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# Green growth and sustainable energy transitions: evaluating the critical role of technology, resource efficiency, and innovation in Europe's low-carbon future

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Renewable energy, technology, and innovation are central to sustainable development, but their effects vary by country because of institutional and economic differences. Previous studies often ignore these differences. By incorporating institutional and economic differences, our study aligns with a growing body of research emphasizing the importance of context in evaluating sustainable development strategies. This nuanced approach allows us to contribute to the ongoing debate by offering a more comprehensive framework for understanding how renewable energy, technology, and innovation interact with the broader economic and institutional environment to drive sustainable outcomes. This study provides a comprehensive analysis to investigate the determinants of green growth in European economies with changing income (higher, upper middle, and lower income) levels. For this study, we have taken data from 1990 to 2023 and employed the Cross Sectional Autoregressive Distributed Lag model. The empirics show that renewable energy consumption and economic growth significantly enhance the green growth across all income groups. One percent increase in renewable energy consumption leads to increase the green growth by 0.1488, 0.0965, and 0.1613 percent respectively while one percent increase in GDP leads to increase green growth by 0.24, 0.26, and 0.32 percent for higher, upper-middle and lower middle-income countries respectively in long run. Nevertheless, technological innovation and globalization show mixed results, being more effective in higher income economies due to stronger institutions and green innovations. More precisely, one percent increase in technological innovation leads to rise green growth by 0.0762 percent in higher income economies while it shows insignificant result for upper and lower middle-income economies. On the other hand, one percent increase in globalization leads to decrease green growth by  $-0.1439$  percent in upper middle income only while it shows insignificant result for high and lower middle-income economies. Furthermore, Natural resource rents generally enhance green growth particularly in lower income economies, supporting resource curse hypothesis. One percent increase in natural resource rents leads to rise green growth by 0.0159 percent in lower middle income. The findings emphasize the need for income specific state policy that promote sustainable development. The policymakers should focus on clean energy investment, green innovation and strengthen institutional frameworks to ensure sustainable green growth.

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## Introduction

**Background of the study.** Sustainable development (SD) enables the nations to survive in the long term. Sustainability empowers the economies to meet their objectives that involve protecting the environment, safeguarding the society and growing the economy. Over approaches that are equally stable to the environment, sustainability benefits the economy all at once and cares about the well-being of society. The world is currently dealing with several issues, but the most significant ones that are also mentioned in the Sustainable Development Objectives are poverty, inequality, injustice, and global climate change (Bai et al., 2022). According to the UN 2030 Agenda of Sustainability, the three main pillars of Sustainable Development are economy, environment & society. Every pillar of SD supports the others (Sharpley, 2020; Chien, 2022). Often, achieving high growth and sustainability are seen as two different goals, especially when technological advancement is thought to have negative ecological repercussions (Hübler et al., 2012). Information and communication technology stimulates productivity and energy efficiency in the digital phase.

The environmental problem of employing technology sets is thought to be solved by innovative technology utilization, especially in the energy industry (Al-Mulali et al., (2015)). Technological developments assist in pointing out environmental issues like greenhouse gas emissions and spur eco-innovation in the nation. The nation is moving towards sustainable development because of the environment as well as the resources saved from greenhouse gas emissions (Sadiq et al., (2023); Sadiq et al., (2023) and Yuan and Zhang 2020). In the fresh years, Information and communication technology in BRICS nations is also growing fast, that has a great influence on its development. According to Franceschini and Pansera (2015), the debate of sustainable development restored the certainty that technological advancements are essential to maintaining both environmental sustainability & Economic Growth (EG). In the outcome, awareness has been raised regarding environmental issues (cleaner technologies are one example) which Paredis (2011) views as means of addressing both environmental & development issues simultaneously. Technological innovation has an instrumental function in the economic, social and environmental spheres which makes it a crucial component of sustainable development (Fokkema et al., 2005; Constantinescu and Frone 2014). To promote the GG (Green Growth), technology innovation is key in lowering the costs of environmental protection.

Besides technological innovation, renewable energy consumption also plays a driving role for sustainability of nations. One significant instrument for encouraging the growth of renewable energy (RE) is the Sustainable Development Goals (Wang et al., 2022). In fact, the SDGs' seventh goal is to assure that everyone is accessible to low-priced, safe and sustainable energy (He et al., 2022). This indicates that nations, businesses and people are devoted to progressing the growth of sources of renewable energy (RE) including wind, hydropower, and solar (Schwerhoff, Sy (2017). Renewable energy is considered as sustainable & clean energy source that may be used to fulfil energy requirements in countries. Several advantages are provided by renewable energy that can assist nations, corporations, and individuals in achieving the SDGs (Cheng et al., 2021). Choudhury et al., (2023) explore the environmental impact on top polluted economies of the world and found that energy consumption and GDP has positive and statistically significant influence on CO<sub>2</sub> emissions. Akbar et al. (2024) investigated the impact of renewable and non-renewable energy sources on carbon emissions in SAARC countries. The study findings show that non-renewable energy consumption enhances CO<sub>2</sub> emissions while renewable energy consumption

decrease CO<sub>2</sub> emissions. Nosheen et al., (2024) also found that renewable energy is the main source of sustainable development.

Examining the connection between renewable energy (RE) and the SDGs is crucial because this can be helpful for nations in achieving sustainable development and uplifting the living standard for their residents (Boubaker, Omri (2022)). SDGs upholding can be improved by renewable energy in a number of ways. According to (Marco-Lajara et al., (2023)), renewable energy can support in lowering greenhouse gas emissions which will support SDG 13 efforts to alleviate the consequences of climate change. Furthermore, SDG 7th, which demands worldwide access to modern, reasonable, safe, secure & sustainable energy (SE) services, can be aided by renewable energy making energy more reachable to people everywhere. Moreover, by generating both direct & indirect jobs in industries like infrastructure development etc., the transition to renewable energy sources can increase employment. This can help achieve SDG 8, which demands productive & full employment, decent work for all, and sustainable economic growth as explored by (Bertheau, 2020).

Natural resource rent is also important indicators of sustainable development. Ji et al. (2023) explored that efficient use of natural resources is required to promote sustainable and sound financial development. Moreover, natural gas rents and mineral rents positively affect global economic growth and hence lead to boost global sustainability (Fu and Liu, 2023). According to Azam et al. (2023), total natural and oil rents positively contribute to sustainable development, while coal rents have negative and forest rent has an insignificant impact over it. Economic indicator like GDP is also crucial when discussing sustainable development. Sustainable development means steady economic growth that does not reduce natural resources or endanger the ability of future generations to meet their own needs (Erdogan, 2024). The connection between GDP and sustainable development relates to wealth distribution and equity (Shao et al., 2023). Often, inadequate economic growth results in socioeconomic inequalities which can be incompatible with sustainable development. Alsagr and Ozturk (2024), Azam et al. (2023) indicate that economic growth contributes significantly and positively to sustainable development. However, others came up with a conclusion that the effect is negative by worsening environmental quality (Anwar et al., 2022).

The current study is distinguished from existing literature in number of ways. Available research has explored the variables separately or investigate the case of individual countries, however, this study applies the European countries aggregation with panel data methods. Moreover, on the natural resource management side, GHG emissions have been focused by earlier studies which is not the case here in this study. This study takes green growth as a proxy for sustainable development to explore something new. Additionally, the current study along with technological innovation and renewable energy consumption as focused variables, used some controlled variables like GDP, GDP square as a proxy for Environmental Kuznets Curve (EKC), Natural resource rents and Globalization index which is further decomposed into economic and political globalization. The results of all these variables will show how these can drive sustainable development (Green Growth) in European countries.

This study is carried out to answer the essential research question that how do technological innovation and renewable energy consumption contribute to green growth, which is taken as a proxy for sustainable development across diverse European economies? This study aims to explore the composite effect of technological innovation and use of renewable energy on the

sustainable development in case of European economies as a proxy variable by using green growth.

**Motivation of the study.** The motivation for the study stems from increasing policy concern for sustainable development within the framework of environmental challenges and economic transformation in emerging economies. By establishing key drivers of green growth, the results provide policy implications that can inform European countries to develop holistic strategies for sustainable development.

**Contribution of the study.** This study makes three significant contributions to the existing literature. First, it extends the analysis of green growth beyond high-income economies by including upper- and lower-middle-income European countries, allowing for an income-based comparative outlook that is largely omitted by previous studies. Second, it provides empirical evidence on the combined effect of technological innovation, renewable energy consumption, natural resource use and globalization over sustainable development that is measured through green growth. Third, it bridges the gap between environmental economics and innovation policy by examining how these forces interact in shaping low-carbon transitions by offering different intuitions for both policy design and academic debate in the European context.

The remainder of this study is organized as Section “Literature review” covers the relevant literature review including literature gap and theoretical framework in subsections. Section “Data and Methods” covers the data and methods including discussion on variables and econometric methodology. The next section contains estimation results based on econometric tests and techniques with discussion over it. Finally, the last section is conclusion of the study and policy recommendations based on findings.

## Literature review

Studies on green growth and its impact have received significant attention in recent years. Keeping our research objectives under consideration, we have accumulated the summary of research across our major themes. Here we have provided some studies that will help us to identify the potential research gaps.

