measures that enable facilities operating in modern biomass industries will limit the  
420 exploitation of endangered biomass resources. Similarly, to prevent the destruction of agricultural  
421 lands, awareness-raising activities on the effective use of agricultural lands could be instituted. Due to

422 the limitation of sampled countries, future research could examine the global perspective of the theme  
423 by expanding the sample size.

424

#### 425 **Data Availability**

426 Sources to data used for the model estimation have been correctly specified in the data sub-section of  
427 Materials and Method.

428

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