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Original Research

Assessment of the environmental kuznets curve within EU-27: Steps toward environmental sustainability (1990–2019)



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ABSTRACT

Reducing environmental pollution is a critical goal in global environmental economics and economic development. The European Union (EU) faces environmental challenges due to its development activities. Here we present a comprehensive approach to assess the impact of carbon dioxide (CO₂) emissions, energy consumption (EC), population structure (POP), economy (GDP), and policies on the environment within the EU using the environmental Kuznets curve (EKC). Our research reveals that between 1990 and 2019, the EU-27 experienced an increase of +1.18 million tonnes of oil equivalent (Mtoe) per year in energy consumption (p < 0.05), while CO₂ emissions decreased by 24.25 million tonnes (Mt) per year (p < 0.05). The highest reduction in CO₂ emissions occurred in Germany (-7.52 Mt CO₂ annually), and the lowest in Latvia (-0.087 Mt CO₂ annually). The empirical EKC analysis shows an inverted-U shaped relationship between GDP and CO₂ emissions in the EU-27. Specifically, a 1% increase in GDP results in a 0.705% increase in carbon emission, while a 1% increase in GDP² leads to a 0.062% reduction in environmental pollution in the long run (p < 0.01). These findings indicate that economic development within the EU has reached a stage where economic growth positively impacts the environment. Overall, this study provides insights into the effectiveness of environmental policies in mitigating degradation and promoting green growth in the EU 27 countries.

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1. Introduction

The global economy has experienced remarkable growth in the

*Corresponding author. Institutes for Agricultural Research and Educational Farm, University of Debrecen, Böszörményi 138, H-4032, Debrecen, Hungary. E-mail address: safwan@agr.unideb.hu (S. Mohammed). past century due to various factors, such as globalization, industry booms, innovation, significant technological development, free international trade at multiple levels and sectors, and the overall economic environment. Nevertheless, this economic progress has adversely affected the environment, leading to heightened environmental pollution and intensified environmental degradation [1]. Economic activities have played a crucial role in promoting

development and growth but have also contributed to the emission of greenhouse gases (GHGs), such as carbon dioxide (CO_2), methane (CH_4), and nitrous oxide (N_2O). These GHGs have contributed to environmental changes, ecosystem degradation, and rapid climate change [2,3]. Various factors influence GHG emissions, including population increase, per capita consumption and output growth, infrastructure choices, innovation, technology, and human behavior [4–6]. However, if mitigation measures are insufficient, economic growth and human and social development potentials will likely decline because of climate change [7].

Environmental degradation remains a main concern in the European Union (EU). Therefore, any operations that involve rising energy consumption (EC) and the exhaustion of carbon (CO₂) have effects on the environment [8,9]. On the other hand, production and consumption activities are linked to energy use; thus, energy is critical for economic growth. Consequently, the EU's foundation for economic activities and social well-being is secure [10]. The EU's historical reliance on fossil fuels as an energy source has notably escalated the emission of potentially hazardous gases [11]. In June 2021, EU-27 countries set the European Green Deal under the auspices of European Climate Law (2021/1119) to reduce GHGs, aiming at 'net-zero' emissions by 2050 [12]. The law's primary goal is to decrease GHG emissions by 55% by 2030 compared to 1990 ('Fit for 55') [13]. The EU has targeted reducing CO₂ emissions by 310 million tons using natural sinks, demonstrating its leadership in pursuing carbon neutrality. The 'Fit for 55' initiative is the most important regulation package to reach the targets of the Paris Agreement, keeping global warming below 1.5 °C. At its establishment, acceptance of 'Fit for 55' was crucial. If a robust framework exists, the EU stands a decent chance of leading the global fight to attain net-zero emissions.

The EU countries have implemented policies related to climate change mitigation, energy security, energy efficiency, use of renewable energy, and others to curb environmental degradation and recorded considerable progress. This article expands the existing literature by providing a holistic approach to evaluating the impact of energy consumption, population structure, economy, and policies on the environmental sector across the EU through the lens of the environmental Kuznets curve (EKC).

The main aims of this study were to (1) comprehensively report on the contemporary changes (trends) in GHG emissions and energy consumption across the EU-27 between 1990 and 2019 and (2) highlight the interaction between GHG emissions, energy consumption, population, and gross domestic product (GDP) by employing the EKC. The study's main focus was to address a crucial question: the effectiveness of EU-27 environmental policies in mitigating environmental degradation. To accomplish these goals, the article is organized into four sections: (1) literature review, which presents a comprehensive overview of the relevant literature, (2) material and method, (3) findings of this research, (4) discussion of research output concerning the research limitations.

2. Literature review

Climate change and global warming are among the worst externalities exclusively being debated by scientists, policymakers, governments, and societies [14]. The primary focus is on the escalating GHG levels, especially CO₂ emissions, which account for 60% of GHGs [15]. Increased GHG emissions reduce environmental quality by increasing environmental pollution [14], which is detrimental to sustainable development [16]. Nations are currently grappling with twin challenges of achieving and maintaining economic growth and environmental quality [17,18] exacerbated by globalization trade openness, urbanization development, and prosperity of the service sector, causing relevant structural changes

in trade, energy, economy, and society [19]. Hence, decreasing pollution has become a significant concept in the environmental economics and economic development model [7]. To understand this nexus, economist Simon Kuznets introduced the EKC in 1955 [20], later formalized by Grossman and Krueger [21]. The EKC indicates that, initially, environmental degradation is caused by economic growth in the first production stage. In addition, the EKC illustrates the probable relationship between economic growth and rising income levels instead of the effect on economic inequality, demonstrating that economic disparity rises and falls as the economy grows [22]. However, the tendency may be inverse; it may decouple the effects of progress on the environment because economic progress can lead to environmental improvements and a virtual reduction in per capita emissions of national production. This relationship is nonlinear between emissions per capita or unit of GDP and wealth [23]. Therefore, EKC serves as a metaphor for understanding the fundamental mechanisms of public policy interventions that reduce economic growth from harming the environment [24].

Several studies have analyzed the nexus between the environment and economic performance indicators such as GDP, trade, urbanization, and others at national, regional, and global levels (Table 1). Many empirical investigations have yet to reach consistent conclusions on the connection between CO₂ and GDP [18] since studies select different input variables, time, and regional scopes.

The findings of several studies exploring the relationship between carbon emissions and economic growth in various countries and regions have been documented in previous research. Le [25] surveyed across ten Association of Southeast Asian Nations (ASEAN) countries and revealed a U-shaped EKC in the early development stages, which later transformed into an N-shaped EKC hypothesis in the long run. This suggests negative and positive causality between carbon emissions and economic growth. Yuelan et al. [26] studied between 1980 and 2016 in China, revealed that energy usage, GDP, total revenue, and government expenditures deteriorated environmental quality. Analysis of the BRICS countries (i.e., Brazil, Russia, India, and China) between 1995 and 2014 showed that the GDP² and GDP had negative and positive effects (valid EKC), while urbanization and globalization have a negative insignificant relationship with CO₂ emission. The report revealed a two-way causality between EC, GDP², financial development, and economic growth with CO₂ emission [27]. Phong [28] studied five ASEAN countries between 1971 and 2014 and confirmed the EKC hypothesis. Results further revealed that except per capita GDP, globalization, EC, urbanization, and financial development increased CO2 emissions.

The latest study by Wang et al. [18] indicated that CO₂ emissions (per capita) nexus with income (per capita) resulted in an "inverted U-shaped" EKC after human capital, trade openness, and natural resource rents were included as variable inputs. Renewable energy consumption had a more substantial carbon emission reduction effect before the EKC turning point, while human capital was after the EKC turning point. Ben Jebli et al. [29] reported an inverted Ushaped EKC between per capita CO₂ emissions and GDP in the 25 OECD (Organization for Economic Co-operation and Development) countries. Balsalobre-Lorente et al. [14] reported that renewable energy (RE) inhibits CO₂ emissions while urbanization exerts vast pressure on environmental quality. Sinha and Shahbaz [30] in India, covering a period between 1971 and 2015, reported an inverted Ushaped EKC. The study revealed that RE had a significant adverse effect on CO₂ emissions. Le and Ozturk [31] studied 47 developing and market economies between 1990 and 2014 and confirmed the inverted U-shaped EKC; however, financial development, globalization, and energy consumption increased CO2 emissions. In Vietnam, industrialization, per capita GDP, and EC raised CO₂

Table 1Survey of econometric literature of EKC implementation showing the nexus between environmental indicators and carbon emissions from a global perspective.

