



Spatial–Temporal Evolution and Factor **Decomposition for Ecological Pressure of Carbon** Footprint in the One Belt and One Road

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Abstract: Low-carbon and green development is important to promote the sustainable economic and social development of countries along the One Belt and One Road. These countries have distinct differences in their ability to withstand carbon emission pressures and their driving factors, due to different stages of development and economic characteristics. This paper presents a model of ecological pressure of the carbon footprint in energy consumption (EP_{cfec}) , defined by three states: light, balanced, and heavy pressure. The EP_{cfec} was calculated from data on 56 countries along the One Belt and One Road from 1994–2014, and analysis of the data's temporal and spatial evolutionary rules was conducted. Furthermore, we used the LMDI method to extract the driving forces of EP_{cfec} and evaluated the contribution of these factors to the overall region and seven sub-regions. The results showed that EP_{cfec} growth slowed over time, with the value of EP_{cfec} reaching 3190.51 in 2014. Resource-rich countries have a greater value of *EP*_{cfec} and are mainly distributed in parts of West Asia, North Africa, and Southeast Asia. The per capita export of goods and services, and the population density on productive land contribute to ecological pressure on the carbon footprint. Energy structure, the influence of international trade on GDP, and energy intensity exerted an inhibitory effect on the ecological pressure of the carbon footprint. This paper proposes mitigation measures for optimizing energy structure, improving energy efficiency, developing low energy consumption, and promoting green international trade. Our results provide support for countries along the One Belt and One Road to mitigate ecological pressures resulting from their carbon footprint.

Keywords: ecological pressure; energy consumption; spatial-temporal evolution; LMDI

1. Introduction

Actively responding to global climate change has become an urgent issue for all countries [1]. In December 2015, 196 countries signed the Paris Agreement to jointly promote a plan to curb global warming [2]. China is the first developing country to propose energy-saving and emission reduction targets. It has committed itself to reducing carbon emission intensity in 2020 to 55–60% of the levels in 2005 [3]. In November 2014, the Chinese government strived to realize the target of decreasing carbon emission as soon as possible [4,5] in accordance with the China–US Joint Statement on Climate Change. Many countries have also successively promulgated their own low-carbon action plans. At present, the situation of carbon emissions is still very serious, and is mainly driven by economic sectors with high energy consumption. Global greenhouse gas emission data show that in 2015, global carbon



emissions were 4.91 billion tons. The contributions of various economic sectors were as follows: electricity production: 29%, transportation: 27%, industry: 21%, commercial and residential: 12%, and agriculture: 9% [6]. The trend of high energy consumption and high emissions has caused tremendous ecological pressure in all countries. Economic activity is one of the most important factors for change in carbon emissions [7]. Decomposing the contribution of various economic development factors to changes in carbon emissions can provide an effective basis for formulating emission reduction policies, and coordinating the balance between economic growth and ecological protection [8].

The "Silk Road Economic Belt" and the "21st Century Maritime Silk Road" (One Belt and One Road) are open international initiatives proposed by the Chinese government. Their goal is to achieve shared benefits for all the countries along the route [9–11]. Green development is an inevitable choice for the countries along the One Belt and One Road, and an objective requirement for the development of the world economy. At present, most of the countries along the One Belt and One Road are developing countries. The development stage and economic characteristics of the various countries are quite different. In 2016, the amount of carbon emissions in the region accounted for 56.1% of global carbon emissions [12], and most of the countries had not established clear emission reduction targets and action paths. Therefore, selecting countries along the One Belt and One Road as targets to explore the issue of carbon emission pressures is of great importance for the global response to climate change issues. It is also an important means for the countries along the route to seek more feasible and effective emission reduction pathways and achieve sustainable development [13].

In recent years, many researchers have used carbon footprint to assess the combined environmental impacts of carbon emissions from human economic activities [14,15]. In 1992, Rees [16] and Wackernagel and Rees [17,18] proposed the concept of the ecological footprint. On this basis, Thomas Wiedmann and Jan Minx [19] proposed the concept of the carbon footprint, which is used to measure the total amount of carbon dioxide emissions directly or indirectly caused by an activity, or the total amount of carbon dioxide accumulated in the product life cycle. At the same time, some scholars have defined the carbon footprint as the productive land area needed to absorb carbon emissions, indicating the ecological footprint of carbon emissions [20]. Since then, others have analyzed carbon footprints from perspectives such as households [21], products [22,23], industries, and sectors [24,25], cities [26], countries [27], and so forth. Most of the research focuses on the industrial and regional levels. Brown et al. [28] took the example of 100 major metropolitan areas in the United States to analyze the carbon emissions from energy consumption in the transportation sector and the construction industry. Hertwich [29] analyzed the carbon footprint of 73 countries and 14 clustered regions. Based on energy consumption data, Chuai and Li [30] studied the temporal and spatial changes in the carbon footprint of six regions in Northeast China, North China, and East China, and concluded that the carbon emissions, carbon footprint, and carbon footprint intensity in East China were significantly higher than those in the other regions. Chen [31] used the Stochastic Impacts by Regression on Population, Affluence and Technology (STIRPAT) model and the partial least squares model to assess Beijing's carbon footprint of energy consumption from the perspectives of population, urbanization, and technology, and analyzed the impact of different factors on the carbon footprint.

Existing studies tend to focus on the carbon footprint itself and do not correlate the carbon footprint with other environmental factors. The introduction of ecological pressure into carbon footprint research can intuitively reflect the impact of carbon emissions on the ecological environment, which is an issue of notable interest. There is little relevant literature at present, and those papers that do exist are usually characterized by carbon footprint ecological pressure [32,33]. In order to seek the source of carbon footprint ecological pressure, it is necessary to decompose the contribution of its driving factors into the various ecological pressure affecting the carbon footprint [34]. Based on studies of the ecological pressure of the carbon footprint of energy consumption, Zhang et al. used logarithmic mean divisia index (LMDI) to decompose the driving forces of different factors. They found that the promotion of economic factors outweighed the inhibitory effects of other factors, and that the ecological pressure brought about by economic development was the most prominent factor [35].

