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A better strategy: using green GDP to measure economic health

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Introduction: Gross Domestic Product (GDP) is the most well-known and widely used measure of a country's economic health. However, GDP fails to account for the depletion of natural resources and the environmental damage that occurs in the pursuit of economic growth, leading to an incomplete and potentially misleading picture of a nation's well-being. To address this shortcoming, Green GDP (GGDP) is proposed as a more comprehensive indicator that incorporates environmental factors into the economic assessment. This study builds on extensive literature reviews, internationally accepted GGDP accounting methods, and scholarly research to propose a new GGDP calculation model that better reflects a country's sustainable development.

Methods: The proposed GGDP model is divided into two main components: natural resource loss and environmental pollution loss. Each component is further broken down into primary factors that are condensed into 13 sub-criteria reflecting a country's capacity for sustainable development. Principal Component Analysis (PCA) is utilized to identify the most representative factors from these sub-criteria and to analyze the relationships among GGDP, these factors, and global mean temperature. Additionally, the Integrated Environmental Sustainability Index (IESI) is used to develop a global temperature mitigation prediction model, which considers the impacts of epidemics, sea and land temperatures, and variations in climate across different regions.

Results: The analysis shows a 74% probability that positive GGDP growth correlates with temperature changes over a 50-year period, indicating that economic activities measured by GGDP are linked to climate change. The GGDP model reveals significant differences between global GDP and Green GDP, with the latter growing at a much slower rate. This slower growth of Green GDP is primarily due to the declining share of GDP from natural resource-dependent activities, which has fallen from 90% in the 1970s to 80% in 2020. This trend underscores the increasing gap between traditional economic growth and sustainable development, suggesting that as countries continue to rely on natural resources, their overall ecological efficiency declines, environmental pressures increase, and the potential for long-term sustainable development diminishes.

Discussion: The findings demonstrate that all factors within the GGDP model are proportional to global temperature, underscoring the significant impact that natural resource utilization and pollution emissions have on economic growth and climate change. The study further evaluates global sustainable development by considering both economic and environmental perspectives. Using Brazil as a case study, the model is applied to assess the values of each component within the GGDP framework, providing a comprehensive analysis of the country's sustainable development challenges and potential solutions. This approach

establishes a method for assessing sustainable development that can be adapted for use in other countries, offering a path forward for integrating environmental considerations into economic policies.

KEYWORDS

GGDP, global temperature prediction model, sustainable development, environmental pollution loss, resource consumption

1 Introduction

1.1 Background

Gross Domestic Product (GDP) is a well-known and widely used indicator of a country's economic wellbeing (Bazan, 1997; Nordhaus, 2006; World Bank, 2011). It determines a country's purchasing power and access to loans, incentivizing nations to propose policies and initiatives that boost their GDPs. The monetary value of final goods and services produced within a country's borders during a specific time period is used to calculate GDP. This approach, however, is flawed because it prioritizes current production over the need to conserve resources for the future. For example, a country with vast forest reserves could boost its GDP by felling trees and producing more wooden furniture, causing significant environmental damage (see Figure 1). Similarly, harvesting more fish to boost GDP could cause irreversible damage to fish stocks with no consequences (Lan et al., 2002).

Economic health has traditionally been measured by Gross Domestic Product (GDP), a metric that captures the market value of all final goods and services produced within a country in a given period (Smith, 1993; Krugman and Wells, 2009). However, GDP has been criticized for its inability to account for environmental degradation, social wellbeing, and the sustainability of economic growth (Costanza et al., 1997; Jackson, 2009). As a result, there is a growing consensus among economists and policymakers that a more comprehensive measure of economic performance is needed, one that incorporates environmental and social factors: the Green Gross Domestic Product (Green GDP) (Pearce and Turner, 1991; Hamilton and Clemens, 1999). Green GDP changes the standard GDP by integrating environmental prices and advantages, such as contamination reduction, source exhaustion, and modifications in ecological quality (Arrow et al., 1996; Dasgupta, 2013). This method acknowledges that economic activity is not separated from the natural environment and that lasting advancement calls for a balance in between financial development and environmental management (Rockström et al., 2009; Stiglitz et al., 2009). The idea of Green GDP has actually been getting grip recently, with a number of countries and international organizations exploring its execution (United Nations, 2013; World Bank, 2014). Studies have shown that incorporating ecological factors into financial measurement can lead to more informed policy decisions and a better understanding of the true costs and benefits of economic activities (Dawson and Graham, 2007; Field et al., 2014). The National Accounts - Analysis of Main Aggregates (2024) highlights the importance of integrating environmental factors into economic indicators to provide a more accurate representation of a country's financial health.

1.2 Literature review

1.2.1 Concept of green GDP

The traditional Gross Domestic Product (GDP), while a reliable measure of economic activity, has consistently stopped working to capture the adverse surfaces associated with ecological injury and resource usage (Hamilton and Clemens, 1999). The development of the Eco-friendly GDP idea addresses this gap by integrating ecological and social aspects into financial assessments, using a more alternative view of a nation's financial wellness (Costanza et al., 1997; Neumayer, 2003). This approach straightens with the broader global movement in the direction of lasting advancement, as supported by the Stiglitz Payment (Stiglitz et al., 2009) which highlights the relevance of taking into consideration environmental and social wellness in financial performance measurement.

The incorporation of all-natural capital and ecological services in financial evaluation is a critical component of the. Eco-friendly GDP structure (Daily, 1997). This not only gives a much more exact depiction of real worth of economic tasks but also urges policies that protect and maintain the environment for future generations (United Nations, 2013). The World Bank (2014) has actually resembled this belief, highlighting the demand for economic indicators that represent the ecological prices and advantages of advancement tasks.

1.2.2 Applications of Green GDP

The practical application of Environment-friendly GDP in policy-making is essential for steering economies toward sustainable development. As discussed by Li and Lin (2017), the incorporation of Eco-friendly GDP into policy decisions is crucial for fostering ecological sustainability without undermining economic growth. Lenaerts and Tagliapietra (2022) further emphasize that green GDP plays a vital role in shaping policies that promote green economic practices, including the adoption of eco-friendly financial policies and the development of green financial instruments. These measures are pivotal in driving sustainable economic strategies that align with both environmental and economic goals.

On the global stage, Green GDP functions as an important tool for comparing ecological performance and encouraging collective initiatives to attend to global ecological challenges (United Nations, 2013). The World Bank (2014) has played a considerable function in promoting Eco-friendly GDP as a system for assessing the environmental implications of economic activities across different countries, thereby influencing international environmental administration and the international fostering of lasting methods.

1.2.3 Quantification of green GDP

Accurate quantification of Environment-friendly GDP necessitates the growth of advanced methods with the ability of valuing natural resources and eco-friendly solutions, and accountability for contamination and resource deficiency costs (Dasgupta, 2013). The United Nations' System of Environmental-Economic Audit (SEEA) represents a major innovation, providing a standard strategy to incorporating ecological metrics right into economic audit (United Nations, 2013). This structure is essential to the calculation of Eco-friendly GDP and its reliable utilization in policy formulation and decision-making processes.

The introduction of sophisticated computational designs, such as neural networks, has actually additionally boosted the metrology of Green GDP. Zhu et al. (2023) have shown the potential of these designs in forecasting environmental impacts and improving the precision of Environment-friendly GDP calculations. Such modeling techniques are essential for getting insights into the lasting ecological effects of economic tasks and for developing sustainable economic policies that secure the earth's ecological communities.

2 Global temperature impact prediction model based on GGDP

Today, numerous scholars on the planet have actually recommended an approach to determine GGDP. Therefore, by comparing the benefits and disadvantages and performance of different well-known methods, this paper initially selects a major index that can change GDP to determine the economic health of a country and combines the existing data to put forward the last accounting approach of GGDP (Deng, 2002).

2.1 Method comparison

Currently known methods are.

2.1.1 Accounting for resource consumption

An example of a model that accounts for resource consumption in green GDP calculations is the "Adjusted Net Savings" (ANS) indicator developed by the World Bank (Nordhaus, 2006).

GDP is gross savings, PCD is depreciation of physical capital and NRD is depletion of natural resources.

2.1.2 Ecological footprint model

An example of a model that uses the ecological footprint method to estimate green GDP is the "Ecological Footprint" developed by the Global Footprint Network (World Bank, 2011) is the total land area, average productivity and is biocapacity. Biocapacity is a measure of an ecosystem's ability to produce resources and absorb waste (Bazan, 1997).

2.1.3 Inclusive Wealth Index IWI

The Inclusive Wealth index considers not only traditional economic indicators such as GDP, but also the country's natural, human and productive capital. Natural capital includes the country's natural resources, such as forests, minerals and



water. Human capital refers to the skills and knowledge of the population, while productive capital includes the infrastructure and equipment used for production. IWI is calculated as follows: represent production capital, human capital and natural capital respectively. P is the social (shadow) price of all kinds of capital.