Country	Period	Sector	GHGs	Input variable	Output	Reference
208 countries	1990 -2018	Industry	CO ₂	Renewable energy (RE), GDP, trade, human capital index, natural resources rents	Income level and CO ₂ emissions had EKC relation	[18]
PIIGS countries (Portugal, Ireland, Italy, Greece, and Spain)	1990 -2019	Industry	CO ₂	Economic complexity index (ECI), foreign direct investment, RE, urbanization process	ECI. and CO ₂ emissions had inverted-U and further <i>N</i> -shaped EKC	[14]
Seven Countries (Emerging industrialized economy)	1995 -2016	Industry	CO ₂	GDP, RE, institutional quality	EKC validated. Renewable energy reduces pollution. Weak institutions reduce environmental quality	[87]
18 Emerging economies/countries	1990 -2015	Industry	CO ₂	Trade openness RE, non- RE, GDP, GDP ² .	RE affects CO ₂ emissions negatively and reverses renewable energy	[88]
BRICS countries	1995 -2014	Industry	CO ₂	EC, financial development, GDP, GDP ² , and globalization urbanization.	GDP and GDP ² had negative and positive effects (valid EKC)	[27]
26 E U. countries	1995 -2018	Industry	CO ₂	ECI, Brexit, and other crisis episodes	Tourism, energy use, real GDP per capita increased emissions.	[6]
47 countries	1990 -2014	Industry	CO ₂	Government consumption expenditure, government expenditures, GDP, GDP ² , EC, financial development, globalization, and institutional quality	In either scenario EKC hypothesis EKC confirmed. EC, financial development, and globalization increased CO ₂ emissions	[31]
India	1971 -2015	Industry	CO ₂	GDP and RE	Inverted U-shaped EKC Renewable energy has a significant negative effect on CO ₂ emissions.	[30]
25 OECD countries		Industry (energy and trade sector)		Trade, RE, GDP, and non-RE consumption	Inverted U-shaped EKC. Increased non- RE increased ${\rm CO_2}$ emission.	[29]
134 countries	1996 -2015	Industry	CO ₂	Trade openness, population aging, RE, and natural resource rent	Inverted U-shape EKC in lower-middle- income countries. Renewable energy and population aging improve the environment	[33]
Ten ASEAN countries	1993 -2014	Industry	CO ₂	GDP, foreign investment, trade openness index, and urbanization	Earlier inverted U-shaped and a later N-shaped EKC	[25]
27 E U. countries	1995 -2017	Industry	CO ₂	Economic complexity Index, energy intensity or consumption	Inverted U-shaped curve. 10% rise in energy intensity increased CO ₂ emissions by 3.9%	[37]
Venezuela	1971 -2013	Industry	CO ₂	GDP, GDP ² , EC, financial development, and government expenditure	EKC confirmed. Government expenditure positively impacted on environmental degradation.	[35]
Five ASEAN countries	1971 -2014	Industry	CO ₂	GDP, financial development, EC, and globalization urbanization	1 0	[28]
Malaysia	1970 -2016	Industry	CO_2	Natural resources (GDP)	Conventional EKC pattern not supported.	[89]
16 lower income, 26 lower middle income, 26 higher middle income, and 31 high-income countries			CO ₂	Trade openness, financial development, GDP, EC, and urbanization,	Inverted U-shaped nexus between ecological footprint and GDP in high- and middle-income countries	[34]
56 countries	2003 -2018	Industry	CO ₂	Income inequality	Income inequality had a double-threshold effect on the economic growth and carbon per capita emission.	[17]

emissions, while globalization lowered CO₂ emissions in the long run [32]. Meanwhile, Wang et al. [33] also reported RE and population aging to improve environmental quality.

According to the research findings, income inequality had a double-threshold effect on economic growth and carbon per capita emission. It redefined the EKC hypothesis from an inverted Ushaped curve to an N-shaped curve depending on the income groups making decoupling carbon emission and economic growth complex [17]. Another study by Al-mulali et al. [34] found an inverted U-shaped EKC between ecological footprint and GDP in the economic development stage of high- and middle-income countries with better energy saving, efficiency, and use of renewable energy. Therefore, improving financial investments targeting the reduction of carbon emissions is vital since financial investments reduce environmental damage [34]. Financial development enhanced environmental quality while government expenditure increased the pollution level in Venezuela [35]. Therefore, comprehensive energy and economic policies targeting cleaner production should be advocated due to the positive nexus between economic complexity and environmental quality [14]. A robust and efficient financial sector can play a role in environmental protection by promoting environmentally friendly and energy-efficient projects and advancing loans for businesses with conditions of reducing CO₂ emissions [27].

The reviewed literature indicates the relationship between CO₂ emissions GDP, income inequality, renewable energy consumption, and CO₂ emissions within the EKC hypothesis, wherein CO₂ was an indicator of environmental harm. Although most studies have focused on these variables, only one study included the population as a variable input. By employing the Vos viewer for bibliographic analysis [36], recent research between 2018 and 2019 focused on understanding the relationship between climate change, environmental degradation, economic development, gross domestic product, energy use, and sustainability using EKC (Fig. 1). In this context, a study by Balsalobre-Lorente et al. [14] confirmed the existence of an inverted U- and later N-shaped EKC nexus between CO₂ emissions and economic complexity, foreign direct investment, renewable energy, and urbanization process in PIIGS countries (i.e., Portugal, Italy, Ireland, Greece, and Spain) in the period of 1990–2019. Accordingly, in the 25 EU countries from 1995 to 2017,

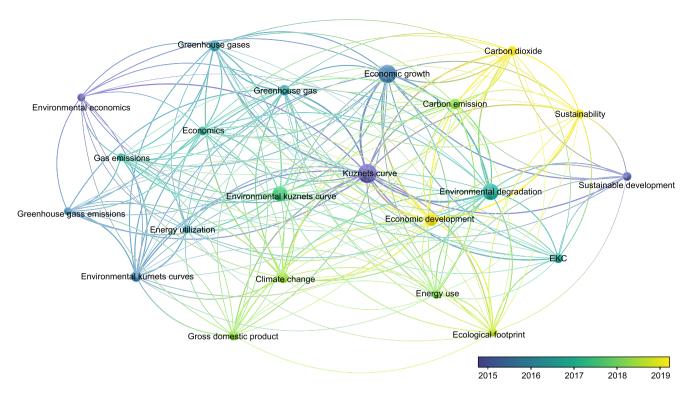


Fig. 1. Bibliographic analysis using Vosviewer software exhibits the relationship between climate change, environmental degradation, economic development, GDP, GHGs, energy use, and sustainability.

a relationship between CO₂ emissions and economic complexity produced an inverted U-shaped EKC with a 10% rise in energy intensity, increasing CO₂ emissions by 3.9% [37]. On the other hand, another study [38] explored the connection between environmental degradation and energy innovation and economic growth in 33 EU countries from 1996 to 2017. Results indicated that energy innovation negatively and significantly impacted environmental degradation, while GDP produced a U-shaped EKC. Cai et al. [5] also revealed no cointegration between the consumption of clean energy, real GDP, and CO2 emissions in France and Italy except in German and other non-EU countries, such as Canada and the U.S. Overall, different studies showed non-unidirectional relationship effects between economic indicators and environmental quality. emphasizing the necessity for continuous evaluation. Besides, our study differs from these by introducing population as a controlling variable input. The EU27 exhibits varying population structures, leading to different patterns in energy consumption, land resource utilization, and urbanization development. Earlier, Wang et al. [33] reported the aging population could positively impact environmental quality.

3. Material and method

3.1. Data collection and trend analysis

Data for EU countries (1990–2019) were collected from the website of the European Commission, DG Energy, Unit A4. Data included (1) CO₂ emissions - National total (including International Aviation, without Land Use, Land-Use Change, and Forestry (LULUCF)) (Mt CO₂), (2) energy consumption (Mtoe), (3) GDP-market prices (Mrd EUR at current prices), and (4) population (thousands of people).

To capture the trend and magnitude of change in each variable,

the Mann–Kendall test (MK_t) [39] and Sen's slope estimator (ρ) [40] were employed. MK_t has been widely used in trend analysis, and it is calculated as shown in equation (1):

$$MK_t = \sum_{k=1}^{n-1} \sum_{j=k+1}^{n} sign(x_j - x_k)$$
 (1)

where x_j, x_k are the values for the studied variable in years j and k (j > k); n is the sum of the years; and $\text{sign}(x_j - x_k)$ is the sign function.