At present, factor decomposition is mostly used in the study of carbon emission driving factors, and a flood of related research has been undertaken in this area since the mid-1980s. There are four main types of methods used. The first category comprises those that focus on identifying the factors affecting carbon emissions through structural decomposition techniques (SDA) under the input–output analysis framework, combined with input–output tables and models [36]. Feng [37] found that the economic growth of the United States during 1997–2007 was the key factor driving the increase of carbon emissions. The change in the energy structure over this period was an important factor in restraining carbon emission growth. Such studies often rely on input-output tables for non-continuous intertemporal analysis, and the types of impact factors are single, which makes it difficult to examine the impact of multiple types of socioeconomic factors. The second category uses the Computable General Equilibrium (CGE) model and scenario analysis to calculate the impact of different socioeconomic factors and their combinations on carbon emissions, in accordance with general equilibrium conditions. This type of research is also dependent on input-output tables, however it is not straightforward to uncover the underlying continuum of a time series [38,39]. The third category comprises the use of econometric models to study the impact of various factors on carbon emissions. There are many such methods, but all of them can reveal only statistical laws, not the intrinsic driving factors [40,41]. The fourth category uses the Kaya identity, along with extensions and improvements. They use factorial decomposition techniques (IDA) to calculate the contribution of different factors. This type of method has a variety of decomposition forms and can be selected based on research needs [42–45]. The logarithmic mean Divisia index (LMDI) method is such a method. It was proposed by Ang et al. in 1998, who deduced its mathematical properties, as well as summarizing a large number of applications of IDA models. Its advantages include using of diverse forms, simple calculations, and the ability to find intrinsic driving factors [46–49]. The method is used in many research fields, such as for the decomposition of energy consumption or carbon emission factors. For example, Vinuya et al. [50,51] used the LMDI decomposition model to find that economic and population growth is the largest factor driving carbon emissions. The increase in energy use efficiency and the decline in energy intensity can effectively restrain the increase in carbon emissions due to the impact of GDP and population increase. Ma and Stern improved the IDA model and used the LMDI method to decompose the energy intensity changes from 1980–2003, confirming that technological progress is the most important factor in reducing energy intensity [52]. Zhang et al. used the LMDI method to find that the rapid growth of carbon emissions in China's power generation industry is mainly due to the use of coal, and energy efficiency is also an important contributor [53]. Inglesi-Lotz applied LMDI to study the main factors affecting the changes in CO_2 emissions in South Africa. They found that energy intensity had a negative impact on CO_2 emissions during 2008–2014 [54]. The use of LMDI by Kopidou et al. showed that the two main drivers of industrial CO₂ emissions and employment are economic growth and resource intensity, and that the optimization of the energy structure is conducive to national emission reduction [55].

The One Belt and One Road is a shared, win–win, open economic and trade cooperation initiative, covering many countries and regions. At present, most of the research on One Belt and One Road focuses on strategic significance, institutional design [11,56], and capacity cooperation [57–59]. The primary focus has been how to promote the economic development of each country. However, there are several studies on the green and coordinated development of the international regional economy. The related literature focuses on the OECD [60,61], East Asia, ASEAN [62,63], the European Union and its sub-regions [17,64], and a few international regions, meaning less attention is paid to the coordinated development of the regional economy and environment in the One Belt and One Road area. In fact, most of the countries along the "One Belt and One Road" are developing or underdeveloped countries, many of which are not concerned with the problem of low carbon and emission reduction. In addition, the development stages and economic characteristics of the countries along the route are quite different. Therefore, selecting countries along the One Belt and One Road as the target

and exploring the carbon footprint ecological pressures of those countries is of great significance to sustainable development; it can also contribute to global emission reduction targets.

In summary, previous research has mainly focused on the measurement and causes of carbon emissions and carbon footprints. There has been little research on ecological pressure caused by carbon footprints, and it does not directly reflect the impact of carbon emissions on the ecological environment.

The LMDI method has been widely used in carbon emission and carbon footprint research. In previous work looking at causal factors, the choice of factors was not varied, with the main focus always being the country. Explanations of changes to ecological pressure resulting from the global carbon footprint were insufficient. In addition, the scope of the countries included was relatively limited, and the differences in economic characteristics and resource endowments among countries are relatively small, which limits the applicability of the research. By selecting countries along the One Belt and One Road with large differences in economic characteristics and resource endowments—based on an ecological-pressure-based model of energy consumption and the carbon footprint—we adopted the LMDI decomposition method to introduce the driving factors that reflect the links between the global carbon footprint and ecological pressure. Factors include the per capita export goods and services trade volume, and the influence of international trade on GDP. Our approach makes the study of the ecological pressures arising from energy consumption as measured by the carbon footprint more straightforward.

The rest of the paper is organized as follows: In Section 2, the concepts of the energy consumption carbon footprint (*CFEC*) and carbon footprint ecological pressure (EP_{cfec}) are proposed, and the measurement and factor decomposition model of EP_{cfec} are explained and constructed. Section 3 introduces the research scope, data sources, and regional divisions. Section 4 analyzes the calculation results and spatial evolution of EP_{cfec} from 1994–2014, and uses the LMDI method to decompose EP_{cfec} . The conclusions and recommendations are given in Section 5.

2. Methodology

2.1. Carbon Footprint Measurement Model of Energy Consumption

The carbon footprint is a quantified value that measures carbon emissions. It is based the concept of an ecological footprint and can directly measure the response of a natural system to carbon emissions from human activities [65]. In this paper, we use a measure of carbon footprint based on energy consumption, mainly related to coal, oil, and natural gas. The carbon footprint is measured using the productive land area needed to store energy consumption (see Equation (1)). According to the data provided by the IPCC (2006), productive land includes forests, grasslands, arable land, gardens, and other agricultural lands, of which forest and grassland carbon reserves account for 93% of the total productive land [66]:

$$CFEC = \sum_{i=1,}^{3} (Q_i \times EF_i) / \beta \quad (i = 1, 2, 3)$$
(1)

where *CFEC* refers to the carbon footprint of energy consumption (hm²); Q_i is the energy consumption of the *i*th energy (t); and *EF*_i is the energy carbon emission factor (t/tec), for which the study takes the average from several agencies, such as the EIA, as shown in Table 1 [67–69]. β is the conversion coefficient for converting carbon emissions into land area, for which the adopted WWF's value is 6.49 t/hm² [70].

Research Institute	Coal	Oil	Natural Gas
EIA	0.702	0.478	0.389
IEEJ	0.756	0.586	0.449
Energy Research Institute National Development and Reform Commission	0.747	0.582	0.443
Mean	0.735	0.549	0.427

Table 1. Carbon emission factors of various agencies.

2.2. The Econometric Model of EP_{cfec}

The ecological pressure of the carbon footprint in energy consumption (EP_{cfec}) refers to the pressure of carbon emissions on the natural ecosystem and is the ratio of the carbon footprint to productive land area. The calculation is as follows:

$$EP_{cfec,} = \frac{CFEC}{S_f + S_a + S_c + S_p},$$
(2)

where *CFEC* is the carbon footprint as described above; and S_f , S_a , S_c , and S_p refer to the area of forest, cultivated land, permanent cropland, and permanent ranching land, respectively, with the unit hm².

When $EP_{cfec} \in (0, 1)$, the productive land can fully absorb the carbon emissions resulting from energy consumption, and the carbon emissions, therefore, exert little pressure on the ecological environment. The smaller the value of EP_{cfec} , the lower the ecological pressure. This is termed the light pressure state.

When $EP_{cfec} = 1$, the carbon emissions generated by energy consumption are equal to those of being absorbed by productive land; that is, the productive land has reached its capacity to absorb the carbon emissions from energy consumption. This is the balanced state.

When $EP_{cfec} \in (1, +\infty)$, the productive land cannot fully absorb the carbon emissions, and the ecological environment faces a large amount of pressure from carbon storage. The larger the value of EP_{cfec} , the greater the ecological pressure on the carbon footprint. At this point, it is in the high pressure state.

A schematic diagram of the three pressure states is shown in Figure 1.

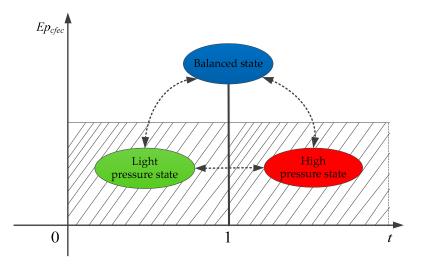


Figure 1. Pressure state diagram.