2.2 Establishment of sustainable development evaluation method based on energy analysis

Energy sustainable indices (ESI) reflect the sustainable development capability of ecological economic system. Only on the premise of improving unit social and economic benefits and controlling unit comprehensive environmental pressure can the level of sustainable development be improved. The higher the ESI value is, the better the sustainability is. Referring to previous studies (Xu, 2012), only when ESI value is lower than 1 can the socio-economic development be good, and the state of eco-economic system be sustainable. When ESI value is higher than 10, the socio-economic development is poor and sustainable development cannot be guaranteed. The improvement of ESI is to fully consider the emission and waste in the calculation process of ESI (Lan et al., 2002). For the impact of waste on sustainable development, the ratio of waste to renewable energy value is added to make the evaluation results more consistent with the actual situation. The specific calculation formula is as follows (World Bank, 2011): ImprovedESI.

Where IESI is an energy sustainable index; EYR is energy yield; ELR is the environmental load rate; EWR is the ratio of waste to renewable energy value.

2.3 Model building

By comparing the advantages and disadvantages of the three methods, and combining with the international common green

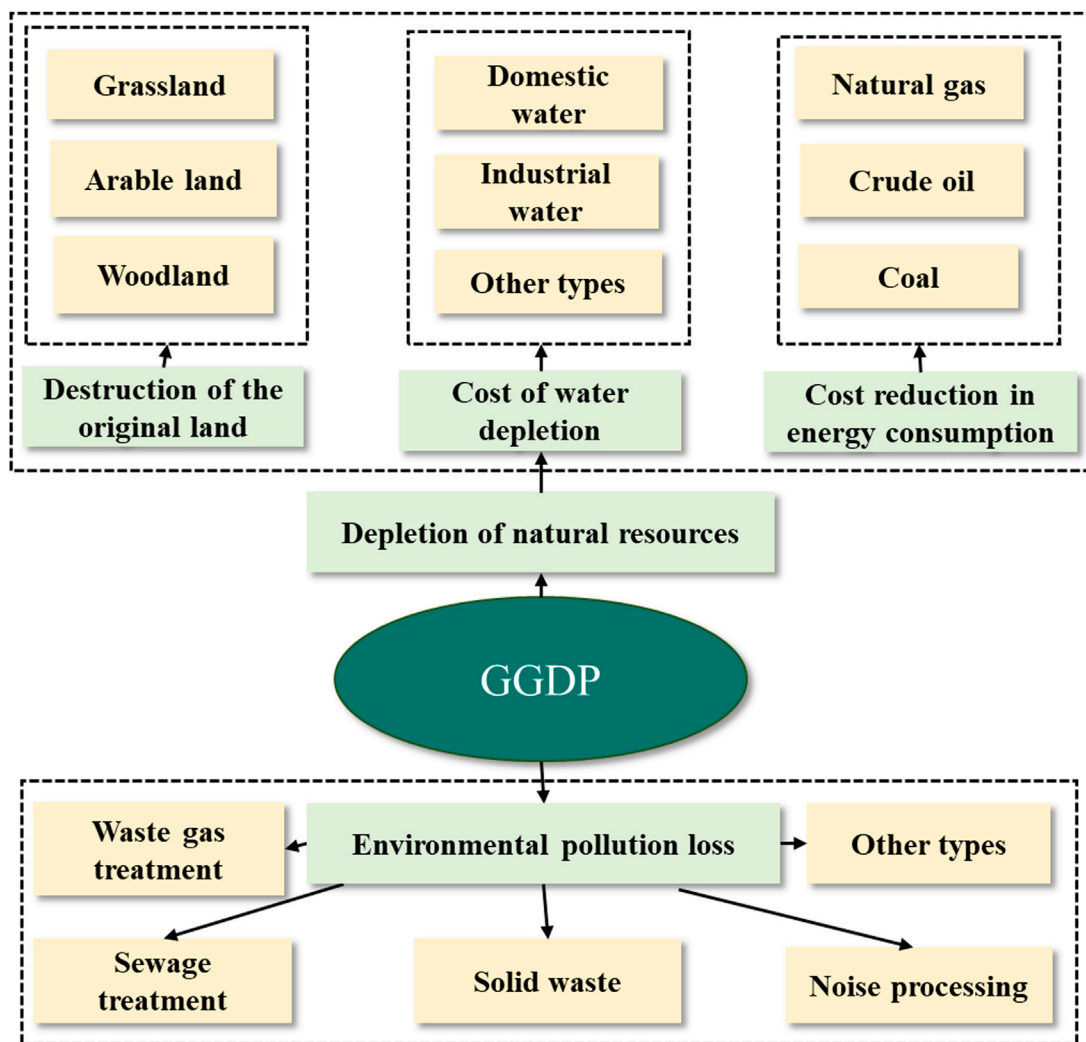


FIGURE 2
Land use change from primary forest.

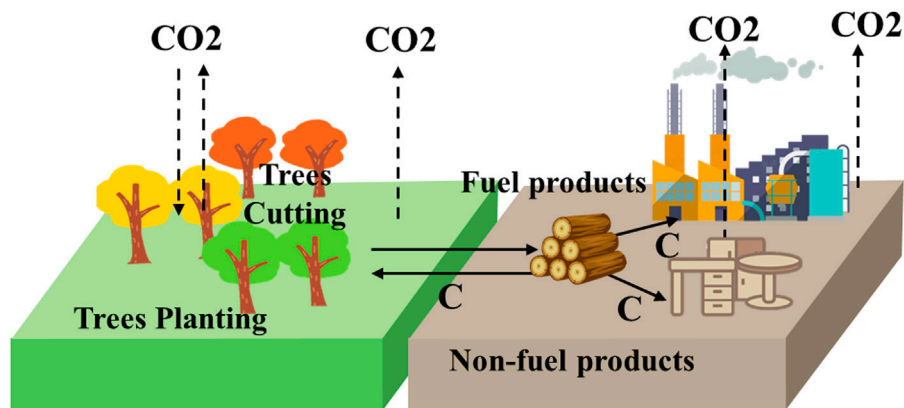


FIGURE 3
Map of carbon cycle in forest systems (Forests and Forest Products).

GDP accounting method, we summarized the following equation: $R + E$ is the cost of environmental damage. R is the loss of natural resources; E is the loss of environmental pollution. After consulting relevant literature and materials, we found several representative factors to further refine the two. Among them, resource loss R includes mineral energy loss, land resource loss, water resource loss, etc., while environmental pollution loss E mainly refers to environmental pollution caused by economic development, including wastewater, waste gas, solid waste, noise pollution, etc. The details are shown in the following [Figure 2](#).

In the second part A: The main consideration is the depletion of different land LD , water resources consumption WRC and energy consumption EC .

In the second part E: We consider the main components: sewage treatment ST , greenhouse gas damage, and solid waste damage SWD .

For CO_2 emissions, we still need to consider the presence of carbon dioxide uptake by forests, so we need to consider net CO_2 emissions that do not participate in the carbon cycle. The diagram [Figure 3](#) is shown below.

2.4 Data processing

2.4.1 The cost of destruction of the original land and the new value of restoration

This section consists of three parts: arable land, woodland and grassland. First of all, according to the international statistical yearbook respectively calculate the value of three types of table, and then determine the price of the unit area of cultivated land, woodland, grassland, take the average, and finally use the unit price of the resource multiplied by the area of the resource, that is, the value of the resource. Namely: land consumption cost = Σ changes in land surface prices average prices of such land in the same year ([World Bank, 2011](#)).

2.4.2 Calculation of the cost of water depletion

Water resources include domestic water and industrial water. The annual water consumption of each type is multiplied by the unit price of the same type of water in the same year, and the sum of the two is the cost of water resources consumption reduction. That is, water consumption reduction cost = Σ annual water consumption of different nature average price of water of that nature in that year.

2.4.3 Cost reduction in energy consumption

Firstly, three indices—coal, crude oil, and natural gas—are selected. Energy costs are calculated by multiplying the annual consumption of each energy source by its respective unit price for that year. The total energy consumption cost is then determined by summing the average costs of these indices. Thus, the cost of energy consumed is calculated as the sum of the annual consumption of various energy sources multiplied by the average price of these resources for the year ([Kureski et al., 2020](#)).