The variance of equation (1) and (∇_{MK_t}) can be denoted as equation (2):

$$\nabla_{MK_t} = \frac{n(n-1)(2n+5) - \sum_{i=1}^{m} t_i(t_i-1)(2t_i+5)}{18}$$
 (2)

However, the Z statistic (Z) can be estimated as equation (3):

$$Z = \begin{cases} \frac{(MK_t) - 1}{\sqrt{\nabla_{MK_t}}}, & \text{if } MK_t > 0\\ 0, & \text{if } MK_t = 0\\ \frac{(MK_t) + 1}{\sqrt{\nabla_{MK_t}}}, & \text{if } MK_t < 0 \end{cases}$$
(3)

The p-value for equation (3) is estimated based on equation (4):

$$p = 0.5 - \frac{1}{\sqrt{\pi}} \int_{0}^{|Z|} e^{-\frac{t^2}{2}} dt \tag{4}$$

The magnitudes of the studied variables are captured using

Sen's slope (Ss) [40] in equation (5):

$$Ss = \frac{x_j - x_k}{i - k}$$
 for $i = 1, 2, 3, ..., n$ (5)

In this case, Sen's slope (Ss) (ρ) is the median of the time series (Γ) for n values and is denoted in equation (6):

$$\rho = \begin{cases} \Gamma_{\frac{n+1}{2}}, & n \text{ is odd} \\ \frac{1}{2} \left(\Gamma_{\frac{n}{2}} + \Gamma_{\frac{n+2}{2}} \right), & n \text{ is even} \end{cases}$$
 (6)

where a positive value of ρ indicates an increasing trend, while a negative value refers to a declining tendency, finally, the results were visualized using the geographical information system (GIS).

3.2. Trend change point detection

To identify the trend-changing points (years) in the time series, a homogeneity nonparametric Buishand range (BR) [41] test was applied. The BR test is based on the null hypothesis assumption that the variables are independent of each other with a normal distribution, while the alternate hypothesis assumes a shift in the mean of the time series. It is probably sensitive to find breaks in the middle of the time series. In the BR test, rescale adjusted partial sums (S_k) are computed by:

$$S_k = \sum_{i=1}^k (X_i - \overline{X}) \ k = 1, 2, 3, ..., n$$
 (7)

where X_i represents the observed data series, \overline{X} represents the mean, and n represents the total data numbers in the time series [42]. The value of S_k rotates around zero for a homogeneous time series. If there is a break in year k, S_k reaches the maximum negative of the maximum positive shift [43]. The equation computes significant change shift:

$$R = \frac{\max_{s_k} - \min_{s_k}}{\overline{X}} \tag{8}$$

3.3. The EKC analysis

The legitimacy of the EKC in 27 European countries has been examined based on the following model, which is formulated depending on different studies [44,45].

$$\begin{split} LN(\text{CO}_2)_{it} &= \beta_1 + \beta_2 LN(\text{GDP})_{it} + \beta_3 \ LN(\text{GDP})_{it}^2 + \beta_4 LN(\text{EC})_{it} \\ &+ \beta_5 LN(\text{POP})_{it} + U_{it} \end{split} \tag{9}$$

In this equation, β_1 is constant; EC, POP, and U_{it} represent energy consumption, population, and the error term. The term (it) subscript indicates panel data: i for countries and t for the time period. The quadratic term of GDP is included to model the nonlinear relationship between GDP (income) and CO_2 (the environment).

The EKC hypothesis is verified in the context of the EU-27 if the estimation of equation (9) proves that $\beta_1 > 0$ and $\beta_2 < 0$. Two important parameters, EC and POP, are included to run the model properly. The sign of the coefficient of energy consumption and

population growth is expected to be positive [46,47]. The data about the variables under investigation covers the period from 1990 to 2019. All data were transformed into the logarithmic form to obtain the elasticities of carbon emissions for explanatory variables: therefore, coefficients were considered long-run elasticities.

The cross-sectional dependency test of Pesaran [48] is employed to ascertain whether each panel is cross-sectionally independent, where the null hypothesis is that the panels are cross-sectionally independent.

We employed a three-stage estimation method to estimate the coefficients of equation (9). In the first stage, the first and second generations of the panel unit root test were applied to examine the stationarity properties of the panel data, where the following unit root tests, Im and Pesaran [49], Levin, Lin and Chu [50], and Maddala and Wu [51], were applied. Moreover, Pesaran [52] altered the I.P.S. test by incorporating cross-sectional dependence (CIPS). However, for each panel data presenting nonstationary, the first difference was taken to make it stationary (H_0 : the panel data are nonstationary); thus, the cointegration relation among the variables was investigated by applying panel cointegration tests developed by Pedroni [53] and Kao [54]. Pedroni [53] developed seven test statistics based on the cointegration regression in equations (10) and (11), which we used to test this null hypothesis that there is no cointegration relation among the variables in equation (9).

$$y_{it} = \alpha_{it} + \delta_{it}t + \beta_i x_{it} + \varepsilon_{it}$$
 (10)

$$\Delta y_{it} = \beta_i \Delta x_{it} + n_{it} (n_{it} = \varepsilon_{it} - \varepsilon_{it-1})$$
(11)

In equation (10), y_{it} and x_{it} are dependent and explanatory variables, respectively, while ε_{it} and α_{it} indicates residual and fixed effects. In equation (11), Δ is the difference operator. Moreover, referring to Kao [54], the Pedroni cointegration relation was verified. We also employed the Westerlund [55] panel cointegration tests (Ga, Gt, Pa, and Pt) based on the error correction model (ECM), as it has several advantages compared to other panel cointegration methodologies.

Firstly, it enables heterogeneity in short- and long-run dynamics. Secondly, it overcomes the problem of cross-sectional dependency. Lastly, this approach relies upon the bootstrap procedure that includes repeating various cointegration methods [56].

The long-run co-integration relation between the dependent and explanatory variables of equation (9) was also estimated. The dynamic ordinary least squares (DOLS) and fully modified ordinary least squares (FMOLS) panel cointegration estimates developed by Pedroni [57] are commonly used in literature. These estimates eliminate the endogeneity and autocorrelation problem and therefore produce efficient estimates.

However, if the panel sections are correlated and have a cross-sectional dependency, the FMOLS and DOLS may produce inefficient estimates despite having high statistical power and robustness. Eberhardt and Teal [58] and Bond and Eberhardt [59] therefore proposed a new panel cointegration estimation method called augmented mean group (AMG) that includes a common dynamic process to account for cross-sectional dependency in panel data through a two-stage regression method as in equation (12). In this context, cross-sectional dependence is a common phenomenon of panel data resulting from policy shocks and sociopolitical linkages among countries. These effects are heterogeneous and of different degrees, leading to inefficient estimation, especially when correlated to explanatory variables [60,61].

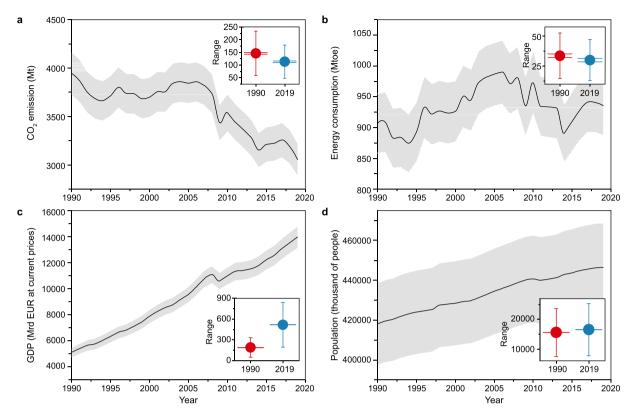


Fig. 2. An overview of temporal trends of the studied variables (CO₂ emissions, energy consumption, GDP, population) for all EU-27 countries (total) between 1990 and 2019 along with box plot (I-shaped box): blue color represents 2019, red color represents 1990 (•: mean; |: mean ± 95% confidence interval; gray shade around the black line represents the standard deviation between E.U. countries for each year from 1990 to 2019).

AMG Stage (i)
$$\Delta$$
EFP_{it} = $b_0\Delta X_{it} + X_{it} = 2 c_{it}\Delta D_t + e_{it} \Rightarrow c_{it} \equiv \mu_{it}$ (12)

AMG Stage (ii)
$$\Delta$$
EFP_{it} = $b_o \Delta X_{it} + X_{it} = 2 c_{it} \Delta D_t + e_{it} \Rightarrow c_{it} \equiv \mu_{it}$ (13)

In equations (12) and (13), Δ X, b, and D represent difference operators, explanatory variables, coefficients of explanatory variables, and dummy variables. Dummy variables capture the impact of time, while c_{it} is the coefficients on the dummy variables, μ_{it} is an error term, t is time, and t is cross sections in the panel data regression. Compared to other panel estimation methods, AMG considers cross-sectional dependencies and heterogeneity issues in estimation, producing more efficient estimates [60,62].

The first stage equation represents a standard Ordinary Least Squares (OLS) regression in the first difference with dummies in the first differences ΔDt , from which we collect the estimated coefficients, which are relabeled as μ_{it} . Because nonstationary variables and unobservables can substantially distort results in pooled levels regressions, this process is extracted from the pooled regression in first differences. In the second stage, this constructed variable μ_{it} is included in each of the N group-specific regressions, which also include linear trend terms to capture omitted idiosyncratic processes that evolve linearly over time. Alternatively, we can subtract μ_{it} from the dependent variable, implying that the common process is imposed on each group with a unit coefficient. In either case, the AMG estimates are derived as averages of the individual b_0 estimates, following the Mean Group approach proposed by Pearson and Smith [63].