2.3. Decomposition Method of EP_{cfec}

The paper applies the LMDI method to the decomposition of ecological pressure factors influencing the carbon footprint. The model introduces two factors related to the EP_{cfec} —the influence

of international trade on GDP, per capita exports of goods and service trade—reflecting linked changes in the ecological pressure between countries. We decomposed the driving factors into five categories: energy structure, energy intensity, the influence of international trade on GDP, per capita export goods and services trade volume, and population intensity of productive land (see Equation (3)). The variables are defined in Table 2.

$$EP_{cfec,} = \frac{C}{Q} \times \frac{Q}{E} \times \frac{E}{Y} \times \frac{Y}{EXP} \times \frac{EXP}{POP} \times \frac{POP}{H} \times \varepsilon = EF \times ES \times EI \times YE \times EP \times PH \times \varepsilon, \quad (3)$$

where *C* is the carbon emission of non-renewable energy as described above; *Q* is the consumption of non-renewable energy; *E* is the total energy consumption including renewable energy and non-renewable energy; *Y* is the gross domestic product; *EXP* is exports of goods and services; *POP* is the total population of every country; *H* is the productive land area; ε means coefficient; *EF* means carbon emission intensity, which is the *i*th energy carbon emission coefficient; *ES* means *the* energy structure that is the proportion of non-renewable energy in the total energy consumption; *EI* means the energy intensity that is the ratio of total energy consumption to GDP; *YE* means the influence of international trade on GDP, which is the ratio of GDP to exports of goods and services; *EP* means the per capita exports of goods and service trade, which is the ratio of exports of goods and services trade to the population; *PH* means the population intensity of productive land, which is the ratio of the population to productive land; and ε means the reciprocal of the conversion coefficient β , as shown in Table 2.

Variable Parameters	Variable Name Variable Meaning		Unit		
EP _{cfec}	Ecological pressure of the carbon footprint	Ecological pressure of the carbon footprint in energy consumption			
С	Carbon emission of non-renewable energy	Carbon emissions arising from non-renewable energy	t		
Q	Consumption of non-renewable energy	Consumption of non-renewable energy	tec		
E	Total energy consumption	Total energy consumption includes renewable energy and non-renewable energy	tec		
Ŷ	GDP	Gross domestic product	dollar		
EXP	Exports of goods and services	Exports of goods and services	dollar		
POP	Total population	Total population	Person		
Н	Productive land area	Productive land area	hm ²		
EF	Carbon emission intensity	The <i>i</i> th energy carbon e dollar mission factor	t/tec		
ES	Energy structure	The proportion of non-renewable energy in the total energy consumption	%		
EI	Energy intensity	Ratio of total energy consumption to GDP	tec/10 ³ dollar		
YE	Influence of international trade on GDP	The ratio of GDP to exports of goods and services	%		
EP	Per capita exports of goods and service trade	The ratio of exports of goods and services trade to population	10 ³ dollar/Person		
PH	Population intensity of productive land	Ratio of population to productive land	person/hm ²		
ε	Coefficient	The reciprocal of the conversion coefficient B			

Table 2. Definition of variables.

The contribution of each individual factor to the ecological pressure was calculated by using the LMDI model in Equation (4), and the rate of change of each individual factor was calculated by the decomposition in Equation (5).

$$\Delta EP_{cfec,} = \Delta EP_{EF} + \Delta EP_{ES} + \Delta EP_{EI} + \Delta EP_{YE} + \Delta EP_{EP} + \Delta EP_{PH} + \Delta EP_{\varepsilon}, \tag{4}$$

where:

$$\Delta EP_{ES} = W^T \times \ln \frac{ES^1}{ES^0} \qquad \Delta EP_{EI} = W^T \times \ln \frac{EI^1}{EI^0} \Delta EP_{YE} = W^T \times \ln \frac{YE^T}{YE^0} \qquad \Delta EP_{EP} = W^T \times \ln \frac{EP^T}{EP^0} \Delta EP_{PH} = W^T \times \ln \frac{PH^T}{PH^0} \qquad W^T = \frac{EP_{cfec}^T - EP_{cfec}^0}{\ln \left(EP_{cfec}^T / EP_{cfec}^0\right)}$$
(5)

where:

$$D = EP_{cfec}^{T} / EP_{cfec}^{0} = D_{EF} \times D_{ES} \times D_{EI} \times D_{YE} \times D_{EP} \times D_{PH} \times D_{\varepsilon}$$
$$D_{ES} = \frac{ES^{T}}{ES^{0}} D_{EI} = \frac{EI^{T}}{EI^{0}} D_{YE} = \frac{YE^{T}}{YE^{0}} D_{EP} = \frac{EP^{T}}{EP^{0}} D_{PH} = \frac{PH^{T}}{PH^{0}}$$
(6)

In Equations (4) and (5), ΔEP_{cfec} is the change in EP_{cfec} ; ΔEP_{EF} , ΔEP_{ES} , ΔEP_{EI} , ΔEP_{YE} , ΔEP_{EP} , ΔEP_{PH} , and ΔEP_{ε} denote the change in carbon intensity, energy structure, energy intensity, influence of international trade on GDP, per capita exports of goods and service trade, population intensity of productive land, and the conversion coefficient from the 0th to the *T*th year, respectively—these are the contribution values; W^T is the weight of influence; *D* represents the change in EP_{cfec} , D_{EF} , D_{ES} , D_{EI} , D_{YE} , D_{EP} , D_{PH} , and D_{ε} are the rates of change in carbon intensity, energy structure, energy intensity, Influence of international trade on GDP, per capita import and export goods and service trade, population intensity of productive land, and the conversion factor from the 0th to the *T*th year, respectively. ES^0 , EI^0 , YE^0 , EP^0 , PH^0 and ES^T , EI^T , YE^T , EP^T , PH^T denote the energy structure, energy intensity, influence of international trade on GDP, per capita export goods and service trade volume, and population intensity of productive land in the 0th year and *T*th year, respectively.

3. Data Sources

According to the bilateral trade agreement of the One Belt and One Road Initiative, combined with the availability of data, 56 countries along the route were selected, as shown in Table 3. Due to the unavailability of data eight countries were excluded: Armenia, Bosnia and Herzegovina, Hungary, Moldova, Montenegro, Nepal, Serbia, and East Timor. The time period used was 1994–2014.

Region	Country Code
Mongolia and Russia	RUS, MNG
Central Asia	KAZ, KGZ, TJK, TKM, UZB
West Asia and North Africa	QAT, ARE, KWT, ISR, BHR, SAU, OMN, TUR, LBN, AZE, IRQ, IRN, JOR, GEO, EGY, YEM, SYR
Central and Eastern	SVN, EST, CZE, SVK, LTU, LVA, POL, HRV, ROM, BLR, BGR, MKD, ALB, UKR
Southeast Asia	SGP, BRN, MYS, THA, IDN, PHL, VNM, LAO, MMR, KHM
South Asia	MDV, LKA, BTN, IND, PAK, BGD, AFG
Eastern Asia	CHN

Table 3. The 56 countries along the One Belt and One Road.