2.4.4 Carbon dioxide carbon neutralization

In the realm of waste gas treatment, carbon dioxide is identified as the principal greenhouse gas due to its profound impact on global temperature and its capacity to serve as a representative indicator for

other atmospheric pollutants ([United Nations Statistics Division, 2007](#)). The financial cost associated with the treatment of CO_2 emissions is calculated by multiplying the annual quantity of CO_2 emissions by the carbon pricing metrics provided by the World Bank. The aggregate expenditure is thus derived from this calculation.

2.4.5 Sewage and solid waste calculation and treatment

The cost of treating sewage and solid waste is determined by multiplying the annual quantity of each type of waste by its corresponding unit treatment price. The total waste disposal cost is obtained by summing these individual costs. Therefore, the formula for waste disposal cost is: the sum of (annual amount of each type of waste average unit price for that type in that year).

3 GGDP global temperature mitigation forecast model based on IESI

This chapter constructs a GGDP global temperature mitigation forecast optimization model based on IESI, used to synthesize various natural resource-related factors and describe their correlation with global temperature and GGDP.

3.1 Correlation analysis

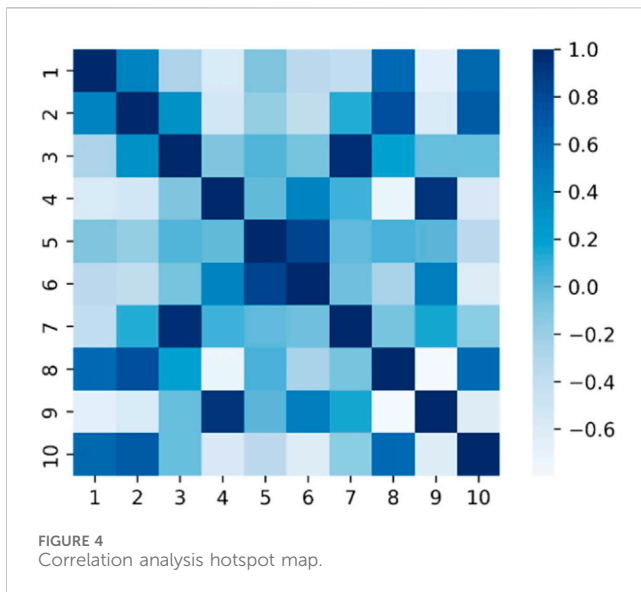
After determining the proportion of GDP attributed to each factor of resource loss (R) and environmental pollution (E), the correlations between these factors, GGDP, and global temperature were calculated. Spearman's rank correlation coefficient was used to estimate the correlation between variables X and Y . This method describes the correlation between variables through a monotone function. If the sets of values for the two variables do not contain duplicate elements, the correlation coefficient (ρ) between them can reach +1 or -1, indicating a perfect monotone relationship ([Wallace, 2000](#)). [Figure 4](#) below illustrates these correlations.

3.2 Principal factor selection

Using principal component analysis (PCA), 14 secondary factors were initially screened and reduced to four principal factors based on their distinct characteristics. [Table 1](#) presents the results of this screening process, showing the loadings of each factor and their contribution to the principal components. Additionally, [Figure 5](#) provides a visual representation of the variance explained by these principal factors, highlighting their significance in the overall analysis.

As shown in the table and figure above, there is no significant correlation among the four selected factors, indicating that these factors can be used as main variables for further analysis.

Additionally, a grey model was constructed to further validate the correlation and accuracy of these main factors. The results of this validation, including the predicted and observed values, are summarized in [Table 2](#). This table highlights the strong correlation and accuracy of the grey model in confirming the significance of the identified principal factors.



Based on the data presented in the table, the posterior difference ratio is calculated to be 0.117, which signifies a high degree of model accuracy. The analysis indicates that all stage ratios of the original sequence are within the range of 0.962–1.04, affirming the suitability of the original sequence for developing a grey prediction model (Deng, 2002). Additionally, the model’s average relative error of 0.99% reflects an excellent fit.

3.3 Numerical solution of model

In the initial phase, calculations for each factor were conducted based on the preliminary model. The parameters utilized in these calculations are elucidated in Table 3 below. This table provides a comprehensive overview of the key parameters, including their values and the assumptions made during the modeling process.

During the process of model refinement, it became evident that some variables had a negligible impact on GGDP. Therefore, factors contributing less than 2% were systematically excluded from the analysis. The four predominant factors retained include energy consumption, carbon dioxide emissions, water resource utilization and wastewater treatment (consolidated as a single factor), and the use of woodland and agricultural land. The refined model is presented in Figure 6.

As illustrated in the graph, among the four factors, total energy consumption exerts the most significant influence on GGDP, with carbon dioxide emissions following closely. Consequently, it is imperative to conduct further research to delve deeper into the relationship between these two factors and their effects on global temperature.

TABLE 1 Factor load coefficient table.

	Factor loading coefficient	Common degree
Energy	0.93	0.865
Forest	0.495	0.245
Water	0.179	0.032
CO ₂ Emissi	0.979	0.958