4. Results

4.1. Trend analysis of EKC parameters within the EU-27 (1990–2019)

Between 1990 and 2019, the EU-27 countries experienced a decrease in CO_2 emissions alongside increases in GDP and population (Fig. 2). In this context, the total CO_2 emission declined from 3945.0 Mt CO_2 in 1990–3054.7 Mt CO_2 in 2019 (Fig. 2). The EC fluctuated between 906.56 Mtoe in 1990 and 935.50 Mtoe in 2019, where the highest value was 990.13 Mtoe in 2006 (Fig. 2). Most countries exhibited a notable decrease in CO_2 emission and an increase in GDP (Fig. 3).

EC increased in the EU-27 countries between 1990 and 2019 (Fig. S1a): the general trend was significant (p < 0.05), and the change rate was growing by +1.18 Mtoe per year. Trends of EC significantly (p < 0.05) increased in Spain (ES), Poland (PL), Austria (AT), Italy (IT), Ireland (IE), Portugal (PT), and Finland (FI) with positive slope and Z (Ss = 0.91-0.14, Zs = 3.46-2.39) with 1999, 2007, 2001, 1998, 1999, 1998, and 2000, respectively; as turning points of trend in time series (Fig. 4a). However, EC decreased in Germany (DE), Romania (RO), Czech Republic (CZ), and Sweden (S.E.) with negative slope and Z (Ss = (-0.54)-(-0.07)) (Fig. S1a). Further, the BR test identified significant turning trend years, including 2006, 1998, 1997, and 2004 for respective countries (Table S1) (Fig. 4a). The highest increase was recorded in Spain (+0.92 Mtoe; p < 0.05), while Germany had the largest drop (-0.55 Mtoe; p < 0.05).

The M.K. test showed a significant decline in CO_2 emissions in the EU-27 (Fig. S1b). The total CO_2 emissions from the EU-27 were reduced significantly (p < 0.05) by 24.25 Mt CO_2 per year. Countries

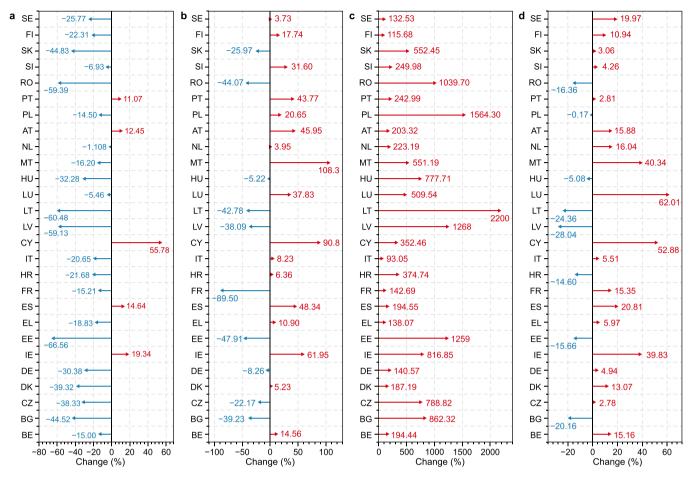


Fig. 3. Changes (%) in EU-27 studied variables between 1990 and 2019: **a**, CO₂ emissions; **b**, energy consumption; **c**, GDP; **d**, population. Blue arrow: decreasing trend; red arrow: increasing trend (Changes (%) = \frac{Value of studied variables_{2019} - Value of studied variables_{2019} - Value of studied variables_{1990}} \times 100) (Abbreviation for the Member States of the European Union (EU) as suggested by: https://ec.europa.eu/eurostat/statistics-explained/index.php?title=Glossary:Country_codes). Abbreviations: Belgium (BE), Bulgaria (BG), Czechia (CZ), Denmark (DK), Germany (DE), Estonia (EE), Ireland (IE), Greece (EL), Spain (ES), France (FR), Croatia (HR), Italy (IT), Cyprus (CY), Latvia (LV), Lithuania (LT), Luxembourg (LU), Hungary (HU), Malta (MT), Netherlands (NL), Austria (AT), Poland (PL), Portugal (PT), Romania (RO), Slovenia (SI), Slovakia (SK), Finland (FI), Sweden (SE).

such as Germany (DE), Italy (IT), France (FR), Romania (RO), Poland (PL), Denmark (DK), and the Czech Republic (CZ) showed a significant negative (p < 0.05) trend with (Ss = (-7.4)-(-1.2)) and Zs = (-6.8)-(-1.8)) with significant (p < 0.05) trend declining year of 2004, 2008, 1998, and 2007 identified from the BR test (Fig. 4b) (Table S1). On the contrary, Spain (ES), Croatia (HR), and Portugal (PT) showed a positive but non-significant (p > 0.05) trend of CO_2 emissions with a positive slope and Z-statistics (Ss = 1.31-0.01, Zs = 1.85-0.07)). Overall, the highest reduction in CO_2 emission was recorded in Germany (DE), with -7.52 Mt CO_2 annually, while the lowest reduction was recorded in Latvia (LV), with -0.087 Mt CO_2 annually.

Furthermore, GDP and POP also had a positive significant (p < 0.05) trend (1990–2019) (Figs. S1c and d). The annual change rate in the total GDP increased by +301.56 Mrd EUR at current prices per year (p < 0.05), while the POP increased by +1001.3 thousand people per year (p < 0.05). The highest increase in the GDP was in Germany (DE) (p < 0.05), followed by France (p < 0.05), while the lowest increase was in Malta (MT) (Fig. S1c). Regarding the POP, the highest increase was in France (FR) (+342.1 thousand people per year, p < 0.05), with the highest S of 341 and a sharp rise since 2004 as identified from the BR test (Fig. S1d). Notably, Romania (RO) had the highest decreasing rate of -140.57 thousand people per year (p < 0.05) (S = -140.5 and S = -7.7) with a

turning point since 2005 as identified from the BR test (Table S1) (Fig. 4d).

4.2. EKC analysis

4.2.1. Cross-sectional dependency (CD) and Slope Homogeneity tests for EKC analysis

According to the CD and Slope Homogeneity tests, the EU-27 remained expectedly dependent on each other due to the free trade between them during the study period (p < 0.01; Table 2).

4.2.2. Panel unit root tests

Different results were obtained by using different stationarity tests where the null hypothesis was rejected for some variables at their first level (Table 3); however, it was rejected after taking the first difference for all series, which therefore became stationary (p < 0.01) (Table 3).

4.2.3. Panel cointegration tests

In the second stage, five out of seven Pedroni test statistics were supported by the test statistics of Kao [54] and Johansen [64], rejecting the null hypothesis of no cointegration (p < 0.01) (Table 4), which verified the long-run cointegration relationship among the variables of equation (9) (GDP, EC, POP), in the study period (Table 4).

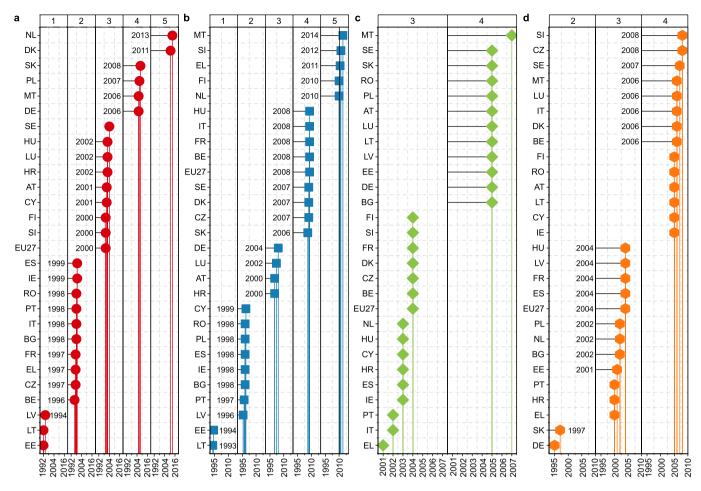


Fig. 4. Changing years in the time series (1990–2019) in EU-27 countries for studied variables based on the BR test: **a**, energy consumption; **b**, CO₂ emissions; **c**, GDP; **d**, population. 1: changes between 1990 and 1994; 2: changes between 1995 and 1999, 3: changes between 2000 and 2004, 4: changes between 2005 and 2010, 5: changes after 2010. Abbreviations: Belgium (BE), Bulgaria (BG), Czechia (CZ), Denmark (DK), Germany (DE), Estonia (EE), Ireland (IE), Greece (EL), Spain (ES), France (FR), Croatia (HR), Italy (IT), Cyprus (CY), Latvia (LV), Lithuania (LT), Luxembourg (LU), Hungary (HU), Malta (MT), Netherlands (NL), Austria (AT), Poland (PL), Portugal (PT), Romania (RO), Slovenia (SI), Slovakia (SK), Finland (FI), Sweden (SE).

In the same context, Westerlund [65] illustrated that (Gt, Ga, Pt, and Pa) statistic tests rejected the null hypothesis of no cointegration (p < 0.01) (Table 5), which also confirms the long-run

Table 2 Cross-sectional dependency (CD) in the panel data.