The information analysis center of the Oak Ridge National Laboratory in the United States Carbon Dioxide Information Analysis Center (CDIAC) provides carbon emission data [71]. The primary energy sources for the countries studied mainly comprise coal, oil, and natural gas. Compiling data on energy consumption and carbon emissions shows that the carbon emissions of countries along the One Belt and One Road are on the rise. The highest value of the carbon emissions in 2012 was 486.41 × 10⁷ t after which a downward trend occurred, showing that various countries began to focus on green and low-carbon development models. Carbon emissions from the energy consumption of coal, oil, and natural gas are also decreasing. From 2000–2012, the carbon emissions from oil and natural gas production have changed by a small amount, and remained within the range 50×10^7 – 70×10^7 t. Of these, coal combustion is the main source of carbon emissions and is growing rapidly. In 2012, the carbon consumption of coal consumption reached 251.9 × 10⁷ t, which was 1.5 times that of oil and natural gas consumption. The trends are shown in Figure 2.

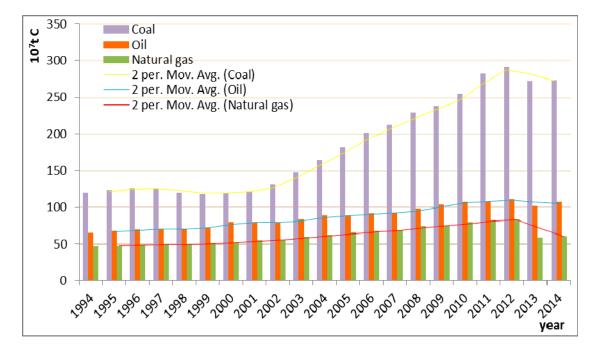


Figure 2. The changes of carbon emissions of different energy consumption modes from 1994–2014.

Data about the productive land area, population, and GDP was taken from the World Bank public data resource [72]. Taking into account the integrity and continuity of the data, a small amount of data was processed using interpolation. In order to analyze regional differences, 56 countries along the route were divided into seven regions: Mongolia, Russia, Central Asia, West Asia, North Africa, Central and Eastern Europe, Southeast Asia, South Asia, and East Asia (Table 3). The spatial layout is shown in Figure 3.

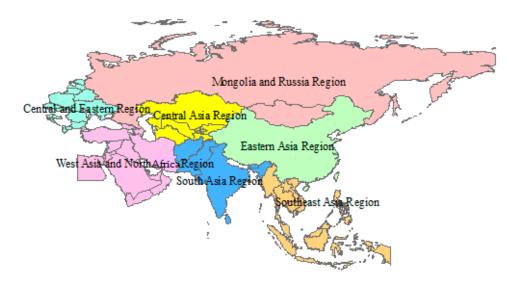


Figure 3. The regional distribution of 56 countries.

4. Results and Analysis

4.1. The Carbon Footprint in Energy Consumption

The changes of the carbon footprint in energy consumption (CFEC) derived from Equation (1) for 56 countries, from 1994–2014, are shown in Figure 4.

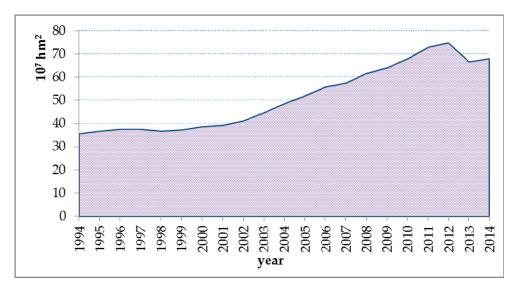


Figure 4. The changes in the sum of CFEC of 56 countries from 1994–2014.

Figure 4 indicates that the carbon footprint of countries along the One Belt and One Road line show a single-peak evolution over time. The value from 1994–2000 remained basically unchanged. With the increase of energy consumption, the carbon footprint increased during the period 2001–2012. In 2012, the peak value of 74.9×10^7 hm² was reached, after which there was a downward trend. This is the same trend as for carbon emissions of countries along the One Belt and One Road during the same period.

4.2. Ecological Pressure of the Carbon Footprint in Energy Consumption

The EP_{cfec} values of 56 countries from 1994–2014 were calculated according to Equation (2), and are shown in Tables 4 and 5.

Country	1994	1996	1998	2000	2002	2004	2006	2008	2010	2012	2014
AFG	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.01	0.01	0.01
ALB	0.01	0.01	0.00	0.00	0.00	0.01	0.00	0.00	0.02	0.02	0.03
AZE	0.29	0.29	0.26	0.25	0.28	0.30	0.28	0.24	0.28	0.29	0.00
BGD	0.08	0.11	0.10	0.12	0.14	0.16	0.19	0.20	0.24	0.28	0.30
BGR	0.36	0.38	0.36	0.32	0.34	0.36	0.38	0.40	0.36	0.36	0.35
BLR	0.29	0.27	0.26	0.24	0.24	0.26	0.28	0.28	0.28	0.29	0.29
BTN	0.01	0.02	0.02	0.02	0.03	0.02	0.02	0.03	0.03	0.04	0.05
CHN	0.23	0.26	0.24	0.25	0.28	0.39	0.48	0.56	0.64	0.73	0.74
CZE	1.20	1.20	1.17	1.19	1.15	1.12	1.18	1.12	1.08	0.98	0.94
EST	0.70	0.74	0.64	0.61	0.85	0.90	0.72	0.77	0.78	0.75	0.81
GEO	0.08	0.06	0.07	0.06	0.04	0.07	0.10	0.08	0.10	0.13	0.13
HRV	0.28	0.34	0.39	0.63	0.69	0.72	0.70	0.67	0.59	0.53	0.43
IDN	0.20	0.22	0.18	0.21	0.24	0.24	0.26	0.30	0.30	0.44	0.30
IND	0.17	0.20	0.21	0.23	0.23	0.25	0.28	0.34	0.37	0.44	0.49
IRN	0.14	0.16	0.15	0.23	0.24	0.24	0.38	0.42	0.45	0.47	0.50
IRQ	0.32	0.32	0.33	0.34	0.39	0.47	0.39	0.40	0.48	0.61	0.62
JOR	0.46	0.50	0.52	0.56	0.62	0.70	0.79	0.83	0.81	0.90	0.96

Table 4. Change values of EP_{cfec} in 1994–2014 in light pressure states.