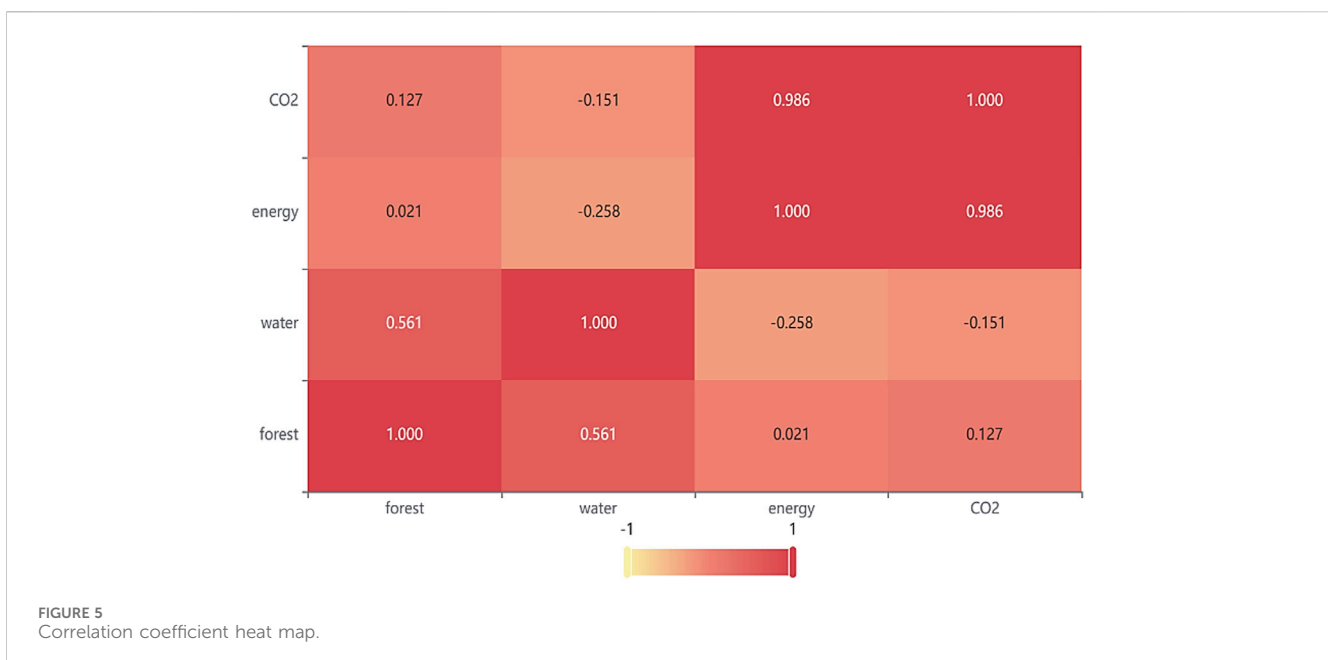
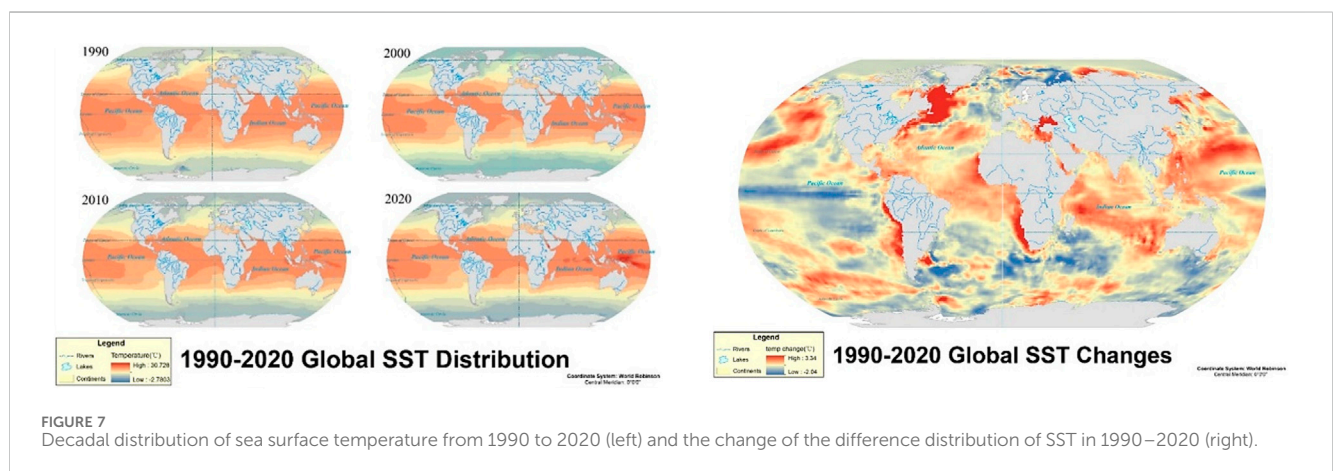
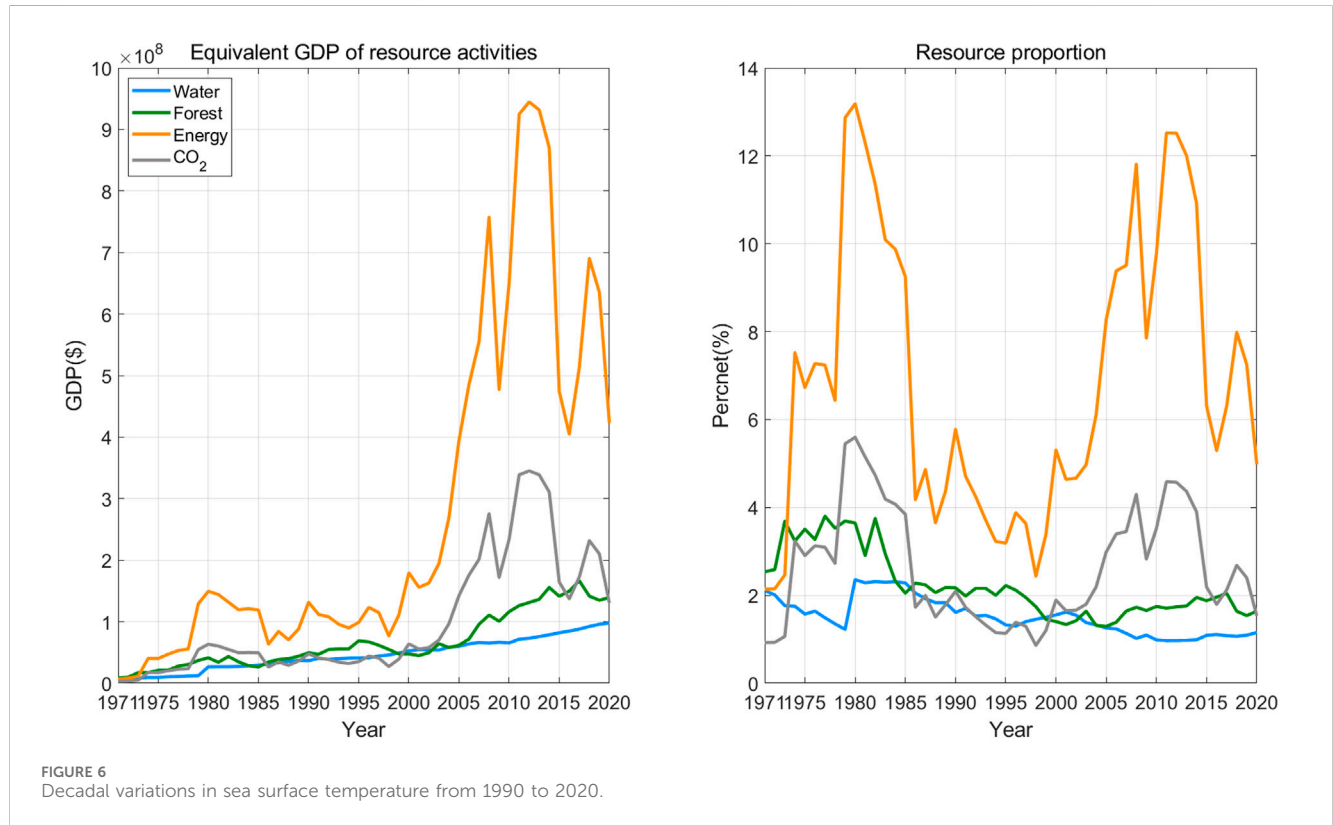


TABLE 2 Grey model construction.

Development coefficient	Gray action	Posteriori ratio value
-0.002	13.753	0.117

TABLE 3 Model parameters of Tien Shan.

Parameters	Country area	Land area	Agri land	GDP	Forest land
Value					
Parameters	Inland waters	Glo water	CO ₂ emissions	Oil prices	value added
Value					



3.4 Considerations of ocean temperature

Oceans cover about 71 percent of the Earth’s surface. In the global carbon cycle, oceans play a significant role in absorbing carbon dioxide compared to land, making them a better indicator

of CO₂’s impact on air temperature. Consequently, the effect of carbon dioxide on sea temperature is the primary focus. The global distribution and regional variation of sea surface temperature (SST) are illustrated in Figure 7 using historical monthly average data from January 1990 to December 2020, combined with statistical methods.

The two figures above illustrate the characteristics and variations in global SST from 1990 to 2020, further confirming the hypothesis. As a result, it is necessary to separately analyze the regression averages of carbon dioxide's impact on sea temperature and land temperature.

3.5 Results presentation and description

The analysis reveals significant relationships between Green GDP (GGDP), Gross Domestic Product (GDP), and global temperature changes. Figure 8 illustrates these relationships, showing how variations in GDP and GGDP correspond with changes in global temperature. This diagram highlights the critical interactions between economic growth and environmental impact, serving as a visual foundation for understanding the broader implications of the study's findings.

As depicted in the graph, there is a discernible upward trend in global average temperatures, while both GDP and GGDP exhibit slight declines. This analysis suggests that an increase in global average temperature is associated with a reduction in GDP and GGDP. Specifically, a 4-degree Celsius rise in average temperature is projected to lead to a 1 percentage point decrease in GDP and a 1.4 percentage point decrease in GGDP. Figure 9 illustrates the interrelationship between carbon dioxide levels, energy consumption, and global temperature, reinforcing the connection between rising CO₂ emissions and global warming.

Extensive research on the macroeconomic impacts of temperature on human activities has identified several factors that contribute to productivity losses as temperatures rise. These factors include:

Human Health: Elevated temperatures can have detrimental effects on human health, resulting in a higher incidence of heart disease in developed nations such as the United States, as well as increased mortality rates from vector-borne diseases like dengue fever and malaria in developing countries.

Labor Productivity: Higher temperatures generally decrease labor productivity.

People are more prone to making mistakes and performing poorly in hot conditions.

Increased temperatures are likely to escalate human conflicts, thereby disrupting normal economic activities.

These findings substantiate the validity of the established model in predicting the global impact of climate change. Figure 10 provides a forecast of GGDP and global mean temperature based on the analysis, indicating the expected economic consequences of ongoing temperature increases. Consequently, GGDP emerges as a robust and comprehensive metric for assessing a nation's economic health.

1. Since 1970, global land temperatures have been steadily rising. Additionally, ocean temperatures are projected to increase over the next 20 years.
2. Over the years, there has been a consistent upward trend in energy consumption, carbon dioxide emissions, and global mean temperatures. This trend indicates a strong positive correlation between global mean temperature and the primary natural resource factors considered in GGDP.
3. Excessive carbon dioxide emissions have a profound impact on global climate change. Projections indicate that CO₂ emissions

will continue to rise annually, consistent with the trend of global warming and the increasing disparity between GGDP and GDP. Therefore, mitigating excessive CO₂ emissions is critical for addressing global warming.

4. The increase in CO₂ concentration leads to significant changes in global temperatures across various regions. Additionally, alterations in precipitation patterns are expected to cause approximately a 10% fluctuation in water resources in different regions. Notably, the absolute value of the correlation coefficient between energy consumption and global mean temperature is higher and more significant.

These findings validate the established model's efficacy in predicting the global impact of climate mitigation efforts. Consequently, GGDP can be effectively utilized as a primary indicator of a country's economic health.

4 Robustness evaluation of GGDP model

4.1 Analysis and results of global carbon GGDP accounting

To comprehensively understand and analyze the interrelationship between various indicators within GGDP and traditional GDP, this study calculates the ratio of each indicator to GDP. Furthermore, it assesses the average annual differential between GDP and GGDP from 1990 to 2010. The comparative analysis is illustrated in Figure 11.