	J (/				
Test	LCO ₂	LGDP	LGDP ²	LEN	LPOP
CD-test statistics p-value	33.33 ^a 00.00	92.34 ^a 00.00	92.34 ^a 00.00	12.00 ^a 00.00	9.46 ^a 00.00

^a Rejection of H_0 at 0.01.

cointegration relationship between LCO₂ and LGDP, LGDP², LEC, and LPOP despite the presence of cross-sectional dependencies.

4.2.4. Long-run estimates of panel cointegration

The signs on coefficient and elasticities of carbon emission for income, income square, energy consumption, and population did not differ significantly in FOMLS, DOLS, and AMG estimation (except for POP). Moreover, due to the data's cross-sectional dependency, we considered the third-generation estimation methodology (i.e., AMG) since it is more reliable [66]. The significant positive sign on LGDP and significant negative sign on income

Table 3 Stationarity panel unit root tests.

Variables	Level	Level				First Difference			
	LLC	IPS	PP-Fisher	CIPS	LLC	IPS	PP-Fisher	CIPS	
LCO ₂	-2.68*	-1.24	121.8*	-2.50*	-6.74*	-12.60*	262.6*	-4.86*	
LGDP	-1.07	-1.31	75.59**	-2.38*	-7.20*	-9.44*	190.50*	-4.40*	
$LGDP^2$	-1.07	-1.31	75.59**	-2.38*	-7.20*	-9.44*	190.50*	-4.40*	
LEC	-4.78*	-3.37*	145.99	-2.99*	-7.13*	-11.21*	229.06*	-4.61*	
LPOP	-2.12*	-0.19	78.59*	-1.24	-2.75*	-3.57*	99.31*	-2.78*	

^{*}and ** significance at 0.01 and 0.05, respectively. The lag length of all unit root tests is determined using the A.I.C. Newey–West bandwidth selection with the Bartlett Kernel is used for the L.L.C. test..

Table 4 Panel cointegration tests.

Tests	Statistics	Probability	Johansen Fisher Panel Coi	Johansen Fisher Panel Cointegration Test		
Panel V-statistics	3.95**	0.042	Number of C.E. (s)	Trace test statistics	Probability	
Panel rho-statistics	1.81	0.9652	None	610.0*	0.0000	
Panel PP-statistics	-2.26**	0.011	At most 1	283.6*	0.0000	
Panel ADF-statistics	-5.92*	0.000	At most 2	152.8*	0.0001	
Group rho-Statistic	1.89	0.971	At most 3	97.48***	0.0097	
Group PP-Statistic	−7.96 *	0.000	At most 4	87.82**	0.0027	
Group ADF-Statistic	-5.00*	0.000	* Probabilities are compu	ted using asymptotic Chi-square dis	stribution	
Kao (1999)	-6.96*	0.000	_			

^{*}and ** rejection of H_0 at 0.01 and 0.05, respectively.

Table 5 Westerlund ECM. panel cointegration tests.

Panel cointegration tests	Statistics	<i>p</i> -value
Gt	-3.141	0.000*
Ga	-12.993	0.067***
Pt	-15.187	0.000*
Pa	-13.486	0.02**

 H_0 : no cointegration within the EU-27 series and four covariates (CO₂ emissions, energy consumption, GDP, population); *, **, and ***rejection of H_0 at 0.01, 0.05, and 0.1 respectively..

square (LGDP²) indicated the existence of an EKC relation between income and CO_2 emissions within the EU-27 countries. Thus, based on the AMG model, an increase of 1% in GDP will lead to a rise of 0.705% in environmental degradation, while a 1% increase in GDP^2 will reduce environmental pollution by 0.062% in the long run (p < 0.01) (Table 6). This implies that economic development in EU-27 reached the stage where economic growth improves the environment. On the other hand, the positive sign on the coefficient of energy consumption and population growth indicates an increase in energy consumption, and population growth will increase CO_2 emissions.

5. Discussion and concluding remarks

EU countries participate in global trade and international agreements, necessitating policy measures considering the carbon leakage effect for GHG emission mitigation. Various laws and policies targeting sectors, such as land, energy, transport, and agriculture, are implemented at the national and EU levels (e.g., European Union's Nitrates Directive) [67]. Research by Simionescu et al. [68] showed that between 2002 and 2019, facilitating environmentally friendly technologies and renewable energy sources by laws, regulatory means, and corruption control ensured reduced GHG emissions. The provision of credit to the private sector in eastern and central EU countries did not reduce GHG emissions because of mismatched grant prioritization. Previously, Lapinskienė et al. [69] revealed that higher taxes on energy production, research, and development reduced GHG emissions in EU countries between 2006 and 2013. However, the same report indicates that

the approximation of research and development impacts on GHG emissions can be problematic because of variations in the timing of conducting and implementing research findings by individual countries. Bellassen et al. [70] showed that inaccuracy and imprecise reporting of soil carbon changes threaten overall GHG emissions information. The main issues are the weak regulatory incentives to boost soil carbon monitoring and soil data is expensive and scarce. The success of land-related climate policy is affected because most soil carbon stock fluxes in European soils are not tracked, making this a significant blind spot. This implies that an additional 70 Mt $\rm CO_2~yr^{-1}$ emissions are omitted from croplands, although new removals of equal magnitude could appear in grasslands and forests [70].

We revealed a significant decline in CO_2 emissions in the EU-27 between 1990 and 2019 (Figs. 2 and 3). Also, the output of this research indicates the existence of an EKC (inverted-U shaped) relation between GDP and CO_2 emissions in the EU-27. Even though our research was conducted within 27 EU countries and POP was involved as a controlled variable, our finding aligns with previous studies, as highlighted in Table 7.

Structural changes in the economy, the use of renewable sources of energy, less carbon-intensive fuels, and improvements in energy efficiency accounted for the reductions [68,71]. Energy production and consumption are essential for economic development; therefore, they remain the key drivers of anthropogenic GHG emissions [72]. The relationship between GDP, energy, and POP variables can be explained by the "energy-GDP-population nexus," which refers to the interdependence between these three factors. GDP positively correlates with energy consumption, as increased economic activity requires more energy [73]. Population growth also contributes to increased energy consumption, which means greater demand for energy services [74]. The population and GDP relationship is typically positive [75] where a larger population generally means a more significant labor force, which can produce more goods and services, leading to increased economic output. However, the relationship between these variables and CO₂ emissions is complex and interdependent.

Energy consumption is one of the main drivers of CO_2 emissions [76], particularly from burning fossil fuels. As nations' GDP increases, so does their energy demand, producing higher CO_2

Table 6Long-run estimation of panel cointegration relation.

Regressors	FMOLS		DOLS		AMG	AMG	
	Coefficient	Probability	Coefficient	Probability	Coefficient	Probability	
LGDP	0.91*	0.00	0.96*	0.000	0.705**	0.000	
LGDP ²	-0.11*	0.00	-0.11*	0.000	-0.062**	0.000	
LEC	1.17*	0.00	1.13*	0.000	0.748*	0.000	
LPOP	-0.11*	0.00	-0.12**	0.007	1.305**	0.000	

^{*}and ** indicate significance at 0.01 and 0.05, respectively.

Table 7An overview of the previous research related to the EKC analysis within Europe.

Variable	Time period	E.U. countries	Input variables	EKC relationship	References
Ecological footprint, CO ₂ emissions	14 European countries	1990 -2014	Fossil fuels consumption, CO ₂ emissions, GDP, GDP ² , RE, Ecological footprint	No	[90]
CO ₂ emissions	27 European countries	1990 -2017	GDP, GDP ² , energy intensity, CO ₂ emissions	Yes	[4]
CO ₂ emissions	34 European countries	1990 -2021	RE, Nuclear energy, Research and Development in E.U., GDP, GDP ² , CO ₂ emissions, energy intensity	No	[91]
CO ₂ emissions	Five European countries	1990 -2015	GDP, foreign direct investment, energy innovation, air transport	Yes	[92]
GHG emissions (CO ₂ , CH ₄ , N ₂ O)	Hungary	1985 -2018	GHG emissions, EC, GDP	Yes	[3]
CO ₂ emissions	Visegrad countries	1990 -2018	RE, non-RE, POP, foreign direct investment, GDP, CO ₂ emissions.	Yes	[93]
CO ₂ emissions	Portugal	1970 -2016	Trade intensity, EC, GDP, CO_2 emissions	Yes	[94]
CO ₂ emissions	BRICS Countries	1990 -2015	GDP, GDP ² , economic complexity index, RE, CO ₂ emissions	Yes	[95]
Nitrous oxide (N ₂ O) emissions	Germany	1970 -2012	Nitrous oxide (N_2O) emissions, GDP, GDP ² , agricultural land used and exports	Yes	[96]
Ecological footprint, CO ₂ emissions	France	1977 -2017	Nuclear and RE, GDP, Ecological footprint, CO ₂ emissions	No	[97]

emissions [77]. In this context, the relationship is not necessarily linear due to many factors, such as the adaptation of renewable energy or implementation of energy-efficient practices in different sectors, which directly decoupled GDP and CO₂ emissions. On the other hand, population and CO₂ emissions have a complex interaction. However, population growth can drive innovation and technological advancements that reduce emissions.