**Nexus between green growth, sustainable development & natural resources.** Xu et al. (2022) studied the green growth, sustainable development & natural resources for BRICS economies for the time 1991–2014. By using panel data methods like slope heterogeneity, cross sectional dependence & the MM quantile regression model, the authors explored that green growth shows a heterogeneously influence the economic performance in BRICS. Furthermore, green innovation is negatively linked, although renewable energy consumption & natural resources were found to be the leading features of economic performance. Omri (2020) discussed technological innovation and SD in 75 low, middle & high-income nations. By putting the DOLS and VECM models, the author found that technological invention is the reason for environmental, economic, and for the social scopes of sustainability in nations only with high income, while it only results in the economic & environmental scopes in economies with middle income. No causal associations were noted in economies with low income. Fokkema et al. (2005) with an intensive view on social & environmental scopes of economic growth found that technology is a major aspect for sustainable development. Some researchers think that, considering a significant improvement in efficiency of environment and manufacturing of goods & services,

technological advancement should be deeply focused for guaranteed sustainable development.

Liu et al. (2024) conducted an asymmetric study of environmental quality towards sustainable development in case of UAE. The natural resources, renewable energy and export diversification are taken as variable of interest using the time period of 1990 to 2019. To explore the empirical findings, the nonlinear autoregressive distributed lag or NARDL and dynamic ordinary least squares or DOLS techniques have been used. Empirical evidence substantiates that export diversification, natural resources and economic growth adversely affect the UAE environment by accelerating carbon emissions. In this respect, positive and negative shocks of export diversification, natural resources, and economic growth increase and decrease carbon emissions, respectively, while positive and negative shocks of renewable energy decrease and increase carbon emissions. Udemba and Yalçıntaş (2021) examined that how FDI, natural resources as well as economic growth determine the environmental performance of Algeria, in other words, to study the sustainable development of the said country. To do so, the authors used the nonlinear autoregressive distributed lag NARDL model using the time frame of 1970 to 2018. The findings disclosed that economic growth and Algerian fossil fuel shocks have opposite consequences on carbon emissions (increasing and decreasing, respectively). However, FDI and natural resource shocks consistently result in decreased carbon emissions which are beneficial for the Algerian environment.

Ahmad et al., (2023) also explored that technological modernization is a crucial factor that encourages SD in China. The author by using the ARDL approach revealed that innovation is helpful for causing SD by decreasing pollution created by environment and improving the China’s growth. The author further discovered that financial expansion also donates significantly to sustainable development of China with the help of decline CO2 emissions. Bekhet & Latif (2018) studied the impact of technological innovation and institutional quality for attaining sustainable growth for the period 1995–2015. The authors exposed that alliance of governance & technological innovation adds positively to sustainable growth of Malaysia. Constantinescu & Frone (2014) argued that technological innovation (TI) has a vital role in advancing SD over its influence on three pillars of sustainability & also with conclusive impact on the attempts to encourage economic strength, social progress and sustainability of environment.

Wang et al. (2022) illustrated the influence of technological modernization on green TFP that led towards sustainable development. Taking the OECD economies for the period 1996–2017, the author used the threshold regression model. It was revealed that technological advancement has significantly and positively influenced the green total factor of productivity that promotes sustainable development. Moreover, Suki et al. (2022) suggested that the company putting on technology innovation supports sustainable growth. The author analyzed technological advancement with its effect on environmental and sustainable development. It was concluded that advanced technology along with usage of reprocessing energy helps performing business developments with no greenhouse gas emissions. So, technology innovation cares for sustainable growth. According to Silvestre, Țircă (2019) studied the role of technological innovation (TI) in diminishing greenhouse gas emissions & attaining sustainable expansion. The authors found that minimizing GHG emissions and a healthy environment offer strong human resources to sustainable development. Thus, technological advancement has an inverse relation with GHG emissions. Güney (2019) investigated how renewable energy and nonrenewable energy affect sustainable development. By using the panel data for 40

developed and 73 developing nations, the author employed the System-GMM and IV (2SLS) techniques and found that renewable energy has a positively and significantly influence the sustainable development both in advanced & developing economies. As there is a rise in renewable energy, the level of SD also goes up as explored by the author. It was also stated that comparatively, the impression of RE is more than that of non-renewable (NR) energy. Moreover, Zhang et al. (2021) investigated the association of ecotechnology innovation, economic effectiveness & the performance of economic wellbeing with sustainable development. By applying the 2 stage super slack based measure (SBM) taking 102 economies, it was found that there exists a positive linkage between advancement in technology and SD. Lyeonov et al. (2019) has studied the impact of green investment and renewable energy on greenhouse gas emissions and sustainable development (SD) in case of European Union from 2008–2016. Using the modes of FMOLS & DOLS, the outcomes showed that renewable energy use reduces the GHG emissions & hence increases environmental well-being which supports sustainable development. Moreover, Yu et al., (2023) analyzed how green growth in China is influenced by natural resource rent, economic stability and the legal framework. Using the ARDL model for the period 1980–2021, the authors found that economic risk is a hurdle towards green growth while natural resource rent and governance indicators drives the green growth in China.

Accordingly, Vasylieva et al. (2019) proposed a study on renewable energy's effects on GHG emissions and sustainable development using the hypothesis of Environmental Kuznets Curve in Ukraine & EU nations. The researchers used the FMOLS and DMOLS methods for the period from 2000 to 2016. It was revealed that renewable energy reduces GHGs, and environmental protection raises sustainable development. Udemba et al. (2024) worked out a study on the significance of financial development, technological innovation, renewable energy, and export diversification to attain long-term environmental quality in case of China. Applying the ARDL model and Granger causality test for the period of 1995Q1 to 2018Q4, it was revealed by the authors that technological innovation and renewable energy have valuable influence on the environmental quality while economic growth and export diversification have damaging impact on the environmental quality of China. Moslehpour et al. (2023) focused on the dynamic connectivity among Technological development, green finance, Energy efficiency and Sustainable development in Vietnam using the period 1991–2020. The authors with the Dynamic Autoregressive Distributed Lags & the Bayesian Autoregressive Distributed Lags model, explored that green finance, advanced technology, renewable energy (RE) output, renewable energy (RE) consumption, population rise & industrialization found to have a positive link with sustainable development in Vietnam.

Similarly, Sueyoshi et al., (2022) by using the nonparametric method called as Data Envelopment Analysis (DEA) explored that the segment of RE in total consumption of energy has positively and strongly influenced sustainability. The author studied how renewable energy affects sustainability at country level for the period 1990–2014 and found that renewable energy strongly and positively improves sustainability of the world. According to Cerqueira et al., (2021), renewable energy and recycling processes of waste are contributing meaningfully to improve the OECD countries' level of sustainable development. The study was an attempt to analyze the connections among renewable energy, recycling & economic development using the data for the period 2000–2016. It was revealed that both recycling & renewable energy are significant policy aspects for creating sustainable development in OECD countries. Liu et al. (2023)

investigated that whether renewable energy is a provides a path towards sustainable development in United states during the period 2013–2022. The test of Granger Causality was practiced examining the linkage between RE and sustainable development. Moreover, the bivariate VAR model revealed that RE influences sustainable development both in positive and negative manners. Similarly, Udemba and Alola (2022) used nonlinear ARDL approach and explored that the positive shocks to renewable energy result in reduction of carbon emission by 23 percent while in case of negative shocks, the carbon emission is raised by 16 percent. This highlights the significant role of renewable energy in attaining environmental sustainability which is a core pillar of sustainable development.

Moreover, Işık et al., (2024) employed the CS-ARDL model and discovered that economic and environmental factors are positively linked with natural resource rents indicating that sustainable environmental practices are important in expanding resource profitability. Moreover, the authors found that governance factors and technological innovation negatively influence the natural resource rents while SDGs also have a positive association with natural resource rents. In their second paper, Işık et al. (2024) investigated the impact of environmental, economic, social, and governance factors on SDG based energy efficiency in case of G-7 countries. Using the CS-ARDL model, the authors came with findings that economic factors relate negatively, while environmental factors relate positively with SDG based energy efficiency. To emphasize the importance of green policies in attaining SDG goals, Iqbal et al. (2024) used more advanced techniques that are panel-correlated standard errors (PCSE) and the Driscoll–Kraay estimations (DKSE) models and explored that green human capital, consumption of renewable energy, and quality institutions are found to considerably minimize CO<sub>2</sub> emissions. However, artificial intelligence, geopolitical risk, natural resource rent, and information communication technology lead to rise CO<sub>2</sub> emissions.

**Gaps in the existing literature.** Although there has been a significant amount of literature that has analyzed the intersection of technological innovation, renewable energy, and sustainable development but has focused mostly on countries like BRICS and USA. Moreover, most of the literature has only analyzed these variables separately, and minimal effort has been directed to the analysis of the individual and joint impacts of technological innovation and consumption of renewable energy on green growth which is a comprehensive stand-in for sustainable development on the wider European continent, encompassing categorization of income levels as well. Additionally, most studies ignore the heterogeneity of technology adoption and policy response in Europe, especially for countries facing diverse economic and environmental issues. This paper closes this research gap by exploring how the advent of renewable energy deployment affects European green growth and how technological change plays a role based on an innovative econometric strategy that accommodates cross-sectional dependence and heterogeneous dynamics in European countries.