Global sustainable development is assessed through the dual lenses of economic performance and environmental impact. An evaluation framework grounded in energy theory has been developed to gauge global sustainability. The findings from this system reveal the following insights.

4.1.1 Green GDP and total GDP growth

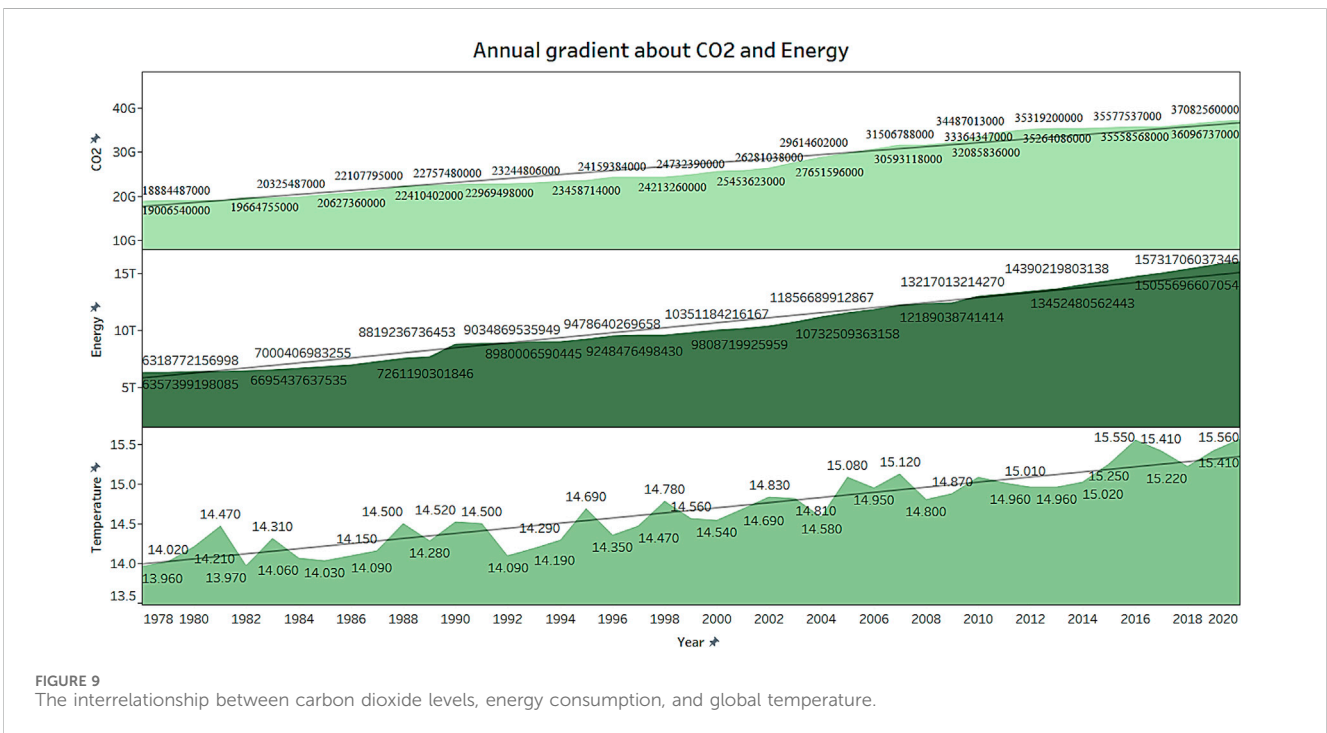
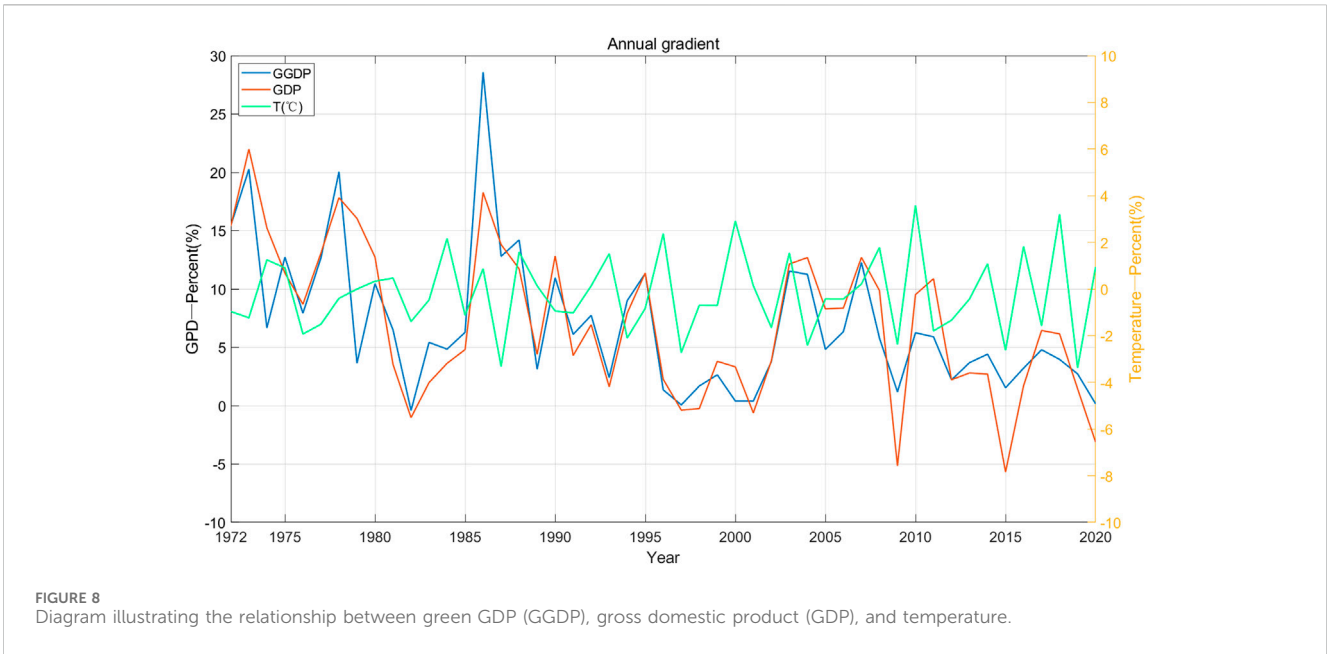
Both Green GDP and total GDP have exhibited continuous growth over the observed period. However, the growth rate of Green GDP is markedly slower compared to that of total GDP. This discrepancy implies that an economic development model heavily dependent on natural resource consumption exacerbates the divergence between GDP and Green GDP, thereby impeding sustainable development.

4.1.2 Natural resource consumption and pollution

The annual consumption and pollution of natural resources have shown a persistent upward trend, while overall ecological efficiency has progressively declined. The rapid pace of economic development has resulted in increased energy consumption, urban expansion, heightened ecological pressure, and a reduced capacity for sustainable development.

4.2 Comparison between green GDP (GGDP) and traditional GDP

1. Enhanced Comprehensiveness: Green GDP incorporates environmental considerations and resource utilization,

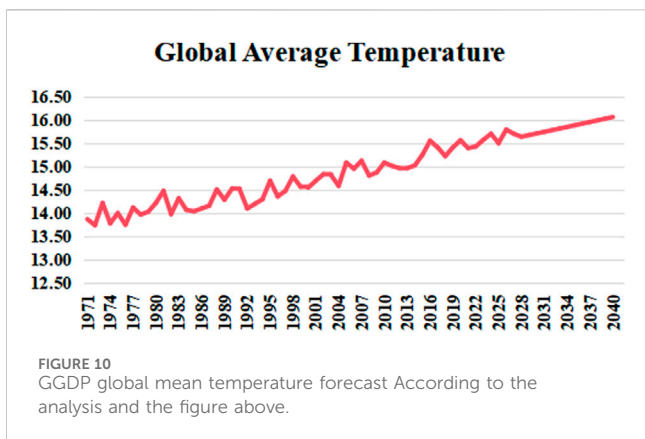


offering a more holistic view of the environmental impacts of economic development.

2. Increased Sustainability: By elucidating the interconnections between economic growth and the natural environment, Green GDP facilitates the alignment of economic progress with environmental stewardship, thereby advancing sustainable development.
3. Greater Accuracy: Green GDP provides a more precise measurement of the interplay between economic activities

and environmental resource consumption, offering a robust foundation for informed policy-making.

4. Green GDP elucidates the specific environmental repercussions of economic development, thereby aiding in the cultivation of public awareness and fostering a greater commitment to environmental sustainability. The following framework, as illustrated in Figure 12, provides a more complete and accurate representation of GGDP, capturing its multifaceted impact on both the economy and the environment.