Additionally, changes in individual behaviors and consumption patterns can significantly impact emissions, regardless of population size. However, reducing CO₂ emissions requires addressing the interconnections between energy, GDP, and population. This can involve promoting sustainable and efficient energy practices, investing in clean energy technology, and making behavioral changes that reduce consumption and waste. The European Environment Agency (EEA) [71] revealed that it is difficult to decouple emissions from economic growth in countries whose GDP is largely based on fossil fuels. However, according to the European neutrality law, it is possible to decouple economic growth from GHG emissions by having specific inclusive sector dialogs and partnerships between stakeholders to reduce the EU's greenhouse gas emissions in a cost-effective manner, such as the use of renewable energy and enhancing removals by sinks, to achieve the climate neutrality objective by 2050. To achieve this, all citizens' and consumers' empowerment in making change possible and providing accurate information to the public is pivotal. This aligns with policy actions related to information and voluntary mitigation strategies. For example, climate change, values, rights, the rule of law, and security conferences on a European platform represent an effective information-sharing framework [68].

To achieve the GHG 2050 neutrality objective, further research on energy, technology, economic cost-benefit analysis, and proper economic and regulatory frameworks will foster the development and diffusion of technological innovations. EEA [78] reported that the reduction in GHG emissions mainly occurred among electricity, heat, and combined heat and power producers. The increased use of renewable energy sources in power generation reduced emissions by 50 million tonnes in 2018 compared to 2017. According to Simionescu et al. [68], the achievement of the 27% use of renewable energy sources as per the EU's Renewable Energy Directive by 2030 will depend on each country's level of economic development, structure of energy consumption, and policies for achieving energy efficiency. Similarly, Mielcarek-Bocheńska and Rzeźnik [79] show

that technological changes in EU countries, the size of countries, and the proportion of changes in GHG emitting sectors also contribute to variations in GHG emissions. Germany's low emissions are attributed to innovative technology, research, and renewable energy sources. However, it is unclear that although Germany has the highest GHG emission reduction, it registered the highest drop in energy efficiency. Unsurprisingly, countries such as Latvia had the lowest emission reduction. Kar [80] shows that economic growth had an insignificant impact in mitigating CO2 emissions in Baltic countries, including Latvia. Surprisingly, these countries had not achieved the required GHG reduction limit by 2020, yet the current updated target by Fit for 55 initiatives by 2030 is much higher. This means these countries must take robust measures in emission trading and non-emission trading system sectors. Coordinated efforts, policies, and instruments benefit from enhanced synergism. In fact, GHG emission reduction requires changes in legislation, financial outlays, organization, and technical changes [79], which should not be taken in isolation but in a nexus with national and international laws because of the carbon leakage effect. However, generally, the results show that implementing GHG emissions reduction policies, such as technology and research, incentives, and economic adjustments, in countries such as Germany, France, Denmark, and the Czech Republic contribute to reducing GHG emissions. However, the extent of the impact is debatable.

Conversely, Petrović et al. [81] show that increased service sector participation in highly developed countries, such as France, Germany, Belgium, and the Netherlands, contributes to reduced transport GHG emissions. At the same time, the reverse is true for transition countries, such as Malta, Latvia, Hungary, Slovakia, and Estonia. This implies that environmental and climate benefits policies are being achieved by EU countries, although at varying levels [82] except for countries with a positive relation. Lee and van de Meene [82] define the environmental benefits of climate change as policies to decrease GHG emissions, positively impact energy consumption, enhance green space, or reduce waste generation. Similarly, Ahmad et al. [83] stress the importance of coupling innovation, green strategies, financial approaches, and environmental goals.

According to Simionescu et al. [68], the policy of providing domestic credit to the private sector reduced GHG emissions and pollution in Eastern and Central European countries from 2002 to

2019, but the specification of the legislative framework conditioning credit granting is suggested as a measure to ensure the proper execution of environmentally friendly technologies. Generally, the decreased emissions and high GDP in some EU countries imply that implementing GHG emission policies and strategies does not conflict with a growing economy [78]. According to our analysis, economic development in the EU has reached the stage where economic growth supports sustainability, as the EKC explains (Table 6). These results align with Dogan and Inglesi-Lotz [84] and Neagu [37]. Jóźwik et al. [85] confirmed the EKC hypothesis only in Poland out of all Central European countries, which implies that in the studied period, EU countries have progressed significantly toward the reduction of GHGs, suggesting that if a similar trend follows, the set target of 2030 and 2050 will be achievable. The positive sign of the coefficients of energy consumption and population growth in the FMOLS model indicates an increase in GHG emissions, in line with the findings of Dogan et al. [86].

Overall, implementing policies targeting energy production and consumption, population dynamics and structure, land, and agriculture of each country are critical to achieving EU countries' 2030 and 2050 GH G reduction targets. The results of this study show that the economic development exhibited by increased GDP reduces GHG emissions and consequently supports environmental protection. Although our results do not provide a detailed analysis of all GHG contributors per country, which may be necessary for the contextualized assessment of individual countries' progress in achieving GHG reduction set targets, our results provide a general overview of the GHG emission trend at the EU level, a perspective that has been lacking. Consequently, our results affect several national, regional, and international stakeholders. A detailed study analyzing the contrast between the reduction in GHG emissions, and the reduced energy efficiency needs to be conducted for Germany and other individual countries to assess trends of GHG emissions, energy efficiency, economic performance, and the effectiveness of specialized national environmental policies. Overall, this study provides valuable insights for decision-makers on the European Commission, offering a robust evaluation of the direct impact of all policies on environmental deterioration between 1990 and 2019.

CRediT authorship contribution statement

Safwan Mohammed: Conceptualization, Formal Analysis, Visualization, Writing - Original Draft, Writing - Review & Editing. Abid Rashid Gill: Formal Analysis, Writing - Original Draft. Main Al-Dalahmeh: Writing - Original Draft. Ali Alkerdi: Writing - Original Draft. Akasairi Ocwa: Writing - Original Draft. Karam Alsafadi: Visualization. Kaushik Ghosal: Writing - Review & Editing. Szilárd Szabó: Writing - Review & Editing. Judit OlÁh: Writing - Review & Editing. Endre Harsanyi: Writing - Review & Editing, Funding Acquisition.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary data