The European countries are some of the world's major and fast-growing economies. Due to their considerable populations, manufacturing centers and resource consumption, they have major influences upon environmental issues everywhere. Because of this impact, knowing sustainable practices and means to drive sustainable growth and development around the world is significant in European nations. That is why the current study with a particular focus on European nations, will show that technological advancement and natural resource management, specifically renewable energy could be the driving forces of sustainable development.

**Theoretical framework.** The foundation of the theoretical instrument, through which natural resources (NR), with technological innovations (patent), Renewable energy consumption, Natural resource rent, Gross domestic product, and globalization affect green growth, is explained in this section. Natural resource rent, technological innovations, Gross Domestic Product (GDP), and globalization are all interconnected and have a significant impact on the concept of green growth. The following paragraphs explain the theoretical mechanism through which these factors affect green growth.

Technological innovation contributes to the change of economies' industrial and financial structures, while facilitating the encouragement of green growth. It is feasible to employ "patents" as a proxy for technological innovations, to identify significant points of origin for innovative technology intended to tackle environmental issues. A rising corpus of mostly policy-oriented research supports the intuition that patenting serves as a helpful, if not perfect, indication of the locations of new idea creation in the "green economy" (Tanner et al., 2019). Hence, following Popp (2010) and Khan et al., (2020) forecast that technology innovation will almost certainly have a favorable impact on GG (green growth) as  $\lambda_1 = \partial GG / \partial TI > 0$ . In the same manner, Manderson & Kneller (2012) and Saleem et al. (2022), also anticipated that technological innovation will have a favorable impact on green growth. SDG 17 aims to achieve all goals through extensive collaboration in all pertinent areas and the use of proactive alliances. The ultimate objective is to contribute to free access to scientific and technological information for sustainable development and growth. Technological innovations also contribute significantly to fostering green growth. Industries may implement eco-friendly technologies to help the environment by making investments in the green technology sector. The literature on how economic activity and globalization affect the rise of green growth offers contradictory findings. Globalization is an international phenomenon that has an impact on people's lives on a social and political level everywhere in the world (Bekun et al., 2020). In terms of negative impacts, however, globalization has a greater positive impact, especially when it comes to the elimination of poverty in rising nations and income disparity. While some studies (Ulucak & Khan, 2020; Christmann & Taylor, 2001; & Xu et al., (2018)) demonstrated that globalization can raise CO2 emissions, other researchers (Acheampong and Boateng, 2019; Akadiri et al., (2020)) demonstrate the contrary. Particularly, because of new business opportunities, the complicated shape of the relationship between transactions, GDP, & GG could vary as income, technology use, and macroeconomic structure change. To make this study more concrete this study also divided globalization into economic and political globalization (PG). In supplement to influencing economic advancement, economic globalization (EG) which is characterized by the cross-border trade of products & services, global capital mobility and technological transfer also channels its effects on green growth sustainability (Ahmad, Wu (2022)) and Political globalization, through international agreements and cooperation, can provide a framework for addressing global environmental challenges. If the technique's impact outweighs the size effect, globalization plus

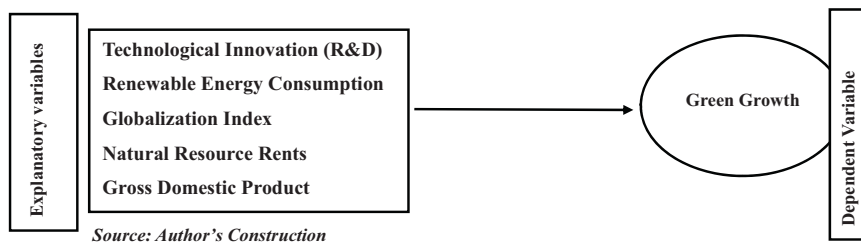
GDP may have a favorable effect on green growth as  $\lambda_2 = \partial GG / \partial GL > 0$ . But be careful, the opposite is also true.

The author explored difficulties with environmental degradation brought on by the natural resources' depletion. Natural resources are the foundation of economic growth (EG) and development. They are essential to produce goods & services, and their availability and quality are critical determinants of economic performance (SDG 17). Using natural resources, though, has negative consequences for the environment, such as pollution, deforestation, and climate change. Green growth goal is to settle economic growth with sustainable environment by promoting the efficient use of natural resources and the development of renewable resources as  $\lambda_3 = \partial GG / \partial NRR > 0$ . Countries aiming for the highest economic growth in this initial stage have drastically decreased environmental quality. Beyond this early phase, the economies' key objectives are to establish environmentally friendly policies to reduce CO2 emissions while achieving sustained financial stability and growth for the economy. Environmental economists and other decision-makers have used the study of EKC's hypotheses involving incomes, pollutants, & other crucial factors in the Gross Domestic Product square function as  $\lambda_4 = \partial GG / \partial GDP > 0$ . The EKC phenomena can be used to elaborate the relationship between environmental deterioration & growth. Following up on Kuznets' research, Grossman & Krueger (1991) divided the relationship between environmental quality & growth into three stages. The EKC channel is also debated for SAARC by (Yikun et al., 2021 and Li et al., 2021) for Globe (Dinda & Stern, 2004) for Asia (Apergis and Ozturk, 2015) for Kenya (Al-Mulali et al., (2015)) for OECD (Jebli et al., 2016) as  $\lambda_5 = \partial GG / \partial GDP^2 > 0$ . Renewable Energy consumption is sometimes termed as a machine of economic mechanism. Academia/ Scholars have practiced econometric methods and revealed that energy consumption (EC) is the key source of green environment sustainability and troubled and improved the climatic conditions of many countries (Caglar (2020)) as  $\lambda_6 = \partial GG / \partial REC > 0$ .

To enhance conceptual clarity, Fig. 1 below shows the visual representation of the theoretical framework. The direction of relationships will be explored empirically and may differ across income groups.

**Data and methods**

**Data and variables description.** This study provides a comprehensive analysis to investigate the determinants of green growth in European economies with changing income (Higher, Upper middle and lower income) levels. For this study we have taken data from 1990 to 2023 from different sources. Our dependent variable green growth is constructed as GDP per capita divided by CO<sub>2</sub> emissions, while independent variables are Natural resource rents, renewable energy consumption, GDP and technological innovations. All variables' data has been taken from WDI except KOFGI, which is obtained from KOF Swiss Economic Institute. Authenticity of data is guaranteed by drawing all variables from internationally accepted and reputable databases like the World



Source: Author's Construction

**Fig. 1 Conceptual framework of the study.** Source: Author's Construction.

**Table 1 Summary of data description and sources.**

Variables	Symbols	Sources	Units
Green Growth (GDP/CO <sub>2</sub> emissions)	GG	WDI	CO <sub>2</sub> (Metric tons per capita)
Technological Innovation (R&D)	TI	WDI	% of GDP
Renewable Energy Consumption	REC	WDI	% of total final energy consumption
Globalization Index	KOFGI	KOF Swiss Economic Institute	Index (0-100)
Natural Resource Rents	NRR	WDI	% of GDP
Gross Domestic Product	GDP	WDI	Constant US \$ 2015

Source: Author's own compilation.

Bank (WDI) and KOF Swiss Economic Institute which are used extensively in empirical economic studies due to their reliability, transparency, and consistency. Table 1 below further clarifies the description and sources of the data used in the current study.

*Justification of selected variables*

Technological innovation (R&D expenditure): R&D fosters technological advancements that lead to more efficient, cleaner production methods and energy solutions. These innovations can reduce CO<sub>2</sub> emissions while supporting economic growth. Increased R&D investments can drive technological changes that decouple GDP growth from environmental harm, supporting sustainable development (Romer, 1990; Porter, Linde (1995); Horbach et al., 2012).

Renewable energy consumption: Shifting from fossil fuels to renewable energy sources reduces CO<sub>2</sub> emissions, a critical factor in green growth. Higher renewable energy consumption helps decouple economic activity from carbon emissions. Increased renewable energy use lowers the carbon intensity of GDP, fostering sustainable economic growth (Stern, 2004; Apergis & Payne, 2010; Omri & Nguyen, 2014).

Globalization Index: Globalization influences both economic growth and environmental impact through trade, investment, and technology transfer. It can increase economic growth but may also raise environmental pressures unless accompanied by green technology diffusion. Globalization can thus stimulate green growth if paired with strong environmental policies (Grossman & Krueger, 1991; Shahbaz et al., 2018).

Natural resource rents (% of GDP): Resource rents from the extraction of natural resources can either promote or hinder green growth depending on governance and institutional quality. Mismanagement often leads to environmental degradation and higher emissions, while prudent use can finance green infrastructure and technology. This supports sustainable development if resource revenues are invested in diversifying the economy and reducing carbon intensity (Sachs & Warner, 2001; Van der Ploeg, Venables (2011); Bhattacharyya & Hodler, 2010).

Gross domestic product (GDP): GDP represents the level of economic output and is an essential part of the green growth measure. While economic growth typically raises emissions (scale effect), it also enhances capacity for adopting cleaner technologies (technique effect). Green growth occurs when GDP increases without a proportional increase in CO<sub>2</sub> emissions, i.e., when growth is decoupled from environmental degradation (Kuznets, 1955; Stern, 2004; Dinda, 2004).