- The predominance of the tertiary sector as the main driver of employment growth, comprising 74.2% of non-agricultural employment in the same year.

The current GDP calculations for Brazil do not incorporate the consumption of natural resources or the environmental degradation costs, potentially leading to adverse long-term impacts on the country’s economic stability and health.

Considering Brazil’s heavy dependence on natural resources, this section examines the effects of variations in CO₂ concentration, temperature, humidity, and other environmental factors on Brazil’s GGDP and GDP over the past 5 decades. It also discusses how these environmental changes might influence the economy and highlights specific observable modifications.

However, our research also revealed that Green GDP has some disappointing features.

5 An analysis of green GDP (GGDP) in Brazil

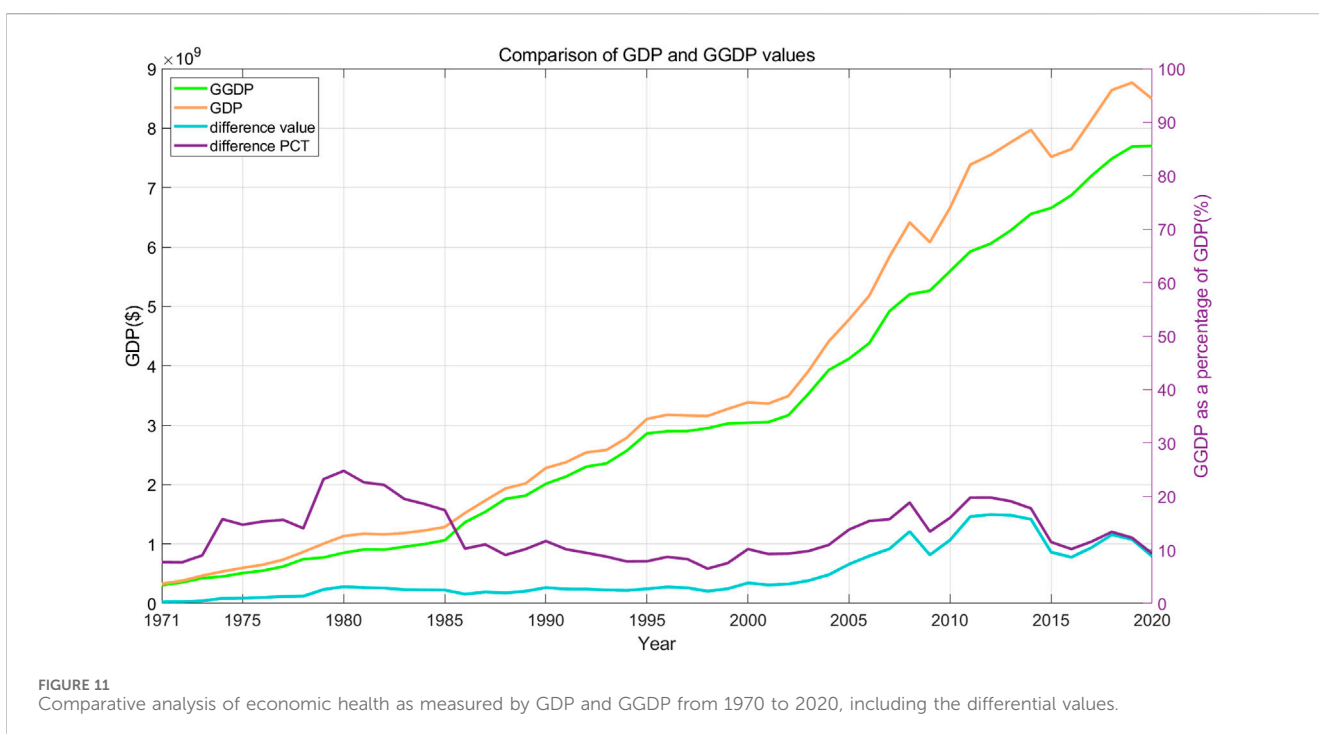
At present, Brazil’s economic growth and development are significantly reliant on its natural resources. The trends in employment growth within Brazil reveal two notable characteristics:

- A swift increase in trade-related agricultural employment, which is classified as non-agricultural employment, representing 10.2% of the total employment in 2011 (Kureski et al., 2020).

5.1 Development of the GGDP accounting model for Brazil

The development of the GGDP accounting model for Brazil involves a detailed comparison between traditional GDP and Green GDP (GGDP) metrics. Figure 13 illustrates the comparison of GDP and GGDP for Brazil, highlighting the differences in economic evaluation when environmental factors are taken into account. This analysis provides insights into the impact of environmental resource utilization on Brazil’s economic performance and demonstrates the added value of incorporating GGDP into national accounting.

The graph provides a comparative analysis of GGDP and GDP. The blue line represents GGDP, the orange line signifies GDP, the yellow line indicates the difference between the two, and the purple line illustrates the ratio of GGDP to GDP. The declining proportion



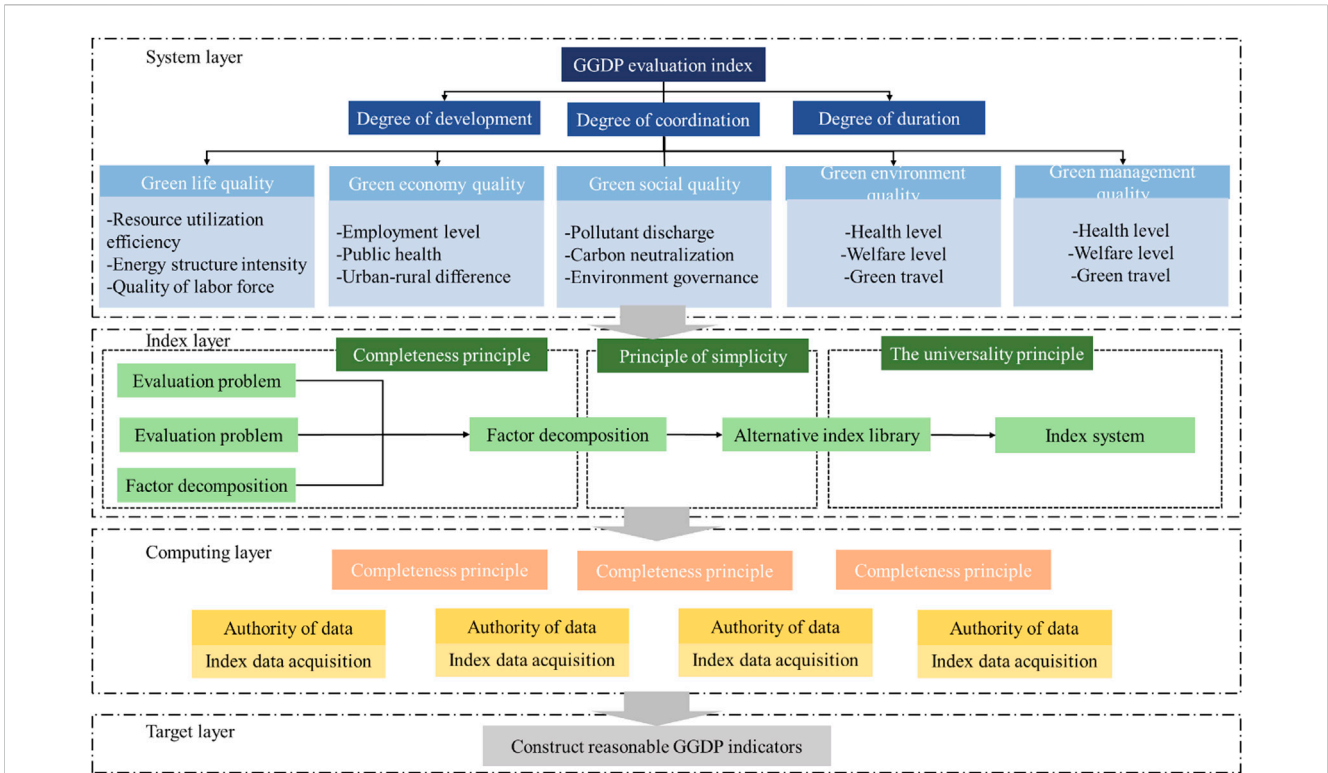


FIGURE 12 The structure of green GDP (GGDP).