Country	1994	1996	1998	2000	2002	2004	2006	2008	2010	2012	2014
-											
KAZ	0.03	0.02	0.02	0.03	0.03	0.04	0.05	0.05	0.05	0.05	0.00
KGZ	0.02	0.02	0.02	0.02	0.02	0.02	0.02	0.03	0.02	0.04	0.03
KHM	0.01	0.01	0.02	0.02	0.02	0.02	0.02	0.03	0.03	0.04	0.05
LAO	0.01	0.01	0.01	0.02	0.02	0.03	0.03	0.01	0.02	0.02	0.01
LKA	0.09	0.12	0.13	0.17	0.19	0.21	0.19	0.18	0.20	0.23	0.27
LTU	0.23	0.19	0.19	0.15	0.19	0.21	0.20	0.23	0.20	0.20	0.18
LVA	0.18	0.20	0.19	0.16	0.18	0.18	0.18	0.18	0.18	0.15	0.14
MKD	0.33	0.37	0.40	0.40	0.33	0.36	0.36	0.35	0.31	0.28	0.24
MMR	0.02	0.03	0.03	0.04	0.03	0.04	0.04	0.03	0.04	0.03	0.07
MNG	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.01	0.01	0.01
PAK	0.09	0.10	0.11	0.11	0.12	0.14	0.16	0.17	0.17	0.17	0.17
PHL	0.19	0.21	0.23	0.25	0.24	0.25	0.22	0.25	0.27	0.27	0.32
POL	0.73	0.79	0.72	0.66	0.72	0.76	0.82	0.82	0.89	0.84	0.80
ROM	0.33	0.35	0.27	0.25	0.27	0.28	0.30	0.28	0.22	0.24	0.20
RUS	0.30	0.28	0.28	0.28	0.29	0.30	0.31	0.30	0.33	0.31	0.00
SAU	0.07	0.06	0.05	0.07	0.08	0.09	0.10	0.10	0.12	0.13	0.14
SVK	0.67	0.68	0.65	0.59	0.70	0.80	0.80	0.76	0.74	0.68	0.63
SVN	0.97	1.13	1.22	1.09	1.20	1.26	1.30	1.11	0.99	0.96	0.81
SYR	0.11	0.12	0.14	0.14	0.11	0.14	0.15	0.19	0.17	0.12	0.08
THA	0.25	0.32	0.32	0.35	0.41	0.48	0.49	0.49	0.53	0.54	0.56
TJK	0.02	0.03	0.02	0.02	0.02	0.02	0.02	0.02	0.02	0.02	0.04
TKM	0.04	0.04	0.04	0.04	0.05	0.06	0.06	0.07	0.07	0.08	0.08
TUR	0.15	0.18	0.19	0.21	0.19	0.21	0.25	0.28	0.29	0.32	0.34
UKR	0.44	0.41	0.33	0.32	0.32	0.34	0.32	0.31	0.30	0.30	0.23
UZB	0.16	0.16	0.18	0.18	0.20	0.19	0.18	0.19	0.16	0.18	0.16
VNM	0.14	0.17	0.22	0.22	0.26	0.33	0.36	0.40	0.44	0.44	0.52
YEM	0.02	0.02	0.02	0.02	0.03	0.03	0.04	0.04	0.04	0.03	0.04

Table 4. Cont.

Table 5. Change values of EP_{cfec} in 1994–2014 in high pressure states.

Country	1994	1996	1998	2000	2002	2004	2006	2008	2010	2012	2014
MYS	0.48	0.66	0.59	0.66	0.70	0.87	0.84	1.03	1.04	1.06	1.18
LBN	0.76	0.82	1.00	0.98	0.99	0.96	0.79	0.96	1.14	1.27	1.36
OMN	0.54	0.54	0.56	0.70	0.71	0.65	1.00	1.09	1.22	1.54	1.59
EGY	1.00	1.23	1.44	1.65	1.39	1.65	1.91	2.12	2.07	2.15	1.99
ISR	3.33	3.74	4.02	4.25	4.28	4.63	4.93	5.35	5.51	5.78	4.78
MDV	1.04	1.52	1.58	2.10	2.08	2.55	3.98	4.42	4.97	5.90	7.09
BRN	11.11	12.04	13.43	11.69	10.05	11.99	12.49	23.59	18.87	22.45	19.55
ARE	7.84	3.97	7.77	8.10	5.88	8.00	8.88	11.09	15.73	18.20	21.95
KWT	11.08	14.16	14.72	14.76	15.85	17.29	20.21	22.54	24.37	27.83	25.71
QAT	19.38	20.34	20.32	21.73	25.84	27.11	35.10	34.53	44.04	57.73	66.29
BHR	68.49	72.48	85.36	84.91	71.52	83.57	91.52	151.32	140.00	126.95	149.36
SGP	1853.18	1480.71	1462.29	1492.58	1452.46	1241.78	1416.69	1758.35	2587.78	1710.52	2876.59

In general, the EP_{cfec} has increased in all countries. Of these, 44 countries are under light pressure and 12 countries are under high pressure. Some countries are already under excessive high pressure (i.e., the EP_{cfec} of SGP reached 2876.59, which is 50.49 times the average value of 56.97 in 2014). Carbon emissions from energy consumption in some countries, such as Singapore, Bahrain, Qatar, Kuwait, UAE, Israel, and Brunei, have far exceeded the digestion capacity of the productive land. On the one hand, these countries are more dependent on non-renewable energy, and on the other hand, the area of productive land is relatively small. Countries with light pressure can also naturally digest carbon emissions from energy consumption, especially countries with an EP_{cfec} of less than 0.1, such as Kazakhstan, Mongolia, and Saudi Arabia. This is directly related to the large area of productive land in these countries. Therefore, protecting or expanding productive land area has a direct impact on reduction of the EP_{cfec} . Figure 5 shows the spatial evolution of the EP_{cfec} in 56 countries in 1994, 2002, 2008, and 2014. It can be seen that there is a significant difference in the EP_{cfec} among these countries. Mongolia, Kazakhstan, Kyrgyzstan, Tajikistan, Afghanistan, Laos, Cambodia, Myanmar, Yemen, and other countries were under light pressure. The EP_{cfec} is always below 0.05 and the change is small. Although China, Iraq, Jordan, and Thailand, among others, are under light pressure, but with a rapid rate of increase—the EP_{cfec} in these countries has increased from a relatively low level to over 0.5, which is related to the increase in carbon emissions in these countries. Oman, Lebanon, and Malaysia have changed from light pressure to heavy pressure, which is related to the small changes in productive land in these countries, the EP_{cfec} has shown a downward trend, and some countries have shifted from a heavy state of pressure to a light pressure state. This is related to the decline in carbon emissions in these countries from a downward trend, and some countries have shifted from a heavy state of pressure to a light pressure state. This is related to the decline in carbon emissions in these countries from coal are decreasing.