**Model specification.** To investigate empirically how technological innovation affects green growth when GDP, REC, natural resource rent, and globalization is present. The empirical

equation may be modeled as:

$$GG_t = \lambda_1 TI_t + \lambda_2 GL_t + \lambda_3 NRR_t + \lambda_4 GDP_t + \lambda_5 REC_t + \mu_t \tag{1}$$

Where GG stands for green growth, which is the dependent variable, representing the level of sustainable economic growth & development achieved whereas minimizing negative environmental effects. GL stands for economic and political globalization and TI represents for technological innovation and its proxy Research and Development expenditure, REC for renewable energy consumption, NRR represents natural resource rent, and GDP shows gross domestic product. The empirical equation may be estimated using regression analysis, with data collected from European countries over a specified period. The results of the analysis will help to determine the extent to which technological innovation contributes to green growth, and whether this contribution is significant even when controlling other factors such as GDP, natural resource endowment, and economic globalization.

**Econometric methodology.**

$$\Delta GG_{it} = \alpha_i + \beta_1 \Delta GG_{it-1} + \beta_2 \Delta TI_{it-1} + \beta_3 \Delta REC_{it-1} + \beta_4 \Delta GDP_{it-1} + \beta_5 \Delta NRR_{it-1} + \beta_6 \Delta GI_{it-1} + \beta_7 \Delta X_{it-1} + \beta_8 \Delta C_{it-1} + \beta_9 \Delta X_{it-2} + \beta_{10} \Delta C_{it-2} + \beta_{11} \Delta X_{it-3} + \beta_{12} \Delta C_{it-3} + \beta_{13} \Delta X_{it-4} + \beta_{14} \Delta C_{it-4} + \beta_{15} \Delta X_{it-5} + \beta_{16} \Delta C_{it-5} + \beta_{17} \Delta X_{it-6} + \beta_{18} \Delta C_{it-6} + \beta_{19} \Delta X_{it-7} + \beta_{20} \Delta C_{it-7} + \beta_{21} \Delta X_{it-8} + \beta_{22} \Delta C_{it-8} + \beta_{23} \Delta X_{it-9} + \beta_{24} \Delta C_{it-9} + \beta_{25} \Delta X_{it-10} + \beta_{26} \Delta C_{it-10} + \beta_{27} \Delta X_{it-11} + \beta_{28} \Delta C_{it-11} + \beta_{29} \Delta X_{it-12} + \beta_{30} \Delta C_{it-12} + \beta_{31} \Delta X_{it-13} + \beta_{32} \Delta C_{it-13} + \beta_{33} \Delta X_{it-14} + \beta_{34} \Delta C_{it-14} + \beta_{35} \Delta X_{it-15} + \beta_{36} \Delta C_{it-15} + \beta_{37} \Delta X_{it-16} + \beta_{38} \Delta C_{it-16} + \beta_{39} \Delta X_{it-17} + \beta_{40} \Delta C_{it-17} + \beta_{41} \Delta X_{it-18} + \beta_{42} \Delta C_{it-18} + \beta_{43} \Delta X_{it-19} + \beta_{44} \Delta C_{it-19} + \beta_{45} \Delta X_{it-20} + \beta_{46} \Delta C_{it-20} + \beta_{47} \Delta X_{it-21} + \beta_{48} \Delta C_{it-21} + \beta_{49} \Delta X_{it-22} + \beta_{50} \Delta C_{it-22} + \beta_{51} \Delta X_{it-23} + \beta_{52} \Delta C_{it-23} + \beta_{53} \Delta X_{it-24} + \beta_{54} \Delta C_{it-24} + \beta_{55} \Delta X_{it-25} + \beta_{56} \Delta C_{it-25} + \beta_{57} \Delta X_{it-26} + \beta_{58} \Delta C_{it-26} + \beta_{59} \Delta X_{it-27} + \beta_{60} \Delta C_{it-27} + \beta_{61} \Delta X_{it-28} + \beta_{62} \Delta C_{it-28} + \beta_{63} \Delta X_{it-29} + \beta_{64} \Delta C_{it-29} + \beta_{65} \Delta X_{it-30} + \beta_{66} \Delta C_{it-30} + \beta_{67} \Delta X_{it-31} + \beta_{68} \Delta C_{it-31} + \beta_{69} \Delta X_{it-32} + \beta_{70} \Delta C_{it-32} + \beta_{71} \Delta X_{it-33} + \beta_{72} \Delta C_{it-33} + \beta_{73} \Delta X_{it-34} + \beta_{74} \Delta C_{it-34} + \beta_{75} \Delta X_{it-35} + \beta_{76} \Delta C_{it-35} + \beta_{77} \Delta X_{it-36} + \beta_{78} \Delta C_{it-36} + \beta_{79} \Delta X_{it-37} + \beta_{80} \Delta C_{it-37} + \beta_{81} \Delta X_{it-38} + \beta_{82} \Delta C_{it-38} + \beta_{83} \Delta X_{it-39} + \beta_{84} \Delta C_{it-39} + \beta_{85} \Delta X_{it-40} + \beta_{86} \Delta C_{it-40} + \beta_{87} \Delta X_{it-41} + \beta_{88} \Delta C_{it-41} + \beta_{89} \Delta X_{it-42} + \beta_{90} \Delta C_{it-42} + \beta_{91} \Delta X_{it-43} + \beta_{92} \Delta C_{it-43} + \beta_{93} \Delta X_{it-44} + \beta_{94} \Delta C_{it-44} + \beta_{95} \Delta X_{it-45} + \beta_{96} \Delta C_{it-45} + \beta_{97} \Delta X_{it-46} + \beta_{98} \Delta C_{it-46} + \beta_{99} \Delta X_{it-47} + \beta_{100} \Delta C_{it-47} + \beta_{101} \Delta X_{it-48} + \beta_{102} \Delta C_{it-48} + \beta_{103} \Delta X_{it-49} + \beta_{104} \Delta C_{it-49} + \beta_{105} \Delta X_{it-50} + \beta_{106} \Delta C_{it-50} + \beta_{107} \Delta X_{it-51} + \beta_{108} \Delta C_{it-51} + \beta_{109} \Delta X_{it-52} + \beta_{110} \Delta C_{it-52} + \beta_{111} \Delta X_{it-53} + \beta_{112} \Delta C_{it-53} + \beta_{113} \Delta X_{it-54} + \beta_{114} \Delta C_{it-54} + \beta_{115} \Delta X_{it-55} + \beta_{116} \Delta C_{it-55} + \beta_{117} \Delta X_{it-56} + \beta_{118} \Delta C_{it-56} + \beta_{119} \Delta X_{it-57} + \beta_{120} \Delta C_{it-57} + \beta_{121} \Delta X_{it-58} + \beta_{122} \Delta C_{it-58} + \beta_{123} \Delta X_{it-59} + \beta_{124} \Delta C_{it-59} + \beta_{125} \Delta X_{it-60} + \beta_{126} \Delta C_{it-60} + \beta_{127} \Delta X_{it-61} + \beta_{128} \Delta C_{it-61} + \beta_{129} \Delta X_{it-62} + \beta_{130} \Delta C_{it-62} + \beta_{131} \Delta X_{it-63} + \beta_{132} \Delta C_{it-63} + \beta_{133} \Delta X_{it-64} + \beta_{134} \Delta C_{it-64} + \beta_{135} \Delta X_{it-65} + \beta_{136} \Delta C_{it-65} + \beta_{137} \Delta X_{it-66} + \beta_{138} \Delta C_{it-66} + \beta_{139} \Delta X_{it-67} + \beta_{140} \Delta C_{it-67} + \beta_{141} \Delta X_{it-68} + \beta_{142} \Delta C_{it-68} + \beta_{143} \Delta X_{it-69} + \beta_{144} \Delta C_{it-69} + \beta_{145} \Delta X_{it-70} + \beta_{146} \Delta C_{it-70} + \beta_{147} \Delta X_{it-71} + \beta_{148} \Delta C_{it-71} + \beta_{149} \Delta X_{it-72} + \beta_{150} \Delta C_{it-72} + \beta_{151} \Delta X_{it-73} + \beta_{152} \Delta C_{it-73} + \beta_{153} \Delta X_{it-74} + \beta_{154} \Delta C_{it-74} + \beta_{155} \Delta X_{it-75} + \beta_{156} \Delta C_{it-75} + \beta_{157} \Delta X_{it-76} + \beta_{158} \Delta C_{it-76} + \beta_{159} \Delta X_{it-77} + \beta_{160} \Delta C_{it-77} + \beta_{161} \Delta X_{it-78} + \beta_{162} \Delta C_{it-78} + \beta_{163} \Delta X_{it-79} + \beta_{164} \Delta C_{it-79} + \beta_{165} \Delta X_{it-80} + \beta_{166} \Delta C_{it-80} + \beta_{167} \Delta X_{it-81} + \beta_{168} \Delta C_{it-81} + \beta_{169} \Delta X_{it-82} + \beta_{170} \Delta C_{it-82} + \beta_{171} \Delta X_{it-83} + \beta_{172} \Delta C_{it-83} + \beta_{173} \Delta X_{it-84} + \beta_{174} \Delta C_{it-84} + \beta_{175} \Delta X_{it-85} + \beta_{176} \Delta C_{it-85} + \beta_{177} \Delta X_{it-86} + \beta_{178} \Delta C_{it-86} + \beta_{179} \Delta X_{it-87} + \beta_{180} \Delta C_{it-87} + \beta_{181} \Delta X_{it-88} + \beta_{182} \Delta C_{it-88} + \beta_{183} \Delta X_{it-89} + \beta_{184} \Delta C_{it-89} + \beta_{185} \Delta X_{it-90} + \beta_{186} \Delta C_{it-90} + \beta_{187} \Delta X_{it-91} + \beta_{188} \Delta C_{it-91} + \beta_{189} \Delta X_{it-92} + \beta_{190} \Delta C_{it-92} + \beta_{191} \Delta X_{it-93} + \beta_{192} \Delta C_{it-93} + \beta_{193} \Delta X_{it-94} + \beta_{194} \Delta C_{it-94} + \beta_{195} \Delta X_{it-95} + \beta_{196} \Delta C_{it-95} + \beta_{197} \Delta X_{it-96} + \beta_{198} \Delta C_{it-96} + \beta_{199} \Delta X_{it-97} + \beta_{200} \Delta C_{it-97} + \beta_{201} \Delta X_{it-98} + \beta_{202} \Delta C_{it-98} + \beta_{203} \Delta X_{it-99} + \beta_{204} \Delta C_{it-99} + \beta_{205} \Delta X_{it-100} + \beta_{206} \Delta C_{it-100} + \beta_{207} \Delta X_{it-101} + \beta_{208} \Delta C_{it-101} + \beta_{209} \Delta X_{it-102} + \beta_{210} \Delta C_{it-102} + \beta_{211} \Delta X_{it-103} + \beta_{212} \Delta C_{it-103} + \beta_{213} \Delta X_{it-104} + \beta_{214} \Delta C_{it-104} + \beta_{215} \Delta X_{it-105} + \beta_{216} \Delta C_{it-105} + \beta_{217} \Delta X_{it-106} + \beta_{218} \Delta C_{it-106} + \beta_{219} \Delta X_{it-107} + \beta_{220} \Delta C_{it-107} + \beta_{221} \Delta X_{it-108} + \beta_{222} \Delta C_{it-108} + \beta_{223} \Delta X_{it-109} + \beta_{224} \Delta C_{it-109} + \beta_{225} \Delta X_{it-110} + \beta_{226} \Delta C_{it-110} + \beta_{227} \Delta X_{it-111} + \beta_{228} \Delta C_{it-111} + \beta_{229} \Delta X_{it-112} + \beta_{230} \Delta C_{it-112} + \beta_{231} \Delta X_{it-113} + \beta_{232} \Delta C_{it-113} + \beta_{233} \Delta X_{it-114} + \beta_{234} \Delta C_{it-114} + \beta_{235} \Delta X_{it-115} + \beta_{236} \Delta C_{it-115} + \beta_{237} \Delta X_{it-116} + \beta_{238} \Delta C_{it-116} + \beta_{239} \Delta X_{it-117} + \beta_{240} \Delta C_{it-117} + \beta_{241} \Delta X_{it-118} + \beta_{242} \Delta C_{it-118} + \beta_{243} \Delta X_{it-119} + \beta_{244} \Delta C_{it-119} + \beta_{245} \Delta X_{it-120} + \beta_{246} \Delta C_{it-120} + \beta_{247} \Delta X_{it-121} + \beta_{248} \Delta C_{it-121} + \beta_{249} \Delta X_{it-122} + \beta_{250} \Delta C_{it-122} + \beta_{251} \Delta X_{it-123} + \beta_{252} \Delta C_{it-123} + \beta_{253} \Delta X_{it-124} + \beta_{254} \Delta C_{it-124} + \beta_{255} \Delta X_{it-125} + \beta_{256} \Delta C_{it-125} + \beta_{257} \Delta X_{it-126} + \beta_{258} \Delta C_{it-126} + \beta_{259} \Delta X_{it-127} + \beta_{260} \Delta C_{it-127} + \beta_{261} \Delta X_{it-128} + \beta_{262} \Delta C_{it-128} + \beta_{263} \Delta X_{it-129} + \beta_{264} \Delta C_{it-129} + \beta_{265} \Delta X_{it-130} + \beta_{266} \Delta C_{it-130} + \beta_{267} \Delta X_{it-131} + \beta_{268} \Delta