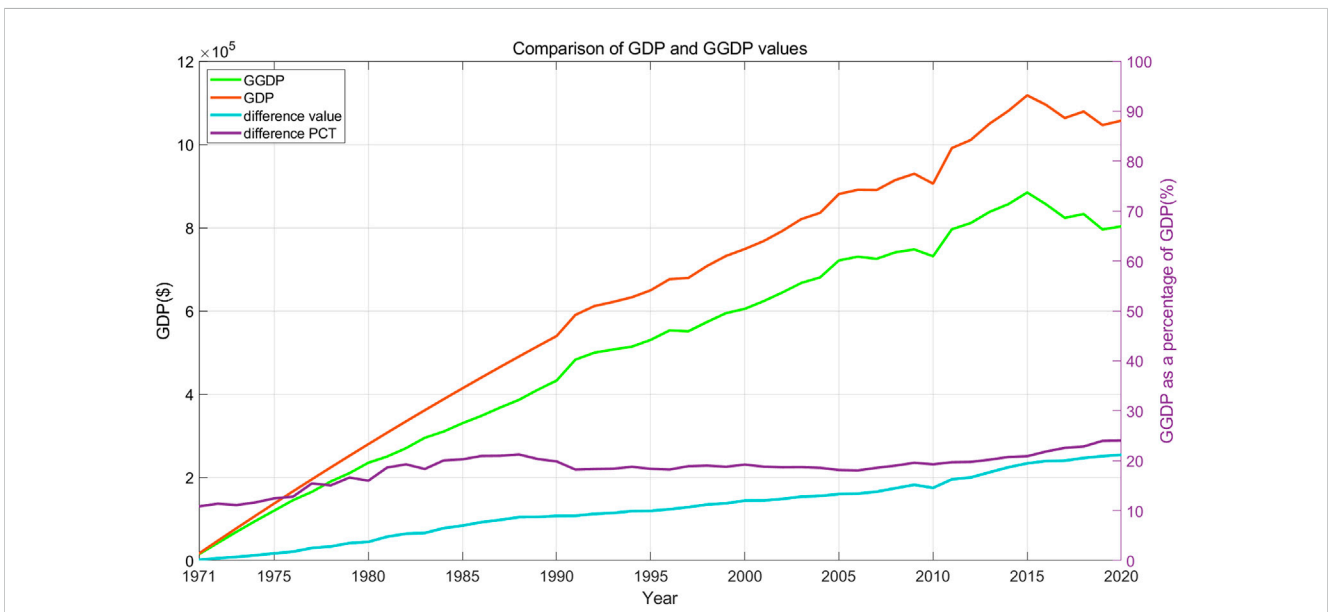


FIGURE 13 Comparison of GDP and GGDP and analysis of their differences.

of GGDP relative to GDP indicates an increasing exploitation of natural resources, suggesting a trend towards intensified environmental exploitation for developmental purposes over time. During the periods 2005–2010 and 2015–2020, both Brazil’s GDP and Green GDP indices experienced declines. Specifically, in

the first period, GDP and GGDP decreased by 1.292×10^5 trillion and 1.462×10^5 trillion, respectively, coinciding with an excessive utilization of forest resources. In the second period, GDP and GGDP declined by \$69.2 billion and \$78.3 billion, respectively, which correlated with the over-exploitation of water and forest

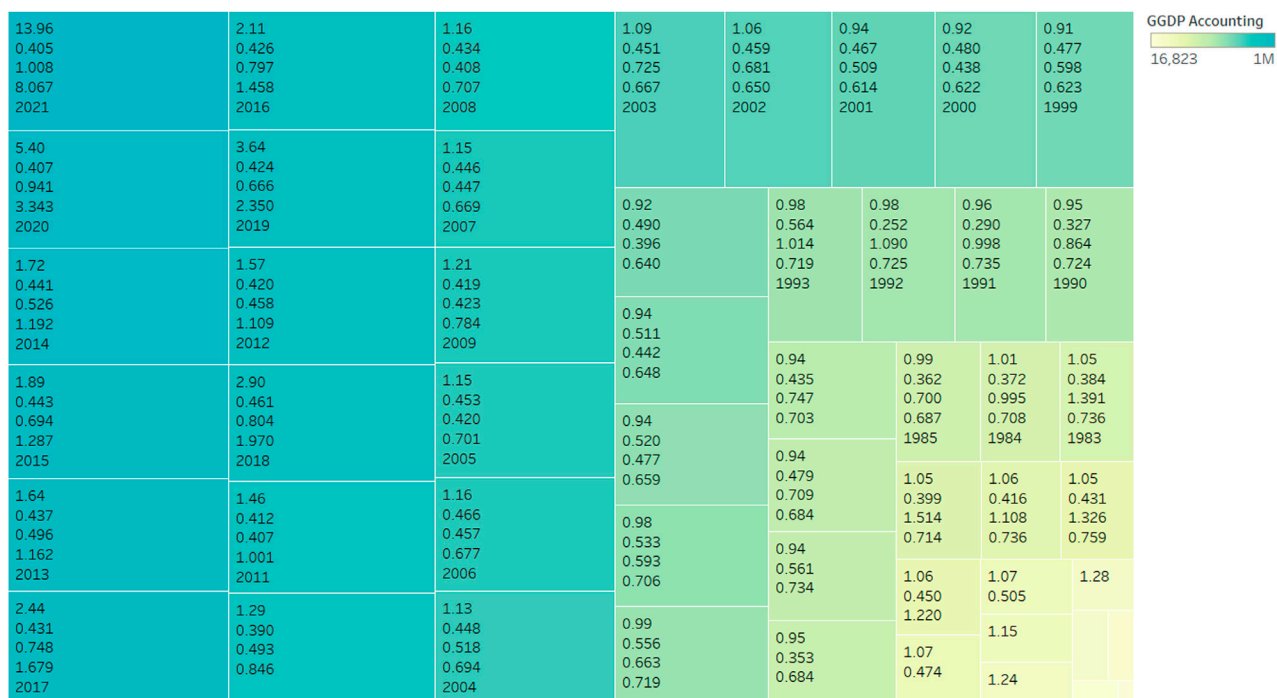


FIGURE 14 The contribution of each factor to the GGDP ratio.

resources. Figure 14 illustrates the contribution of each factor to the GGDP ratio, highlighting how resource use impacts GGDP.

Overall, the ratio of the monetary value of natural resources to GDP exhibited a slight increase, whereas the ratio of the monetary value of environmental resources to GDP remained relatively stable. Notably, the ratio of the monetary value of imported resources to GDP rose rapidly, surpassing the ratio of environmental resources after 2006. Figure 15 provides further detail on the contribution of each primary factor to GDP and the evolution of resource value, emphasizing the significant role of resources such as coal, crude oil, and plant foods in Brazil's economic growth.

These observations suggest that the decline in Brazil's Green GDP index is closely linked to the consumption of imported resources. Additionally, it emphasizes the pivotal role that resources such as coal, crude oil, and plant foods play in Brazil's economic growth, substantially influencing the nation's lasting economic growth trajectory (Wallace, 2000; Xu et al., 2005).

The graph defines the 4 consider our version that may influence GDP: variants in water resources, woodland land area, power use and usage, and CO₂ discharges. In time, it comes to be evident that carbon dioxide discharges and energy consumption exert the most considerable effect on GGDP growth, with a steady rise in consumption degrees.

Overall, Brazil's energy efficiency has actually been on a decline, while the environmental tons proportion has actually been climbing, indicating installing ecological pressures resulting from the country's economic growth. Simultaneously, the Power Sustainability Index (ESI) and the Improved Energy