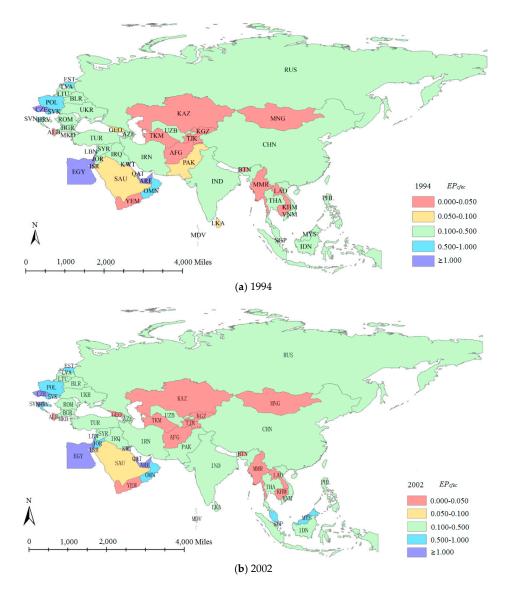


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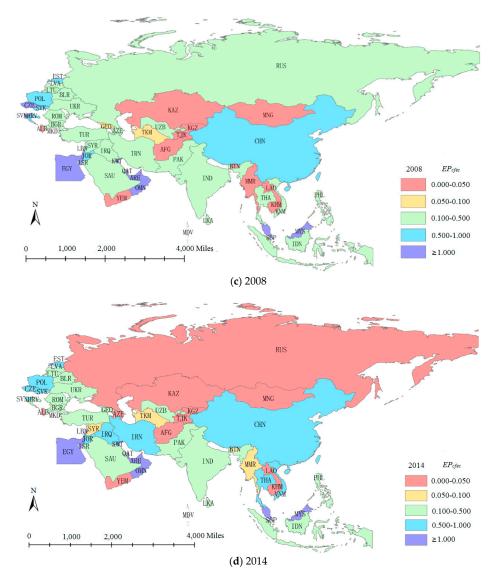
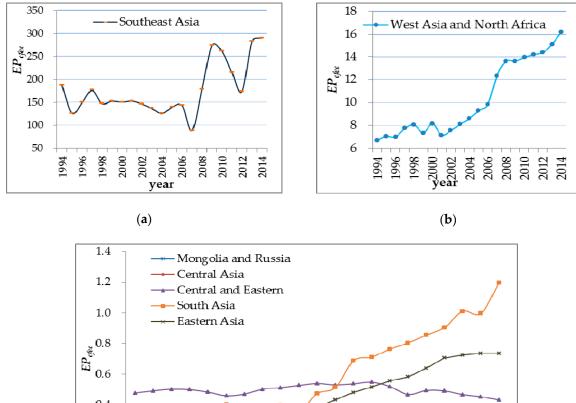


Figure 5. Spatial changes of the *EP*_{cfec} in (**a**) 1994; (**b**) 2002; (**c**) 2008; and (**d**) 2014.

Figure 6 shows the time variation of the EP_{cfec} in seven regions. The areas with greater ecological stress are mainly located in Southeast Asia, West Asia, and North Africa, and the lowest EP_{cfec} value is 6.71. However, in Central Asia, where the level of income is relatively low, and in Mongolia and Russia, where the area of productive land is relatively large, the ecological pressure is relatively low, and the EP_{cfec} value is between 0.05 and 0.17.



0.4 0.2 0.0 2014 year (c)

Figure 6. The EP_{cfec} of the seven regions from 1994–2014.

4.3. Driving Factor Decomposition of EP_{cfec}

4.3.1. Factors Decomposition for the Overall Region

Equations (3)–(5) were used to decompose the ecological pressure of the carbon footprint in energy consumption, and calculate the contribution and rate of change of EP_{cfec} for different factors. Figure 7 shows changes in the contribution value of each factor to EP_{cfec} , and Figure 8 shows the rate of change of each factor. A contribution value greater than 0 indicates that the factor has a promoting effect on the EP_{cfec} , and a value of less than 0 shows an inhibitory effect. The greater the absolute value, the greater the influence.

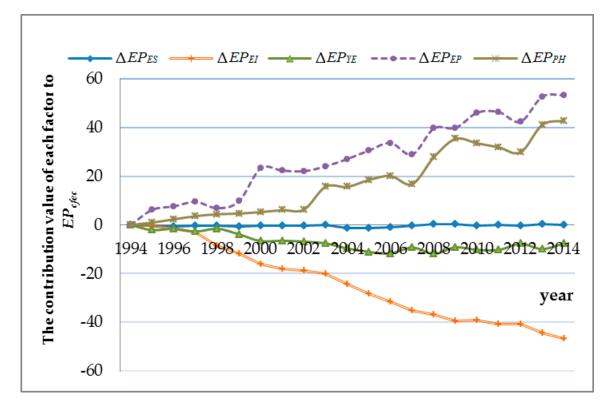


Figure 7. The contribution of *ES*, *EI*, *YE*, *EP*, and *PH* to the *EP*_{cfec}.

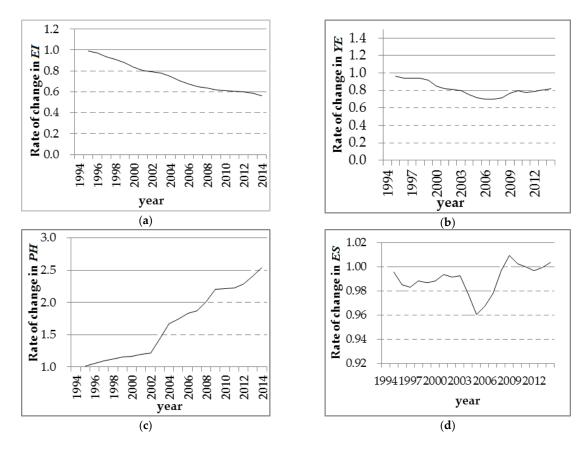


Figure 8. Cont.

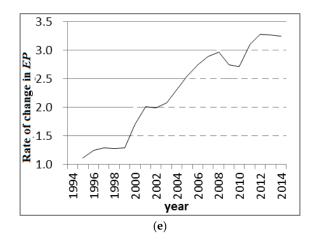


Figure 8. The rate of change in *ES*, *EI*, *YE*, *EP*, *PH* to *EP*_{cfec}.

Note that in the original statistics for Myanmar, the maximum value of the influence of international trade on GDP is 1005.36 and the minimum is 4.98, which is abnormal. Including the factor analysis of the factors in the overall region, the contribution of other factors to the EP_{cfec} was diluted. Therefore, Myanmar was excluded from this part of the analysis.

As can be seen from Figure 7, the five factors that contributed to EP_{cfec} in 2014 are ranked as follows: energy intensity, influence of international trade on GDP, energy structure, population intensity of productive land, and per capita exports of goods and service trade. Among them, three factors (energy intensity, influence of international trade on GDP, and energy structure) play a decisive role; two factors (population intensity of productive land, per capita export goods and service trade) play a catalytic role.

(1) Energy intensity (*EI*): From 1994–2014, EI had an inhibitory effect on the EP_{cfec} gradually increased over time. The contribution in 2014 was –46.64, and the rate of change was 0.55, as shown in Figure 8a. The energy consumption intensity of the countries also decreased significantly over the same period. Countries should continue to implement energy-saving and emission-reduction measures, significantly reduce energy consumption per unit, in order to strengthen the inhibitory effect of energy consumption intensity on the EP_{cfec} .

(2) Influence of international trade on GDP (*YE*): in 1994–2014, the contribution of this factor was between -12 and 0, and the rate of change was between 0.69 and 1, as shown in Figure 8b. This factor has an inhibitory effect on the EP_{cfec} but shows a slight weakening trend. In the same period, *YE* gradually decreased and export dependency rose slowly. This shows that international trade has gradually enhanced the economic growth of countries along the One Belt and One Road and has also increased the ecological pressure arising from the carbon footprint. Therefore, countries along the One Belt and One Road urgently need to develop their foreign trade in accordance with low-carbon, environmental protection, and green models.