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References

- K.R. Abbasi, K. Lv, M. Radulescu, P.A. Shaikh, Economic complexity, tourism, energy prices, and environmental degradation in the top economic complexity countries: fresh panel evidence, Environ. Sci. Pollut. Control Ser. 28 (48) (2021) 68717–68731.
- [2] S. Mohammed, K. Alsafadi, I. Takács, E. Harsányi, Contemporary changes of greenhouse gases emission from the agricultural sector in the EU-27, Geol. Ecol. Landscapes 4 (4) (2020) 282–287.
- [3] S. Mohammed, A.R. Gill, K. Alsafadi, O. Hijazi, K.K. Yadav, M.A. Hasan, A.H. Khan, S. Islam, M.M.S. Cabral-Pinto, E. Harsanyi, An overview of greenhouse gases emissions in Hungary, J. Clean. Prod. 314 (2021) 127865.
- [4] F.V. Bekun, A.A. Alola, B.A. Gyamfi, S.S. Yaw, The relevance of EKC hypothesis in energy intensity real-output trade-off for sustainable environment in EU-27, Environ. Sci. Pollut. Control Ser. 28 (37) (2021) 51137–51148.
- [5] Y. Cai, C.Y. Sam, T. Chang, Nexus between clean energy consumption, economic growth and CO2 emissions, J. Clean. Prod. 182 (2018) 1001–1011.
- [6] F. Fatai Adedoyin, P.O. Agboola, I. Ozturk, F.V. Bekun, M.O. Agboola, Environmental consequences of economic complexities in the EU amidst a booming tourism industry: accounting for the role of brexit and other crisis events, J. Clean. Prod. 305 (2021) 127117.
- [7] A.M. Yusuf, A.B. Abubakar, S.O. Mamman, Relationship between greenhouse gas emission, energy consumption, and economic growth: evidence from some selected oil-producing African countries, Environ. Sci. Pollut. Control Ser. 27 (13) (2020) 15815–15823.
- [8] A.E. Caglar, M.W. Zafar, F.V. Bekun, M. Mert, Determinants of CO2 emissions in the BRICS economies: the role of partnerships investment in energy and economic complexity, Sustain. Energy Technol. Assessments 51 (2022) 101907.
- [9] M. Javid, F. Sharif, Environmental Kuznets curve and financial development in Pakistan, Renew. Sustain, Energy Rev. 54 (2016) 406–414.
- [10] S. Özokcu, Ö. Özdemir, Economic growth, energy, and environmental Kuznets curve, Renew. Sustain. Energy Rev. 72 (2017) 639–647.
- [11] S. Kumar, H. Fujii, S. Managi, Substitute or complement? Assessing renewable and nonrenewable energy in OECD countries, Appl. Econ. 47 (14) (2015) 1438–1459.
- [12] W. Rickels, A. Proelß, O. Geden, J. Burhenne, M. Fridahl, Integrating carbon dioxide removal into European emissions trading. Front. Clim. 3 (2021).
- [13] V. Masson-Delmotte, P. Zhai, H.-O. Pörtner, D. Roberts, J. Skea, P.R. Shukla, A. Pirani, W. Moufouma-Okia, C. Péan, R. Pidcock, Global warming of 1.5 C, An IPCC Special Rep. impacts of global warming of 1 (5) (2018) 43–50.
- [14] D. Balsalobre-Lorente, L. Ibáñez-Luzón, M. Usman, M. Shahbaz, The environmental Kuznets curve, based on the economic complexity, and the pollution haven hypothesis in PIIGS countries, Renew. Energy 185 (2022) 1441–1455.
- [15] I. Ozturk, A. Acaravci, CO2 emissions, energy consumption and economic growth in Turkey, Renew. Sustain. Energy Rev. 14 (9) (2010) 3220–3225.
- [16] F.V. Bekun, Mitigating emissions in India: accounting for the role of real income, renewable energy consumption and investment in energy, Int. J. Energy Econ. Pol. 12 (1) (2022) 188–192.
- [17] Q. Wang, T. Yang, R. Li, Does income inequality reshape the environmental Kuznets curve (EKC) hypothesis? A nonlinear panel data analysis, Environ. Res. 216 (2023) 114575.
- [18] Q. Wang, F. Zhang, R. Li, Revisiting the environmental kuznets curve hypothesis in 208 counties: the roles of trade openness, human capital, renewable energy and natural resource rent, Environ. Res. 216 (2023) 114637.
- [19] R. Li, Q. Wang, Y. Liu, R. Jiang, Per-capita carbon emissions in 147 countries: the effect of economic, energy, social, and trade structural changes, Sustain. Prod. Consum. 27 (2021) 1149–1164.
- [20] S. Kuznets, Economic growth and income inequality, Am. Econ. Rev. 45 (1) (1955) 1–28.
- [21] G.M. Grossman, A. Krueger, Economic growth and the environment, Q. J. Econ. 110 (2) (1995) 353–377.

- [22] N. Shafik, S. Bandyopadhyay, Economic growth and environmental quality: time series and cross-country evidence, in: Policy Research Working Paper Series 904, The World Bank, 1992.
- [23] M. Fodha, O. Zaghdoud, Economic growth and pollutant emissions in Tunisia: an empirical analysis of the environmental Kuznets curve, Energy Pol. 38 (2) (2010) 1150–1156.
- [24] M. Munasinghe, Is environmental degradation an inevitable consequence of economic growth: tunneling through the environmental Kuznets curve, Ecol. Econ. 29 (1) (1999) 89–109.
- [25] D.N. Le, Environmental degradation and economic growth in ASEAN-10: the perspective of EKC hypothesis, Malays. J. Econ. Stud. 56 (1) (2019) 43–62.
- [26] P. Yuelan, M.W. Akbar, M. Hafeez, M. Ahmad, Z. Zia, S. Ullah, The nexus of fiscal policy instruments and environmental degradation in China, Environ. Sci. Pollut. Control Ser. 26 (28) (2019) 28919—28932.
- [27] A. Haseeb, E. Xia, Danish, M.A. Baloch, K. Abbas, Financial development, globalization, and CO2 emission in the presence of EKC: evidence from BRICS countries, Environ. Sci. Pollut. Control Ser. 25 (31) (2018) 31283–31296.
- [28] L.H. Phong, Globalization, financial development, and environmental degradation in the presence of environmental kuznets curve: evidence from ASEAN-5 countries, Int. J. Energy Econ. Pol. 9 (2) (2019) 40–50.
- [29] M. Ben Jebli, S. Ben Youssef, I. Ozturk, Testing environmental Kuznets curve hypothesis: the role of renewable and non-renewable energy consumption and trade in OECD countries, Ecol. Indicat. 60 (2016) 824–831.
- [30] A. Sinha, M. Shahbaz, Estimation of environmental kuznets curve for CO2 emission: role of renewable energy generation in India, Renew. Energy 119 (2018) 703–711
- [31] H.P. Le, I. Ozturk, The impacts of globalization, financial development, government expenditures, and institutional quality on CO2 emissions in the presence of environmental Kuznets curve, Environ. Sci. Pollut. Control Ser. 27 (18) (2020) 22680—22697.
- [32] L.H. Phong, D.T.B. Van, H.H.G. Bao, The role of globalization on CO2 emission in vietnam incorporating industrialization, urbanization, GDP per capita and energy use, Int. J. Energy Econ. Pol. 8 (6) (2018) 275–283.
- [33] Q. Wang, X. Wang, R. Li, Does urbanization redefine the environmental Kuznets curve? An empirical analysis of 134 Countries, Sustain. Cities Soc. 76 (2022) 103382.
- [34] U. Al-mulali, C. Weng-Wai, L. Sheau-Ting, A.H. Mohammed, Investigating the environmental Kuznets curve (EKC) hypothesis by utilizing the ecological footprint as an indicator of environmental degradation, Ecol. Indicat. 48 (2015) 315–323.
- [35] A. Mohammed Saud M, P. Guo, I.u. Haq, G. Pan, A. Khan, Do government expenditure and financial development impede environmental degradation in Venezuela? PLoS One 14 (1) (2019) e0210255.
- [36] N. Van Eck, L. Waltman, Software survey: VOSviewer, a computer program for bibliometric mapping, Scientometrics 84 (2) (2010) 523–538.
- [37] O. Neagu, The link between economic complexity and carbon emissions in the European union countries: a model based on the environmental kuznets curve (EKC) approach, Sustainability 11 (17) (2019) 4753.
- [38] M. Ali, S.A. Raza, B. Khamis, Environmental degradation, economic growth, and energy innovation: evidence from European countries, Environ. Sci. Pollut. Control Ser. 27 (22) (2020) 28306–28315.
- [39] H.B. Mann, Nonparametric tests against trend, Econometrica 13 (3) (1945) 245–259.
- [40] P.K. Sen, Estimates of the regression coefficient based on kendall's tau, J. Am. Stat. Assoc. 63 (324) (1968) 1379–1389.
- [41] T.A. Buishand, Some methods for testing the homogeneity of rainfall records, J. Hydrol. 58 (1) (1982) 11–27.
- [42] M.I. Yuce, M. Esit, V. Kalaycioglu, Investigation of trends in extreme events: a case study of Ceyhan Basin, Turkey, J. Appl. Water Eng. Res. (2022) 1–16.
- [43] G. Yildirim, A. Rahman, Homogeneity and trend analysis of rainfall and droughts over Southeast Australia, Nat. Hazards 112 (2) (2022) 1657–1683.
- [44] N. Shafik, Economic Development and Environmental Quality: an Econometric Analysis, Oxford economic papers, 1994, pp. 757–773.
- [45] Grossman, Krueger, Economic environment and the economic growth, Q. J. Econ. 110 (2) (1995) 353–377.
- [46] A. Acaravci, I. Ozturk, On the relationship between energy consumption, CO2 emissions and economic growth in Europe, Energy 35 (12) (2010) 5412–5420.
- [47] N.M. Pham, T.L.D. Huynh, M.A. Nasir, Environmental consequences of population, affluence and technological progress for European countries: a Malthusian view, J. Environ. Manag. 260 (2020) 110143.
- [48] H. Pesaran, General diagnostic tests for cross-section dependence in panels, Fac Econ. (2004).
- Fac Econ. (2004).
 [49] K.S. Im, M.H. Pesaran, On the Panel Unit Root Tests Using Nonlinear Instrumental Variables, 2003.
- [50] A. Levin, C.-F. Lin, C.-S. James Chu, Unit root tests in panel data: asymptotic and finite-sample properties, J. Econom. 108 (1) (2002) 1–24.
- [51] G.S. Maddala, S. Wu, A comparative study of unit root tests with panel data and a new simple test, Oxf. Bull. Econ. Stat. 61 (S1) (1999) 631–652.
- [52] M.H. Pesaran, A simple panel unit root test in the presence of cross-section dependence, J. Appl. Econom. 22 (2) (2007) 265–312.
- [53] P. Pedroni, Critical values for cointegration tests in heterogeneous panels with multiple regressors, Oxf. Bull. Econ. Stat. 61 (S1) (1999) 653–670.
- [54] C. Kao, Spurious regression and residual-based tests for cointegration in panel data, J. Econom. 90 (1) (1999) 1–44.
- [55] J. Westerlund, Testing for error correction in panel data, Oxf. Bull. Econ. Stat.