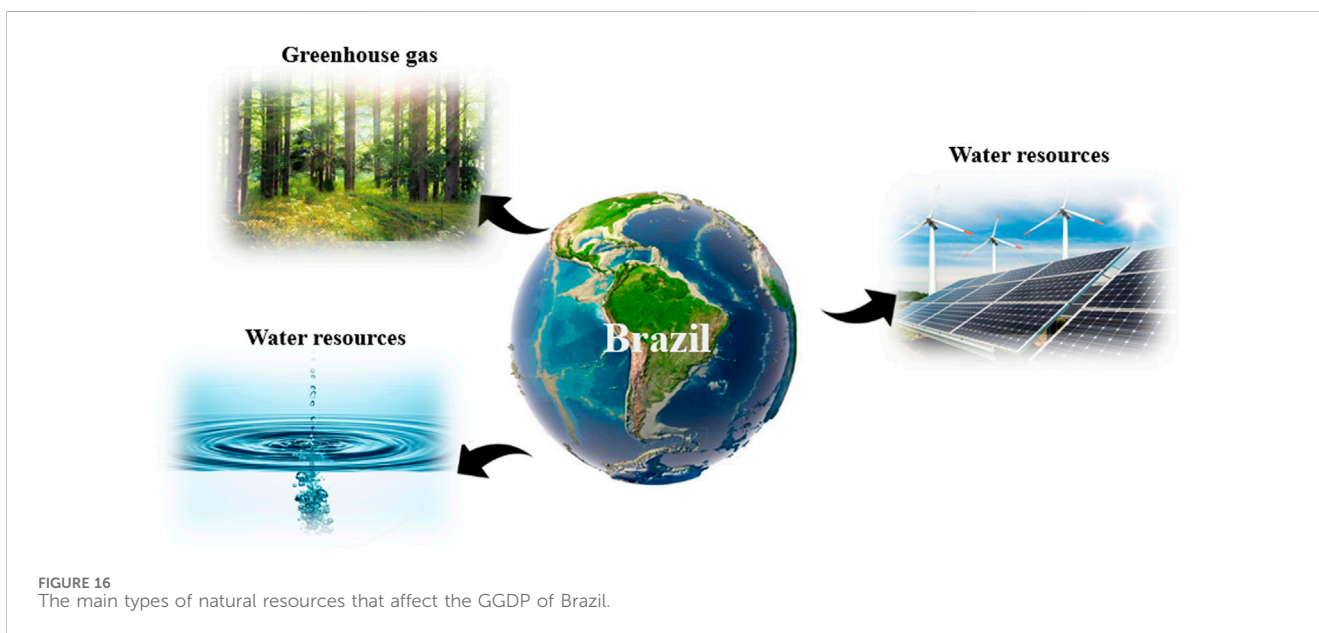
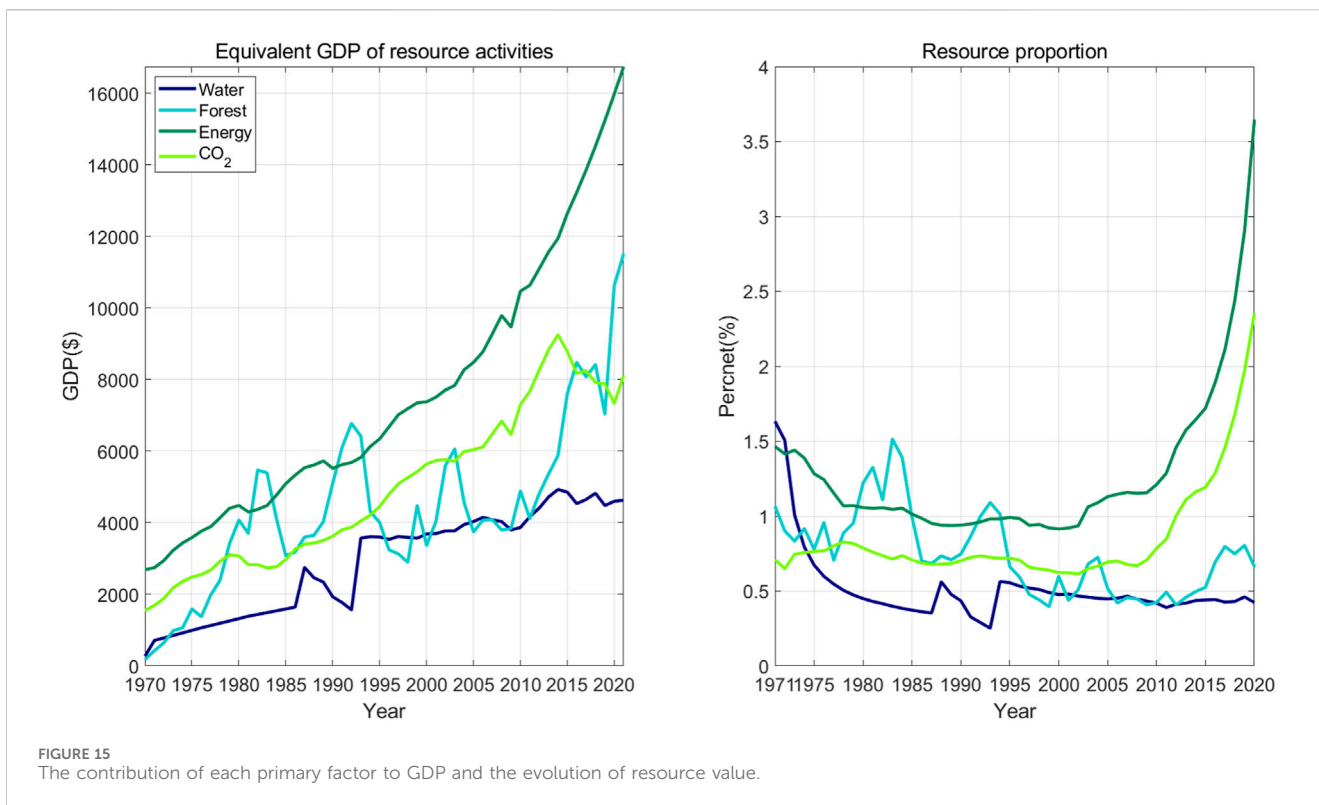
Sustainability Index (Improved ESI) have actually shown a decreasing trend from 1970 to 2020. These indices show a regular pattern, carefully lining up with the growth trend of energy performance (Xu et al., 2024).

Between 2010 and 2015, ESI worths were especially high, both going beyond 10, signifying inadequate total lasting advancement in Brazil during this period. Nonetheless, the ESI and Improved ESI values went down below 10 in 2013 and 2011, specifically, and further reduced to 3.848 and 2.225 by 2019. This suggests an enhancement in Brazil's lasting growth methods and suggests a potential change in the direction of even more lasting advancement methodologies.

5.2 Findings and analysis

5.2.1 Soybean production:

As one of the globe's leading producers and merchants of soybeans, Brazil's financial growth is considerably bolstered by this industry. Nonetheless, the manufacturing of soybeans is a significant vehicle driver of deforestation, causing damaging ecological repercussions such as loss of biodiversity and increased greenhouse gas exhausts. Subsequently, mitigating carbon dioxide emissions is essential for fostering GDP development. For Brazil to efficiently take on the GGDP structure, it needs to lower its dependence on unsustainable farming practices, such as monoculture, and transition to even more lasting agricultural techniques that secure and restore all-natural ecosystems.



5.2.2 Amazon rainforest

Brazil nurtures a substantial part of the Amazon rainforest, which supplies necessary community services, including carbon sequestration, water policy, and biodiversity preservation. The deterioration of this jungle positions lasting dangers to the nation’s economic wellness and overall wellness (United Nations Statistics Division, 2010). To execute the Inclusive Wealth Index (IWI), Brazil has to focus on the conservation and repair of

all-natural communities, especially the Amazon rainforest, and invest in sustainable land-use techniques that enhance the country’s natural funding.

5.2.3 Energy sustainability indices

The Energy Sustainability Index (ESI) and the Improved Energy Sustainability Index (Improved ESI) have shown a continuous decline, reaching values of 3.848 and 2.225, respectively, by 2019. Figure 16

illustrates these trends, highlighting the significant drop in these indices over time. These indices, which integrate financial and eco-friendly sustainable development research, underscore the immediate demand for Brazil to boost its sustainable development practices. Enhancing these practices is crucial for aligning economic growth with environmental preservation, ensuring long-term sustainability.

Must Brazil take on the GGDP index as the major step of financial wellness, it will demand a change in economic policies in the direction of more sustainable advancement methods, emphasizing the conservation and repair of natural environments. Although this transition may demand substantial short-term investments, it holds the potential for fostering more sustainable and resilient economic growth in the long term, yielding benefits for both current and future generations.

6 Conclusion

Over the past 50 years, there has been a 74% probability that the growth rate of Green GDP (GGDP) is positively correlated with temperature change. Utilizing the GGDP calculation model, we can analyze the development patterns and discrepancies between global GDP and green GDP, as well as examine the relationship between ecological efficiency, GGDP, and global average temperature. Although both green GDP and total GDP have shown continuous growth, the growth rate of green GDP is markedly slower, with its proportion decreasing from 90% in the 1970s to 80% in 2020. This decline highlights that an economic development model heavily reliant on natural resources has exacerbated the gap. Projected future CO₂ emissions are expected to increase annually, aligning with the trends of global warming and the expanding gap between GGDP and GDP. Hence, reducing excessive carbon dioxide emissions is crucial to mitigating global warming. The model we have developed aids in decision-making by allowing the control of parameters and variables to predict outcomes under different scenarios across various countries, demonstrating its wide applicability. Furthermore, by analyzing and modeling historical data, we can forecast future trends and developmental directions. This analysis underscores the profound benefits of

GGDP for both individual countries and the global community, showcasing its potential as a valuable tool in promoting sustainable economic growth and environmental stewardship.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

Author contributions

XZ: Funding acquisition, Investigation, Project administration, Resources, Software, Supervision, Writing—original draft. YC: Conceptualization, Data curation, Formal Analysis, Methodology, Validation, Visualization, Writing—review and editing.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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