(3) Energy structure (*ES*): from 1994–2014, the contribution of the ES to the EP_{cfec} was between –1.3 and 0.36, and the rate of change was between 0.96 and 1.01, as shown in Figure 8c. The impact on the EP_{cfec} fluctuated, but not by much. In general, it showed inhibitory effects and only acted as a promoter in 2008 and 2011 during which time its promotion was small. The ES mainly depends on the proportion of non-renewable energy, such as coal, oil, and natural gas. This proportion has been relatively large in the countries along the One Belt and One Road, and has remained at about 78%. Therefore, there is a very great potential for reducing the ecological pressure of the carbon footprint in energy consumption by adjusting the energy structure.

(4) Population intensity of productive land (*PH*): This factor has a significant role in increasing the EP_{cfec} and is rising year by year. In 2014, the contribution was 42.86 and the rate of change was 2.57, as shown in Figure 8d. In the same period, the *PH* increased continuously, and the population

intensity in 2014 was 2.57 times that of 1994. Countries along the One Belt and One Road are mostly developing countries and underdeveloped countries, with a large population base and strong growth inertia. It is hard to reduce the population size in the short term. Therefore, the key to reduce the EP_{cfec} is to protect or expand the area of productive land, such as forests and grasslands.

(5) Per capita exports of goods and services (*EP*): This factor made the largest contribution to the EP_{cfec} . From 1999–2014, the contribution to the EP_{cfec} rose dramatically from 9.72–53.39, an increase of 5.49 times, and the rate of change climbed from 1.35 to 3.24, an increase of 2.4 times, as shown in Figure 8e. In the same period, the EP also experienced rapid growth, which promoted the economic growth of all countries. This shows that the export trade of goods and services in certain countries is in a period of high growth, accompanied by high energy consumption. This is consistent with the *YE* factor. Therefore, the development of low-carbon and green international trade will be the key to reducing the EP_{cfec} . While actively conducting foreign trade, countries along the One Belt and One Road need to provide new impetus for foreign trade development through means such as technological innovation and structural adjustment.

4.3.2. Decomposition Analysis of Factors in Different Sub-Regions

Changes of the studied ecological pressure drivers across the seven sub-regions are summarized in Table 6.

Period	Region	ΔEP_{ES}	D _{ES}	ΔEP_{EI}	D _{EI}	ΔEP_{YE}	DYE	ΔEP_{EP}	D _{EP}	ΔEP_{PH}	D _{PH}
	1	0.00	0.99	-0.01	0.91	-0.01	0.97	0.01	1.05	0.00	1.00
	2	0.00	0.99	0.00	0.95	-0.01	0.89	-0.01	0.87	0.00	1.03
	3	-0.06	0.99	-0.81	0.90	-0.30	0.96	2.19	1.38	0.88	1.13
1994–1999	4	-0.01	0.98	-0.05	0.90	-0.01	0.98	0.34	2.09	0.01	1.02
1//1 1///	5	-5.67	0.97	4.47	1.03	-23.13	0.88	19.01	1.12	13.81	1.09
	6	0.02	1.08	-0.01	0.98	-0.02	0.92	0.05	1.19	0.01	1.06
	7	0.00	1.02	-0.04	0.87	0.01	1.03	0.04	1.20	0.01	1.03
	1	0.00	0.97	-0.04	0.77	-0.04	0.78	0.07	1.56	0.00	0.99
	2	0.00	0.96	-0.01	0.81	-0.02	0.70	0.02	1.40	0.01	1.11
	3	-0.10	0.99	-1.88	0.77	-1.84	0.78	12.85	6.27	2.28	1.37
2000-2004	4	-0.02	0.96	-0.17	0.71	-0.10	0.82	0.56	3.13	0.06	1.12
	5	-1.07	0.99	-12.37	0.93	-54.74	0.72	59.16	1.45	51.80	1.40
	6	0.02	1.10	-0.03	0.90	-0.04	0.86	0.12	1.48	0.01	1.02
	7	0.02	1.06	-0.10	0.68	-0.07	0.79	0.23	2.47	0.02	1.08
	1	0.00	0.98	-0.07	0.65	-0.03	0.85	0.10	1.94	0.00	0.98
	2	0.00	0.99	-0.03	0.64	-0.01	0.88	0.04	1.92	0.01	1.19
	3	-0.03	1.00	-3.19	0.70	-3.81	0.65	18.54	7.95	5.69	1.88
2005-2009	4	-0.03	0.95	-0.29	0.56	-0.16	0.73	0.78	4.75	0.06	1.12
	5	6.77	1.04	-43.42	0.78	-59.68	0.70	102.36	1.82	123.80	2.02
	6	-0.01	0.99	-0.11	0.77	-0.09	0.81	0.72	6.43	0.11	1.30
	7	0.04	1.13	-0.16	0.64	-0.18	0.60	0.58	5.16	0.04	1.12
	1	0.00	0.98	-0.05	0.60	0.00	1.00	0.08	2.01	0.00	0.98
	2	0.00	1.00	-0.04	0.49	0.00	1.07	0.04	1.88	0.02	1.32
	3	-0.01	1.00	-3.73	0.69	-3.51	0.71	21.77	8.44	8.93	2.40
2010-2014		-0.04	0.91	-0.33	0.49	-0.23	0.62	0.86	6.13	0.03	1.06
	5	11.02	1.05	-66.20	0.73	-69.80	0.72	133.50	1.87	187.59	2.40
	6	0.13	1.29	-0.15	0.75	0.04	1.08	1.12	9.14	0.19	1.44
	7	0.06	1.14	-0.26	0.55	-0.12	0.76	0.78	6.25	0.06	1.15

Table 6. Fa	ctor decomp	osition of the EP	for the seven	regions studied	from 1994–2014.
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Note: Region number: 1. Mongolia and Russia region; 2. Central Asia region; 3. West Asia and North Africa region;

4. Central and Eastern region; 5. South East region; 6. South Asia region; 7. Eastern Asia region.

(1) Mongolia and Russia

The energy structure, energy intensity and the Influence of international trade on GDP caused depressing effects, among which energy intensity played the greatest role. Per capita export goods and services trade has a significant role in increasing the EP_{cfec} , and contributed the most to the EP_{cfec} . The population intensity of productive land has shown a weak promotion effect, and its contribution gradually decreased.

(2) Central Asia

The energy structure, energy intensity and influence of international trade on GDP had a suppressing effect on the EP_{cfec} , among which the energy intensity showed the greatest inhibition. The per capita export goods and services trade caused fluctuations. The ΔEP_{EP} was -0.01 from 1994–1999, showing the inhibition effect. During the period from 2000–2004, it was transformed into a promoting factor. During the period of 2010–2014, the ΔEP_{EP} reached 0.04, and the rate of change was 1.88. The intensity of productive land population always showed a promotive effect, and the effect was greater than that of the per capita export goods and services trade volume.

(3) West Asia and North Africa

The energy structure, energy intensity, and the influence of international trade on GDP inhibited the EP_{cfec} in this region. The inhibitory effect of energy intensity gradually increased, and there was little change in the energy structure. The inhibitory function of YE increased at first and then weakened. The per capita value of the trade in goods and services and the population intensity of productive land played a catalytic role. The effect of the former was greater than that of the latter, and both indicators showed an upward trend over time.