C_{it-131} + \beta_{269} \Delta X_{it-132} + \beta_{270} \Delta C_{it-132} + \beta_{271} \Delta X_{it-133} + \beta_{272} \Delta C_{it-133} + \beta_{273} \Delta X_{it-134} + \beta_{274} \Delta C_{it-134} + \beta_{275} \Delta X_{it-135} + \beta_{276} \Delta C_{it-135} + \beta_{277} \Delta X_{it-136} + \beta_{278} \Delta C_{it-136} + \beta_{279} \Delta X_{it-137} + \beta_{280} \Delta C_{it-137} + \beta_{281} \Delta X_{it-138} + \beta_{282} \Delta C_{it-138} + \beta_{283} \Delta X_{it-139} + \beta_{284} \Delta C_{it-139} + \beta_{285} \Delta X_{it-140} + \beta_{286} \Delta C_{it-140} + \beta_{287} \Delta X_{it-141} + \beta_{288} \Delta C_{it-141} + \beta_{289} \Delta X_{it-142} + \beta_{290} \Delta C_{it-142} + \beta_{291} \Delta X_{it-143} + \beta_{292} \Delta C_{it-143} + \beta_{293} \Delta X_{it-144} + \beta_{294} \Delta C_{it-144} + \beta_{295} \Delta X_{it-145} + \beta_{296} \Delta C_{it-145} + \beta_{297} \Delta X_{it-146} + \beta_{298} \Delta C_{it-146} + \beta_{299} \Delta X_{it-147} + \beta_{300} \Delta C_{it-147} + \beta_{301} \Delta X_{it-148} + \beta_{302} \Delta C_{it-148} + \beta_{303} \Delta X_{it-149} + \beta_{304} \Delta C_{it-149} + \beta_{305} \Delta X_{it-150} + \beta_{306} \Delta C_{it-150} + \beta_{307} \Delta X_{it-151} + \beta_{308} \Delta C_{it-151} + \beta_{309} \Delta X_{it-152} + \beta_{310} \Delta C_{it-152} + \beta_{311} \Delta X_{it-153} + \beta_{312} \Delta C_{it-153} + \beta_{313} \Delta X_{it-154} + \beta_{314} \Delta C_{it-154} + \beta_{315} \Delta X_{it-155} + \beta_{316} \Delta C_{it-155} + \beta_{317} \Delta X_{it-156} + \beta_{318} \Delta C_{it-156} + \beta_{319} \Delta X_{it-157} + \beta_{320} \Delta C_{it-157} + \beta_{321} \Delta X_{it-158} + \beta_{322} \Delta C_{it-158} + \beta_{323} \Delta X_{it-159} + \beta_{324} \Delta C_{it-159} + \beta_{325} \Delta X_{it-160} + \beta_{326} \Delta C_{it-160} + \beta_{327} \Delta X_{it-161} + \beta_{328} \Delta C_{it-161} + \beta_{329} \Delta X_{it-162} + \beta_{330} \Delta C_{it-162} + \beta_{331} \Delta X_{it-163} + \beta_{332} \Delta C_{it-163} + \beta_{333} \Delta X_{it-164} + \beta_{334} \Delta C_{it-164} + \beta_{335} \Delta X_{it-165} + \beta_{336} \Delta C_{it-165} + \beta_{337} \Delta X_{it-166} + \beta_{338} \Delta C_{it-166} + \beta_{339} \Delta X_{it-167} + \beta_{340} \Delta C_{it-167} + \beta_{341} \Delta X_{it-168} + \beta_{342} \Delta C_{it-168} + \beta_{343} \Delta X_{it-169} + \beta_{344} \Delta C_{it-169} + \beta_{345} \Delta X_{it-170} + \beta_{346} \Delta C_{it-170} + \beta_{347} \Delta X_{it-171} + \beta_{348} \Delta C_{it-171} + \beta_{349} \Delta X_{it-172} + \beta_{350} \Delta C_{it-172} + \beta_{351} \Delta X_{it-173} + \beta_{352} \Delta C_{it-173} + \beta_{353} \Delta X_{it-174} + \beta_{354} \Delta C_{it-174} + \beta_{355} \Delta X_{it-175} + \beta_{356} \Delta C_{it-175} + \beta_{357} \Delta X_{it-176} + \beta_{358} \Delta C_{it-176} + \beta_{359} \Delta X_{it-177} + \beta_{360} \Delta C_{it-177} + \beta_{361} \Delta X_{it-178} + \beta_{362} \Delta C_{it-178} + \beta_{363} \Delta X_{it-179} + \beta_{364} \Delta C_{it-179} + \beta_{365} \Delta X_{it-180} + \beta_{366} \Delta C_{it-180} + \beta_{367} \Delta X_{it-181} + \beta_{368} \Delta C_{it-181} + \beta_{369} \Delta X_{it-182} + \beta_{370} \Delta C_{it-182} + \beta_{371} \Delta X_{it-183} + \beta_{372} \Delta C_{it-183} + \beta_{373} \Delta X_{it-184} + \beta_{374} \Delta C_{it-184} + \beta_{375} \Delta X_{it-185} + \beta_{376} \Delta C_{it-185} + \beta_{377} \Delta X_{it-186} + \beta_{378} \Delta C_{it-186} + \beta_{379} \Delta X_{it-187} + \beta_{380} \Delta C_{it-187} + \beta_{381} \Delta X_{it-188} + \beta_{382} \Delta C_{it-188} + \beta_{383} \Delta X_{it-189} + \beta_{384} \Delta C_{it-189} + \beta_{385} \Delta X_{it-190} + \beta_{386} \Delta C_{it-190} + \beta_{387} \Delta X_{it-191} + \beta_{388} \Delta C_{it-191} + \beta_{389} \Delta X_{it-192} + \beta_{390} \Delta C_{it-192} + \beta_{391} \Delta X_{it-193} + \beta_{392} \Delta C_{it-193} + \beta_{393} \Delta X_{it-194} + \beta_{394} \Delta C_{it-194} + \beta_{395} \Delta X_{it-195} + \beta_{396} \Delta C_{it-195} + \beta_{397} \Delta X_{it-196} + \beta_{398} \Delta C_{it-196} + \beta_{399} \Delta X_{it-197} + \beta_{400} \Delta C_{it-197} + \beta_{401} \Delta X_{it-198} + \beta_{402} \Delta C_{it-198} + \beta_{403} \Delta X_{it-199} + \beta_{404} \Delta C_{it-199} + \beta_{405} \Delta X_{it-200} + \beta_{406} \Delta C_{it-200} + \beta_{407} \Delta X_{it-201} + \beta_{408} \Delta C_{it-201} + \beta_{409} \Delta X_{it-202} + \beta_{410} \Delta C_{it-202} + \beta_{411} \Delta X_{it-203} + \beta_{412} \Delta C_{it-203} + \beta_{413} \Delta X_{it-204} + \beta_{414} \Delta C_{it-204} + \beta_{415} \Delta X_{it-205} + \beta_{416} \Delta C_{it-205} + \beta_{417} \Delta X_{it-206} + \beta_{418} \Delta C_{it-206} + \beta_{419} \Delta X_{it-207} + \beta_{420} \Delta C_{it-207} + \beta_{421} \Delta X_{it-208} + \beta_{422} \Delta C_{it-208} + \beta_{423} \Delta X_{it-209} + \beta_{424} \Delta C_{it-209} + \beta_{425} \Delta X_{it-210} + \beta_{426} \Delta C_{it-210} + \beta_{427} \Delta X_{it-211} + \beta_{428} \Delta C_{it-211} + \beta_{429} \Delta X_{it-212} + \beta_{430} \Delta C_{it-212} + \beta_{431} \Delta X_{it-213} + \beta_{432} \Delta C_{it-213} + \beta_{433} \Delta X_{it-214} + \beta_{434} \Delta C_{it-214} + \beta_{435} \Delta X_{it-215} + \beta_{436} \Delta C_{it-215} + \beta_{437} \Delta X_{it-216} + \beta_{438} \Delta C_{it-216} + \beta_{439} \Delta X_{it-217} + \beta_{440} \Delta C_{it-217} + \beta_{441} \Delta X_{it-218} + \beta_{442} \Delta C_{it-218} + \beta_{443} \Delta X_{it-219} + \beta_{444} \Delta C_{it-219} + \beta_{445} \Delta X_{it-220} + \beta_{446} \Delta C_{it-220} + \beta_{447} \Delta X_{it-221} + \beta_{448} \Delta C_{it-221} + \beta_{449} \Delta X_{it-222} + \beta_{450} \Delta C_{it-222} + \beta_{451} \Delta X_{it-223} + \beta_{452} \Delta C_{it-223} + \beta_{453} \Delta X_{it-224} + \beta_{454} \Delta C_{it-224} + \beta_{455} \Delta X_{it-225} + \beta_{456} \Delta C_{it-225} + \beta_{457} \Delta X_{it-226} + \beta_{458} \Delta C_{it-226} + \beta_{459} \Delta X_{it-227} + \beta_{460} \Delta C_{it-227} + \beta_{461} \Delta X_{it-228} + \beta_{462} \Delta C_{it-228} + \beta_{463} \Delta X_{it-229} + \beta_{464} \Delta C_{it-229} + \beta_{465} \Delta X_{it-230} + \beta_{466} \Delta C_{it-230} + \beta_{467} \Delta X_{it-231} + \beta_{468} \Delta C_{it-231} + \beta_{469} \Delta X_{it-232} + \beta_{470} \Delta C_{it-232} + \beta_{471} \Delta X_{it-233} + \beta_{472} \Delta C_{it-233} + \beta_{473} \Delta X_{it-234} + \beta_{474} \Delta C_{it-234} + \beta_{475} \Delta X_{it-235} + \beta_{476} \Delta C_{it-235} + \beta_{477} \Delta X_{it-236} + \beta_{478} \Delta C_{it-236} + \beta_{479} \Delta X_{it-237} + \beta_{480} \Delta C_{it-237} + \beta_{481} \Delta X_{it-238} + \beta_{482} \Delta C_{it-238} + \beta_{483} \Delta X_{it-239} + \beta_{484} \Delta C_{it-239} + \beta_{485} \Delta X_{it-240} + \beta_{486} \Delta C_{it-240} + \beta_{487} \Delta X_{it-241} + \beta_{488} \Delta C_{it-241} + \beta_{489} \Delta X_{it-242} + \beta_{490} \Delta C_{it-242} + \beta_{491} \Delta X_{it-243} + \beta_{492} \Delta C_{it-243} + \beta_{493} \Delta X_{it-244} + \beta_{494} \Delta C_{it-244} + \beta_{495} \Delta X_{it-245} + \beta_{496} \Delta C_{it-245} + \beta_{497} \Delta X_{it-246} + \beta_{498} \Delta C_{it-246} + \beta_{499} \Delta X_{it-247} + \beta_{500} \Delta C_{it-247} + \beta_{501} \Delta X_{it-248} + \beta_{502} \Delta C_{it-248} + \beta_{503} \Delta X_{it-249} + \beta_{504} \Delta C_{it-249} + \beta_{505} \Delta X_{it-250} + \beta_{506} \Delta C_{it-250} + \beta_{507} \Delta 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sectional averages which adjusts for omitted variable bias and potential endogeneity. This makes it effective for investigating the changing impact of technological innovation and renewable energy on sustainable development (green growth) across different income groups. As discussed above, the current study extracts those factors which are causing damaging effects on green growth. Before applying the suitable estimation techniques to the selected data series, we need to verify if the variables under study are stationary. For this, we exercise the co-integration test to check the order of integration by considering unit root test. In this context, the time under examination considers the changes and transformations that occur, as these primarily lead to structural breaks. The other typical unit root test for example ADF; DF-GLS and KPSS generally don't give structural breaks, resulting in inconsistent & biased outcomes (Umar et al., (2020)). The benefit of this technique is that, unlike the more traditional unit root test approaches that are frequently considered, the structural breaks that occur in this part do not exert any influence while the data is being processed. Following stationary testing, Bayer and Hanck's (2013) teachings are used for looking into the co-integrative link between the variables that were currently considered. It is notable to observe that this upgraded co-integration approach tends to produce results that are far more precise in nature. This is achievable because the results of numerous tests are merged to provide a single result. Engle and Granger (1987), Johansen (1991); Boswijk (1994) & Banerjee et al. (1998) are few of the tests that fall under this category. As a result, the equations of Bayer-Hanck Fisher are shown below:

$$EG - JOH = -2[\ln(PEG) + \ln(PJOH)],$$

$$EG - JOH - BO - BDM = -2[\ln(PEG) + \ln(PJOH) + \ln(PBO) + \ln(PBDM)] \quad (3)$$

In the above Eq. 2, the symbols PEG; PJOH; PBO & PBDM stand for the probability values corresponding to each of the co-integration tests that were described earlier.

Furthermore, the Fisher statistics setup establishes the nature of the long-term relationship between the central variables. Following the study uses the Maki co-integration test, which was created by (Maki, 2012) and considers a variety of structural breaks. Maki co-integration test, in contrast to other co-integration tests that involves structural breaks (Hatemi-j, 2008), has a very well-defined magnitude and qualities of influence, according to writers like (Cheng et al., 2021). This is because ignoring the presence of additional structural breaks can lead to biased and inaccurate estimates of the model parameters and their associated statistical significance. Multiple structural breaks can have significant impacts on the underlying relationships between variables. Therefore, it is important to conduct a thorough analysis that considers the possibility of multiple structural breaks and accounts for their potential effects on the data.

The present test additionally considers the DOLS & CCR, which were proposed by (Saikkonen (1992)), (Watson (1993)), and (Park, 1992), as well as the Fully Modified OLS that was popularized by (Phillips and Hansen, 1990). The test is considered in order to find any potential long-term impacts of Technological-innovation, Globalization, GDP, Natural resource, and GDP square on GG in BRICS. To get around this correlation problem, Phillips & Hansen (1990) also devised the Semi-Parametric method called FMOLS. The FMOLS test tends to be asymptotically unbiased and proactive in nature, they highlighted. The FMOLS and CCR have been replaced by another approach that was originate by (Park, 1992).

The primary distinction among the Fully Modified OLS and the CCR estimators in this context stems from the fact that, whereas the FMOLS concentrates on the reconstruction of data as well as the appropriate parameters, the CCR estimator frequently concentrates solely on changing the data (Hassan, Abubakar (2025)). In addition, the DOLS estimate makes use of lags and leads to the loss against simultaneity biases and small number of samples error. In particular, when analyzing, DOLS & FMOLS estimation methods must contend with the challenge of resolving the endogeneity as well as serial correlation issues by controlling the disturbance factors (Su et al., (2020)). The current study also makes use of the Causality test, which was first created by (Geweke (1982)) & (Hosoya (1991)) & later refined by (Breitung & Candelon, 2006) for the purpose to take into consideration the causal link.