- 69 (6) (2007) 709-748.
- [56] D. Persyn, J. Westerlund, Error-correction—based cointegration tests for panel data, STATA J. 8 (2) (2008) 232—241.
- [57] P. Pedroni, Purchasing power parity tests in cointegrated panels, Rev. Econ. Stat. 83 (4) (2001) 727–731.
- [58] M. Eberhardt, F. Teal, Productivity Analysis in Global Manufacturing Production. 2010.
- [59] S. Bond, M. Eberhardt, Accounting for Unobserved Heterogeneity in Panel Time Series Models, University of Oxford, 2013, pp. 1–11.
- [60] J. Coakley, A.-M. Fuertes, R. Smith, Unobserved heterogeneity in panel time series models. Comput. Stat. Data Anal. 50 (9) (2006) 2361–2380.
- [61] M.H. Pesaran, Time Series and Panel Data Econometrics, Oxford University Press, 2015.
- [62] B.H. Baltagi, The Oxford Handbook of Panel Data, Oxford Handbooks, 2015.
- [63] M.H. Pesaran, R. Smith, Estimating long-run relationships from dynamic heterogeneous panels, J. Econom. 68 (1) (1995) 79–113.
- [64] S. Johansen, Likelihood-based Inference in Cointegrated Vector Autoregressive Models, Oxford University Press on Demand, 1995.
- [65] J. Westerlund, Testing for error correction in panel data, Oxf. Bull. Econ. Stat. 69 (6) (2007) 709–748.
- [66] M.H. Pesaran, A simple panel unit root test in the presence of cross-section dependence, J. Appl. Econom. 22 (2) (2007) 265–312.
- [67] M.A. Harty, P.J. Forrestal, C.J. Watson, K.L. McGeough, R. Carolan, C. Elliot, D. Krol, R.J. Laughlin, K.G. Richards, G.J. Lanigan, Reducing nitrous oxide emissions by changing N fertiliser use from calcium ammonium nitrate (CAN) to urea based formulations. Sci. Total Environ. 563–564 (2016) 576–586.
- [68] M. Simionescu, W. Strielkowski, B. Gavurova, Could quality of governance influence pollution? Evidence from the revised Environmental Kuznets Curve in Central and Eastern European countries, Energy Rep. 8 (2022) 809–819.
- [69] G. Lapinskienė, K. Peleckis, Z. Nedelko, Testing environmental Kuznets curve hypothesis: the role of enterprise's sustainability and other factors on GHG in European countries, J. Bus. Econ. Manag. 18 (1) (2017) 54–67.
- [70] V. Bellassen, D. Angers, T. Kowalczewski, A. Olesen, Soil carbon is the blind spot of European national GHG inventories, Nat. Clim. Change 12 (4) (2022) 324–331
- [71] EEA, Analysis of key trends and drivers in greenhouse gas emissions in the EU between 1990 and 2015, in: E.E. Agency (Ed.), Luxembourg, 2017.
- [72] S.A. Sarkodie, Winners and losers of energy sustainability—global assessment of the sustainable development goals, Sci. Total Environ. 831 (2022) 154945.
- [73] R. Hannesson, Energy and GDP growth, Int. J. Energy Sect. Manag. 3 (2) (2009) 157–170.
- [74] D.I. Stern, The environmental Kuznets curve after 25 years, J. Bioecon. 19 (1) (2017) 7–28.
- [75] G. Kuno, Pradipto, Non-trivial relationship between scaling behavior and the spatial organization of GDP in Indonesian cities, PLoS One 17 (11) (2022) e0277433
- [76] X. Ma, C. Wang, B. Dong, G. Gu, R. Chen, Y. Li, H. Zou, W. Zhang, Q. Li, Carbon emissions from energy consumption in China: its measurement and driving factors, Sci. Total Environ. 648 (2019) 1411–1420.
- [77] N. Mughal, A. Arif, V. Jain, S. Chupradit, M.S. Shabbir, C.S. Ramos-Meza, R. Zhanbayev, The role of technological innovation in environmental pollution, energy consumption and sustainable economic growth: evidence from South Asian economies, Energy Strategy Rev. 39 (2022) 100745.
- [78] EEA, Trends and Drivers of EU Greenhouse Gas Emissions, European Environment Agency, 2020, pp. 1–36.
- [79] P. Mielcarek-Bocheńska, W. Rzeźnik, Greenhouse gas emissions from agriculture in EU countries—state and perspectives, Atmosphere 12 (11) (2021) 1396.
- [80] A.K. Kar, Environmental Kuznets curve for CO2 emissions in Baltic countries: an empirical investigation, Environ. Sci. Pollut. Control Ser. 29 (31) (2022) 47189–47208.
- [81] N. Petrović, N. Bojović, M. Petrović, V. Jovanović, A study of the environmental Kuznets curve for transport greenhouse gas emissions in the European Union, Facta Univ. – Ser. Mech. Eng. 18 (3) (2020) 513–524.
- [82] T. Lee, S. van de Meene, Comparative studies of urban climate co-benefits in Asian cities: an analysis of relationships between CO2 emissions and environmental indicators, J. Clean. Prod. 58 (2013) 15–24.
- [83] M. Ahmad, Z. Ahmed, Y. Bai, G. Qiao, J. Popp, J. Oláh, Financial inclusion, technological innovations, and environmental quality: analyzing the role of green openness, Front. Environ. Sci. 10 (2022).
- [84] E. Dogan, R. Inglesi-Lotz, The impact of economic structure to the environmental Kuznets curve (EKC) hypothesis: evidence from European countries, Environ. Sci. Pollut. Control Ser. 27 (11) (2020) 12717–12724.
- [85] B. Jóźwik, A.-V. Gavryshkiv, P. Kyophilavong, L.E. Gruszecki, Revisiting the environmental kuznets curve hypothesis: a case of central Europe, Energies 14 (12) (2021) 3415.
- [86] E. Dogan, R. Ulucak, E. Kocak, C. Isik, The use of ecological footprint in estimating the Environmental Kuznets Curve hypothesis for BRICST by considering cross-section dependence and heterogeneity, Sci. Total Environ. 723 (2020) 138063.
- [87] F.V. Bekun, B.A. Gyamfi, S.T. Onifade, M.O. Agboola, Beyond the environmental Kuznets Curve in E7 economies: accounting for the combined impacts of institutional quality and renewables, J. Clean. Prod. 314 (2021) 127924.
- [88] M.W. Zafar, F.M. Mirza, S.A.H. Zaidi, F. Hou, The nexus of renewable and nonrenewable energy consumption, trade openness, and CO2 emissions in the

- framework of EKC: evidence from emerging economies, Environ. Sci. Pollut. Control Ser. 26 (15) (2019) 15162–15173.
- [89] R.A. Badeeb, H.H. Lean, M. Shahbaz, Are too many natural resources to blame for the shape of the Environmental Kuznets Curve in resource-based economies? Resour. Pol. 68 (2020) 101694.
- [90] H. Altıntaş, Y. Kassouri, İs the environmental Kuznets Curve in Europe related to the per-capita ecological footprint or CO2 emissions? Ecol. Indicat. 113 (2020) 106187.
- [91] L.C. Voumik, M. Rahman, S. Akter, Investigating the EKC hypothesis with renewable energy, nuclear energy, and R&D for EU: fresh panel evidence, Heliyon 8 (12) (2022) e12447.
- [92] D. Balsalobre-Lorente, O.M. Driha, N.C. Leitão, M. Murshed, The carbon dioxide neutralizing effect of energy innovation on international tourism in EU-5 countries under the prism of the EKC hypothesis, J. Environ. Manag. 298 (2021) 113513.
- [93] N.C. Leitão, C.C. Dos Santos Parente, D. Balsalobre-Lorente, J.M. Cantos Cantos, Revisiting the effects of energy, population, foreign direct investment, and economic growth in Visegrad countries under the EKC scheme, Environ. Sci. Pollut. Control Ser. 30 (6) (2023) 15102–15114.
- [94] N.C. Leitão, Testing the role of trade on carbon dioxide emissions in Portugal, Economies 9 (1) (2021) 22.
- [95] N.C. Leitão, D. Balsalobre-Lorente, J.M. Cantos-Cantos, The impact of renewable energy and economic complexity on carbon emissions in BRICS countries under the EKC scheme. Energies 14 (16) (2021) 4908.
- [96] M.A. Zambrano-Monserrate, M. Fernandez, An environmental kuznets curve for N2O emissions in Germany: an ARDL approach, Nat. Resour. Forum 41 (2) (2017) 119–127.
- [97] U.K. Pata, A. Samour, Do renewable and nuclear energy enhance environmental quality in France? A new EKC approach with the load capacity factor, Prog. Nucl. Energy 149 (2022) 104249.