(4) Central and Eastern

The energy structure, the Influence of international trade on GDP and the energy intensity had an inhibitory effect on the EP_{cfec} , with the effect is moving from small to large and increasing year by year. The per capita export goods and service trade volume, and the population intensity of productive land were both promoting effects. The former had a greater effect than the latter, and both indicators are increasing over time.

(5) Southeast

The influence of international trade on GDP had an inhibiting effect in this region. The energy structure has been transformed from inhibition to promotion, and the promotion shows an increasing trend. The energy intensity has been transformed from promotion to inhibition. The per capita exports of goods and services, the trade volume, and the population intensity of productive land still show a positive effect. The role of these factors has increased over time. The rapidly expanding population size of the region is the main reason for the promotion of the population intensity of productive land. The main reason for this is the consumption of non-renewable energy, which is the main reason for the transformation of the energy structure into a promoting effect.

(6) South Asia

Energy intensity always played an inhibitory role, and caused and increasingly strong effect year by year. The overall performance of the energy structure has increased, but the promotion effect has been relatively small, and showed a slight inhibitory effect during 2005–2009. The Influence of international trade on GDP is generally low. From 2010–2014, it began to show a small promoting effect and needs more attention. The per capita export goods and services trade volume, and the population intensity of productive land showed a promoting effect. The former is greater than the latter, and the promotion of both indicators showed an upward trend. The area is also dominated by

primary energy consumption, which is the most important reason that the overall performance of the energy structure is the driving force.

(7) Eastern Asia

The energy intensity had a suppressing effect on the EP_{cfec} and the overall trend of suppressing effect enhanced. The influence of international trade on GDP is generally suppressed. The energy structure, the per capita export goods and service trade, and the population intensity of productive land are all contributing factors, and the impact is enhanced year by year. Among them, the per capita export of goods and services trade has the greatest promotion effect, while the energy structure and the productivity of productive land tended to be more moderate, indicating that the energy structure adjustment and productive land protection measures in the region work to good effect.

The driving effects of the various factors in the seven sub-regions are shown in Table 7.

Factors	ES	EI	YE	EP	PH
Mongolia and Russia region	×	×	×	\checkmark	
Central Asia region	×	×	×	Ó	
West Asia and North Africa region	×	×	×		
Central and Eastern region	×	×	×		
South East region	0	0	×		
South Asia region	0	×	\bigcirc	\checkmark	
Eastern Asia region		×	×	\checkmark	

Note: $\sqrt{\text{represents that the factor contributes to the } EP_{cfec}$; × represents the factor having an inhibitory effect on the EP_{cfec} ; and \bigcirc is used to show the role of the factor changes over time.

4.4. Discussion

From the above research results, we can see that the ecological pressure of the carbon footprint in energy consumption is creating a serious challenge for the countries along the One Belt and One Road. Therefore, all the countries along the One Belt and One Road should pay more attention to the issue of ecological pressure, including setting clear emission reduction targets and emission reduction paths, in order to change the development model, realize green and low-carbon development, and to coordinate the balance between economic growth and environmental protection. In the future, with the aim of reducing the ecological pressure of the carbon footprint, governments should take green international trade and protecting the productive land as the primary methods, while continuing to reduce energy intensity and optimize the energy using structure (i.e., the clean energy occupying the most percentage of all the energy in the development). The green development of international trade requires all countries to strengthen cooperation and coordinate with each other, and fully consider the impact on other countries' environment in terms of export trade and capacity cooperation. These countries should fully consider their own resource endowments and regional advantages, choose a decompression route, focus on developing the domestic economy and international trade, reduce the ecological pressure of the carbon footprint, and provide new impetus for economic growth. For instance, the Mongolia and Russia region should continue to exert the inhibitory effect of the population intensity of the productive area, and Eastern Asia, South Asia, and Southeast Asia should further optimize the energy structure, increase its inhibitive influence, and reduce its effect promoting the EP_{cefc}. Economically underdeveloped regions, such as Central Asia, should learn from the experience of other countries by choosing green low-carbon industries as soon as possible and adopting a sustainable development model.

5. Conclusions

Utilizing data from the countries along the One Belt and One Road from 1994–2014, we measured the ecological pressure of the carbon footprint in energy consumption (EP_{cefc}), and revealed the

temporal and spatial dynamic changes EP_{cefc} . Furthermore, by using the LMDI method to decompose its driving factors, contributions to ecological pressure arising from the carbon footprint in energy consumption was analyzed for the entire region and in seven sub-regions of the One Belt and One Road.

In general, the value of the EP_{cefc} along the One Belt and One Road increased annually as a whole. From 1994–2014, the value of the EP_{cefc} of 56 countries along the One Belt and One Road increased annually. The growth rate increased at first and then decreased, which indicates that the overall EP_{cefc} of the countries along One Belt and One Road is still high. However, the growth trend has gradually slowed down. In terms of sub-regions, the region with the most ecological stress on the carbon footprint was found to be Southeast Asia, and the lowest was Central Asia. The proportion of carbon footprint ecological pressure in Central and Eastern Europe and Mongolia was gradually decreasing over the study period. The proportion in Southeast Asia, South Asia, and China was gradually increasing, and Central Asia remains basically unchanged. The per capita export of goods and services, and the population density on productive land contribute to ecological pressure on the carbon footprint. Energy structure, the influence of international trade on GDP, and energy intensity exerted an inhibitory effect on the ecological pressure of the carbon footprint. The former had a greater effect than the latter, while the energy structure has less effect, and positive and negative fluctuations. The effect of overall per capita export of goods and services, is greater than the population intensity of productive land.

From the perspective of the seven sub-regions, the effects of various factors on the EP_{cefc} are different, and the influence of each driving factor on different regions is also quite different. Energy intensity shows inhibition in all regions. During the period of 2010–2014, the inhibitory effects were as follows from small to large: Central Asia, Mongolia, Russia, South Asia, East Asia (China), Central and Eastern Europe, West Asia, North Africa and Southeast Asia. The energy structure had both inhibitory and promoting effects, but the effect was always small. Areas where it caused an inhibitory effect include Mongolia, Russia, Central Asia, West Asia, North Africa, and Central and Eastern Europe. Regions where it was a promoter include South Asia and East Asia. The impact of the Influence of international trade on GDP was generally inhibitory effect. The impact of the two factors of per capita exports of goods and services and the population intensity of productive land showed a promoting effect as a whole, and the effect of the former was greater than the latter. However, in the Mongolia–Russia region, the population intensity of productive land shows a slight increase.

This study mainly discusses issues regarding the ecological pressure of the carbon footprint in energy consumption, and other measurements of the ecological pressure of the carbon footprint from various angles, such as transportation and tourism. The conclusions of the paper can only be explained by the carbon emissions generated by energy consumption; however, this has its limitations. The regional division of land was mainly based on geography and not according to economic characteristics. The generality of the analysis conclusions is affected by this. Additionally, the productive land area conversion coefficient directly used the WWF's value to measure the carbon footprint in energy consumption without considering the difference in storage capacity of different types of productive land for carbon emissions.

Author Contributions: Q.S. established the research framework; Y.G. and F.M. jointly established the research model; C.W., B.W., X.W., and W.W. collected the data and carried out the result calculations; Q.S. provided the data acquisition channel; and Q.S. and Y.G. analyzed the results and wrote the paper together.

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