**Results and discussion**

Table 2 gives the summary statistics of all the variables. The total number of observations has 1496, which is reliable sample size for statistical analysis. The most of the variables Mean and Median are fairly close except lnIT which is negatively skewed. The high mean was observed for a variable lnGG (21.5), indicating that these countries are relatively high growth. The value 2.37 of lnNRR variable for standard deviation shows mostly dispersed while the standard deviation of the lnKOFGI is 0.23 showing low variation in globalization across sample. Here, the skewness values indicate that most of the variables are negatively skewed which means that variables have longer tail to the left side. A very high kurtosis value reported for the variable lnIT (21.83), while other variables lnNRR, lnREC, and lnKOFGI are moderately leptokurtic, meaning that some outliers and sharp peaks in distribution.

Table 3 demonstrates the findings of the Cross-Sectional Dependence (CSD) for all variables used in the model. This test was proposed by Pesaran (2004) with an argument that cross sectional dependence exists in panel data. We reported that the CD test statistics are highly significant for all variables, indicating that there is a strong cross-sectional dependency across countries or panel. The average and absolute correlation values for variable lnKOFGI (0.95), lnGG (0.86) and lnGDP (0.84) showing a very

Variable	LnGG	lnTI	lnGDP	lnNRR	lnREC	lnKOFGI
Obs	1,496	1,496	1,496	1,496	1,496	1,496
Mean	21.518	-0.181	9.567	-1.12	2.317	4.243
Median	21.6537	0.03998	9.73546	-0.9062	2.5177	4.32301
Std. Dev.	0.933	1.338	1.129	2.378	1.197	0.23
Variance	0.87047	1.79155	1.27498	5.65555	1.43383	0.05272
Min	18.78	-12.045	6.558	-12.114	-2.303	3.216
Max	23.87	6.609	11.63	5.872	4.418	4.498
Skewness	-0.345	-2.632	-0.406	-0.784	-0.941	-1.654
Kurtosis	2.73	21.831	2.299	4.685	4.324	5.895

high positive correlation among different cross sections. Even variable  $\ln\text{NRR}$  and  $\ln\text{TI}$  showing lower correlation coefficients but still exhibits significant dependence. After recognizing the cross-sectional dependence, next is to explore the stationarity property of the variables.

We have employed the first generation (LLC and IPS) proposed by Levin et al. (2002) and Im, Pesaran and Shin (2003) respectively. Moreover, the second generation (CADF and CIPS) panel unit root tests proposed by Hansen (1995) and Pesaran (2007) respectively, is also applied to determine stationarity of the variables. In Table 4, across all four panel unit root tests showing that most of the variables are non-stationary at level  $I(0)$  but become stationary after first differencing  $I(1)$ . If we look at the table variable  $\ln\text{GG}$  at level it is not stationary at all tests so we fail to reject the null hypothesis of a unit root test. At first difference, all tests strongly reject the null hypothesis because green growth variable is stationary at first difference. A variable  $\ln\text{IT}$  also non-stationary at level in all panel while stationary at first difference. Gross Domestic Product variable is stationary at level  $I(0)$  in panel LLC, CADF and CIPS while at first difference stationary at all panels. A  $\ln\text{REC}$  stationary at mixed order reported in the table. Natural resource rent variable is non-stationary at level while stationary at first difference. In sum, both the first- and second-generation panel unit root tests resulted in the mixed order of variables for stationarity which means that we will have to perform the ARDL model approach for estimation. Before going towards the process of estimation, it is quite relevant to explore the existence of long run relationship among the variables. For this objective, we have applied for the Westerlund (2005) and Pedroni (1999, 2004) cointegration tests. In Table 5 we have applied both tests for checking either long run relationship exists among the variables in the panels. We observed that the values of Pedroni cointegration test are statistically significant, suggesting that there is long run relationship exists among the variables. While Westerlund test values are insignificant showing no long run relationship exists. But literature suggests if you still want to explore long run relationship you can apply on the basis of Pedroni test which suggests the long run relationship exists among variables.

The estimated outcomes from the Cross Sectional Autoregressive Distributed lag (CS-ARDL) model are shown in Table 6. The variable technological innovation ( $\ln\text{TI}$ ) indicates a positive and significant impact on green growth in both the short and

long run in the European Economies. Its coefficient in the short run is 0.067 while in the long run is 0.03 and statistically significant. Our finding is consistent with Bekhet & Latif (2018), Sun et al. (2023) and Kwilinski et al. (2025), who also discovered the positive impact of technology innovation on sustainable development. GDP is also a significant positive impact in both the periods, indicating that economic growth is a major indicator of green growth in this region. Its long run coefficient value is 0.32 showing that if one percent increase in GDP leads to green growth by 32 percent in the European region. Renewable energy consumption ( $\ln\text{REC}$ ) has also reported positive and significant impact on green growth. It shows that clean energy adaptation is important for achieving environmental goals. This finding is supported by Güney (2019), Moslehpour et al. (2023) and Khan et al. (2025). The remaining variables  $\ln\text{NRR}$  and  $\ln\text{KOFGI}$  both are insignificant, demonstrating that in the context of Europe, resource dependency and globalization don't play a direct role in driving green growth in the long run, may be due to strict environmental laws and diversified economies in Europe. It is a usual recognition that economies are heterogeneous in their gross national income and can have different behavior of various economic variables. The results presented in Table 6 can be misleading if these were driven by only a certain group of economies. To avoid this uncertainty about the findings we have decomposed our data set into three groups based on income. The subsample classification contains the higher income nations, upper middle-income countries & lower middle-income countries. This study estimates the same model (CS-ARDL) separately for three subsamples.

The empirically estimated results of the CS-ARDL presented in Table 7 based on different income groups (Higher income, Upper middle income and Lower middle income) countries. It will explain that how different macroeconomic variables influence green growth in European economies. If we look at the long run estimates,  $\ln\text{TI}$  coefficient value is 0.076 which is positive and significant in higher income economies, indicating that R&D and clean innovation promoting green growth in these regions while insignificant impact reported in upper and lower middle-income countries due to weak innovation ecosystems in these economies. Economic growth variable ( $\ln\text{GDP}$ ) shows a positive and significant relationship with green growth in all three income groups. This can be explained that if one percent increase in GDP leads to increase green growth by 0.24, 0.26 and 0.32 percent for higher, upper-middle and lower middle-income countries in long run. The findings are consistent with Iqbal et al. (2024). Renewable energy consumption ( $\ln\text{REC}$ ) also reported positive and significant impact in all three income groups. The renewable energy consumption has a universal importance but if we look at the policy and investment in renewable are paying off particularly in higher income economies. Natural resource rent ( $\ln\text{NRR}$ ) coefficient value is 0.015 which is significant impact on green growth only in lower-middle income countries, while in higher income and upper middle-income economies the relationship is insignificant,

**Table 3 Findings of the Cross-Sectional Dependence test.**

Variable	CD-test	p-value	Corr	Abs(Corr)
$\ln\text{GG}$	154.59	0.000	0.862	0.862
$\ln\text{TI}$	42.29	0.000	0.236	0.620
$\ln\text{GDP}$	152.08	0.000	0.848	0.848
$\ln\text{NRR}$	38.73	0.000	0.216	0.370
$\ln\text{REC}$	79.30	0.000	0.442	0.609
$\ln\text{KOFGI}$	171.36	0.000	0.955	0.955

**Table 4 Panel Unit Root test.**

Variable	LLC		IPS		CIPS		CADF	
	I(0)	I(1)	I(0)	I(1)	I(0)	I(1)	I(0)	I(1)
$\ln\text{GG}$	4.8167	-14.1075***	0.2167	-5.6811***	-1.891	-3.792***	-1.825	-5.225***
$\ln\text{TI}$	0.1890	-10.6808***	-1.9464	-4.2552***	-2.004	-2.998***	-1.853	-4.122***
$\ln\text{GDP}$	-3.8214***	-17.8966***	-0.6192	-4.6497***	-3.181***	-3.983***	-3.077***	-4.408***
$\ln\text{REC}$	-3.0013***	-16.8060***	-1.5685	-5.3679***	-2.572***	-3.900***	-2.352***	-5.213***
$\ln\text{NRR}$	0.6460	-18.8022***	-1.6593	-5.1598***	-2.024	-4.095***	-1.873	-5.340***
$\ln\text{KOFGI}$	-19.6178***	-11.4233***	-5.1407***	-4.6250***	-2.459***	-3.910***	-2.929***	-5.300***

**Table 5 Westerlund and Pedroni cointegration test.**

Pedroni cointegration test			
Test Statistic	Value	p-value	
Modified Phillips-Perron t	2.7258	0.0032	
Phillips-Perron t	-2.5531	0.0053	
Augmented Dickey-Fuller t	-1.6303	0.0515	
Westerlund cointegration test			
Statistic	Value	Z-value	P-value
Gt	-1.355	5.498	1.000
Ga	-3.765	6.875	1.000

**Table 6 CS-ARDL outcomes.**

Variables	Coefficient	Std. Err.	Z	P >  z
Long Run Estimates				
lnTI	0.0335	0.0157	2.13	0.033
lnGDP	0.3297	0.0470	7.02	0.000
lnREC	0.1585	0.0303	5.22	0.000
lnNRR	-0.0017	0.0043	-0.40	0.686
lnKOFGI	-0.0098	0.1263	-0.08	0.938
Short Run Estimates				
L(lnGG)	-0.8957	0.0321	-27.92	0.000
D(lnTI)	0.0671	0.0318	2.11	0.035
D(lnGDP)	0.6369	0.0921	6.91	0.000
D(lnNRR)	-0.0018	0.0083	-0.22	0.829
D(lnREC)	0.2990	0.0549	5.45	0.000
D(lnKOFGI)	-0.0378	0.2385	-0.16	0.874
ECM(-1)	-1.8957	0.0321	-59.09	0.000

**Table 7 CS-ARDL Income Group.**

Variable	High income	Upper Middle income	Lower Middle income
Long Run Estimates			
lnTI	0.0762 (0.0253)***	-0.0143 (0.0270)	0.0108 (0.0254)
lnGDP	0.2478 (0.0913)***	0.2626 (0.0186)***	0.3279 (0.1476)**
lnREC	0.1488 (0.0339)***	0.0965 (0.0149)***	0.1613 (0.0671)**
lnNRR	0.0060 (0.0078)	-0.0106 (0.0086)	0.0159 (0.0085)*
lnKOFGI	-0.1372 (0.1992)	-0.1439 (0.0862)*	0.0861 (0.1693)
Short Run Estimates			
L(lnGG)	-0.9686 (0.0591)***	-0.8597 (0.0486)***	(-0.8724 (0.0944)***
D(lnTI)	0.1481 (0.0513)***	-0.0257 (0.0509)	0.0273 (0.0435)
D(lnGDP)	0.5419 (0.1697)***	0.4892 (0.0382)***	0.6105 (0.2717)**
D(lnREC)	0.2906 (0.0644)***	0.1811 (0.0296)***	0.2930 (0.1164)**
D(lnNRR)	0.0130 (0.0153)	-0.0176 (0.0162)	0.0316 (0.0185)*
D(lnKOFGI)	-0.3237 (0.3777)	-0.2788 (0.1736)*	0.1692 (0.3344)
ECM(-1)	-1.9686 (0.0591)***	-1.8597 (0.0486)***	-1.8724 (0.0944)***

indicating that these economies are less resource dependance. The globalization variable which is weakly negative and significant (-0.143\*) in upper middle-income economies while insignificant in high and lower-middle income economies in Europe.

**Table 8 ARDL PMG.**

Variable	Coefficient	Std. Error	z	P-value
Long Run Estimates				
lnTI	-0.2845	0.0523	-5.44	0.000
lnGDP	0.3762	0.0745	5.05	0.000
lnNRR	-0.0594	0.0147	-4.05	0.000
lnREC	0.1477	0.0168	8.78	0.000
lnKOFGI	1.9313	0.2481	7.79	0.000
Short Run Estimates				
D(lnTI)	6128.761	6128.716	1.00	0.317
D(lnGDP)	0.4873	0.0526	9.26	0.000
D(lnNRR)	-0.0126	0.0072	-1.76	0.079
D(lnREC)	0.2994	0.0482	6.22	0.000
D(lnKOFGI)	0.1042	0.1129	0.92	0.356
C	-3465.47	3464.503	-1.00	0.317

Now in the short run the error correction mechanism (ECM) value is negative and significant in all income groups indicating that the system converges from short run to long run. In the short run green growth which is negative and highly significant in all income groups. The speed up adjustment is (-0.96\*\*\*) fast in high income economies, and slightly slow in lower middle-income economies. Technological innovation is significant in high income economies (0.148\*\*\*) showing that high income economies can immediately influence innovation for short run benefits while in other economies have insignificant results are reported. It indicating that due to delays or inefficiencies it will take time in convergence from short run to long run. Economic growth and Renewable energy consumption have positive and significant impact on the green growth of all income groups. Natural resource rent (lnNRR) is significantly only in the lower middle-income economies with coefficient value is (0.0316\*\*) indicating that short run dependency on natural resources. Thus, the overall results indicating that some deterrents are important like GDP and renewable energy, while the effectiveness of other variables like innovation capabilities, globalization depends upon the countries institutional capabilities and income.

Interestingly, the results are quite consistent in the above conclusions. In Table 8, Pool Mean Group results indicates that in the long run, Gross Domestic Product, Globalization and Renewable energy consumption have a positive and statistically significant effect on the green growth of Europe, indicating that economic development, global integration and clean energy contribute to the sustainable development in the European economies over the time. On the other hand, technological innovation and natural resource rents have negative and significant impact on green growth. The reason is that may be in these economies' innovation may not be green-oriented which degrade the environment. Although in the short run Gross Domestic Product and REC have a positive and significant impact, while natural resource rents (lnNRR) have a significant but negative impact reported. Globalization and technological innovation have statistically insignificant impact on the short run, indicating that their effects will take time. Green growth is calculating this ratio involves dividing the GDP of a country by its energy-related CO2 emissions. The resulting value indicates how much economic value is being produced for each ton of CO2 emitted. This metric can help policymakers and analysts to evaluate the effectiveness of strategies aimed at decoupling Economic growth from carbon (CO2) emissions. Overall results indicating that short run mainly attentive on economic activity & renewable energy while long term dynamics highlighting the need for environmentally focused innovation & better resource management in these regions.

### Conclusion and policy recommendations

This study provides a comprehensive empirical analysis to investigate the role and importance of Technological innovation and Renewable Energy consumption in promoting green growth or attaining sustainable development in European economies categorized into high income, upper middle income and lower middle-income groups. Using the Cross Sectional ARDL model on data spanning from 1990 to 2023, the findings of the study show that economic growth (GDP) is the important and consistent indicator of green growth in both the short and long run across all income groups. It indicates that higher income associated with improved environmental outcomes like better institutions, better infrastructure and higher investments in clean technologies. Technological innovation has a positive and significant impact on green growth in high-income European economies, probably due to stronger institutional frameworks, access to clean technologies and larger investment in environmentally friendly research and development. However, its effect in middle- and lower-income economies are statistically insignificant or negative, suggesting that in less developed economies, technological innovation may be either pollution-intensive or insufficiently oriented towards environmental sustainability. Similarly, the renewable energy consumption has also positive and statistically significant impact on green growth indicating that clean energy is the important indicator for achieving sustainable environmental goals.

These findings have several economic implications. First, they support the environmental Kuznets curve may hold in higher income economies of the Europe while less evident for low-income economies. Secondly, the divergence in the role of globalization and technological innovation across all income economies emphasizing the need for differentiated policy strategies that will handling the structural weakness and institutional capabilities of developing economies.

A practical example that validates the applicability of this study is the Germany's Energiewende which was a long-term strategic initiative aimed to shift Germany to a greener energy future. Germany has greatly invested in technological innovation, specifically in energy efficiency, smart grid and alternative energy sources like wind and solar power. These innovations have played a critical role in discontinuing the correlation between economic development and environmental deterioration, and Germany takes the lead among European green growth. The implications of this study, particularly on technological innovation and renewable energy in achieving sustainable development, are clearly evident in the German example.

Keeping in consideration this study's findings, the current study suggests that the policymakers especially middle- and lower-income economies should prioritize aligning technological innovation with environmental goals by supporting R&D, cleantech startups and innovation incentives. Additionally, the positive role of renewable energy across all income groups highlights the importance of scaling up clean energy infrastructure and offering policy incentives like subsidies, tax breaks and feed-in tariffs. As far as renewable energy is concerned, government must prioritize substantial investment in renewable energy infrastructure and providing the above-mentioned incentives which can help to enhance the green growth outcomes. Due to globalization, economies should focus on strengthening the environmental standards that aligned with the sustainable objectives.

Furthermore, in the context of uncertainty like climate policy shifts, war, covid etc., rising uncertainties can harm the countries' efforts for sustainability (Ongan et al., (2025)). Therefore, it is recommended that European countries need to embrace adaptive green strategies. These encompass investments in resilient low-carbon technologies and the promotion of renewable energy

using flexible incentives. Public-private partnerships and regional collaboration can facilitate optimal technological diffusion and ensure stable green growth with uncertain conditions.

However, there are some limitations to this study. First, although the CS-ARDL model is capable of modeling short- and long-run relationships, it makes homogeneity of long-run coefficients that might not completely capture unique country-specific dynamics. Second, data unavailability in certain countries can limit generalizability of findings. Future research can use other proxies for sustainable development, i.e., environmental quality indices or inclusive growth indicators instead of green growth. Moreover, nonlinear or country-specific models can be employed in future research to realize more heterogeneous effects. Finally, combining policy variables like environmental policy or innovation-friendly fiscal policy can further increase the explanatory power of the model as well as its practical relevance.

### Data availability

Data will be available on request to the corresponding author.

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### Author contributions

Yangfan Lu did the formal analysis, proof reading and reviewed the draft. Sareer Ahmad did the supervision, conceptualized, and project administration, Shaista Noureen wrote the original draft, conducted the econometric analysis. Muhammad Salman wrote the literature review, methodology and review & editing of the manuscript.

### Competing interests

The authors declare no competing interests.

### Ethical approval

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Not applicable

### Additional information

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