

Article

Spatiotemporal Pattern and Driving Forces of Arable Land-Use Intensity in China: Toward Sustainable Land Management Using Emergy Analysis

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Abstract: The level of arable land-use intensity has important impacts on food security and rural sustainable development. Using the emergy method, we investigate the spatial disparities and driving forces of arable land-use intensity in China from 1999 to 2008 at the national, regional and provincial levels. The empirical results show that chemical fertilizer was the largest component of agricultural inputs and that agricultural diesel oil recorded the highest growth rate. The degree of heterogeneities in arable land-use intensity in China showed a decreasing trend, which resulted mainly from the differences among the eastern, northeastern, central and western regions. The regional disparities in labor, pesticides and plastic sheeting decreased from 1999 to 2008. The per capita annual net incomes of household operations and the agricultural policies had a significant positive correlation with total inputs, fertilizer inputs, pesticide inputs and agricultural plastic sheeting. In addition, the nonagricultural population had a greater impact on agricultural plastic sheeting. Finally, we suggest that there is an urgent need to focus on the effects of chemical fertilizer and pesticide inputs on the ecological environment. Agricultural support policies should be introduced for the poor agricultural production provinces.

Keywords: arable land; land-use intensity; emergy; sustainable land management; food security

1. Introduction

Land use is an economic and social activity, wherein goods and services are obtained through the combined use of human labor and land resources [1–5]. Arable land is one of the necessary resources for human survival [6,7]. The level of arable land-use intensity has important impacts on food security and rural sustainable development. China suffers from a relative scarcity of arable land. The per capita arable land area was only 1.39 acres in 2006, less than 40% of the world average at the time. Additionally, accelerating urbanization and a growing population have drastically reduced China's arable land [8,9]. To maintain sustainable development in agriculture, it is imperative first to ensure the quantity and quality of arable land. The lack of incentives for improving the quality of arable land may pose more threats to food security than the reduction in arable land area [6,10]. Since the initiation of economic reforms in China, increasing arable land use and production intensity have played critical roles in meeting the growing demand for food [11,12]. Therefore, intensive use of arable land has become a vital strategy for alleviating the shortage of arable land and ensuring food security [13]. The structural characterization of arable land-use intensity reflects the internal structure of and trends in land-use intensity. In recent years, in addition to considering the indicators used to measure arable land-use intensity, researchers have explored the factors affecting its development by analyzing the structural features of land-use intensity [14]. The structural features of land-use intensity can also reveal the associated impacts on the surrounding ecological environment. Certain studies have indicated that these features will vary with economic development and that capital investment will be gradually replaced by labor input [15]. Thus, it is particularly significant to study the structural features, regional disparities and driving forces of arable land-use intensity in China [6,12,15,16].

In China's economically developed regions, there is evidence of an apparent decline in the labor force on China's arable land, a gradual decline in the proportion of capital investment in increasing production and a rapid growth in labor-saving inputs through capital investment [17]. In contrast, the economically underdeveloped areas show a higher degree of labor intensity and low capital intensity [17]. It is apparent that these studies analyzed the structural features of arable land-use intensity in China from the viewpoints of labor intensity and capital intensity. The structural features of arable land-use intensity should be further refined, such as in terms of fertilizer intensity or pesticide intensity, so that they specifically reflect the changing trends in intensive land use. In addition, past studies have not focused on the changing differences in these internal structures among various areas of the country. Using methods, such as monetary evaluation, to measure arable land-use intensity, previous studies on the structural features of land-use intensity have always considered prices of agricultural inputs to be a unified dimensionless quantity. However, the price method suffers from two main defects. The biggest shortcoming is that spatiotemporal disparities in the prices of labor and production are remarkably obvious, and there are sometimes large deviations from the real quantities of labor and material inputs [6]. Additionally, price is a quantitative measure that does not hold any information about the quality of the input/product. In contrast, energy evaluation offers insight into the quality of the inputs in agricultural production. Emergy is the available energy (exergy) of one kind that is used in transformations directly and indirectly to make a product or service [16]. Emergy accounts for and, in effect, measures quality differences between forms of energy [16]. Furthermore, emergy can be used to evaluate the indirect and direct work of nature and does not suffer from the

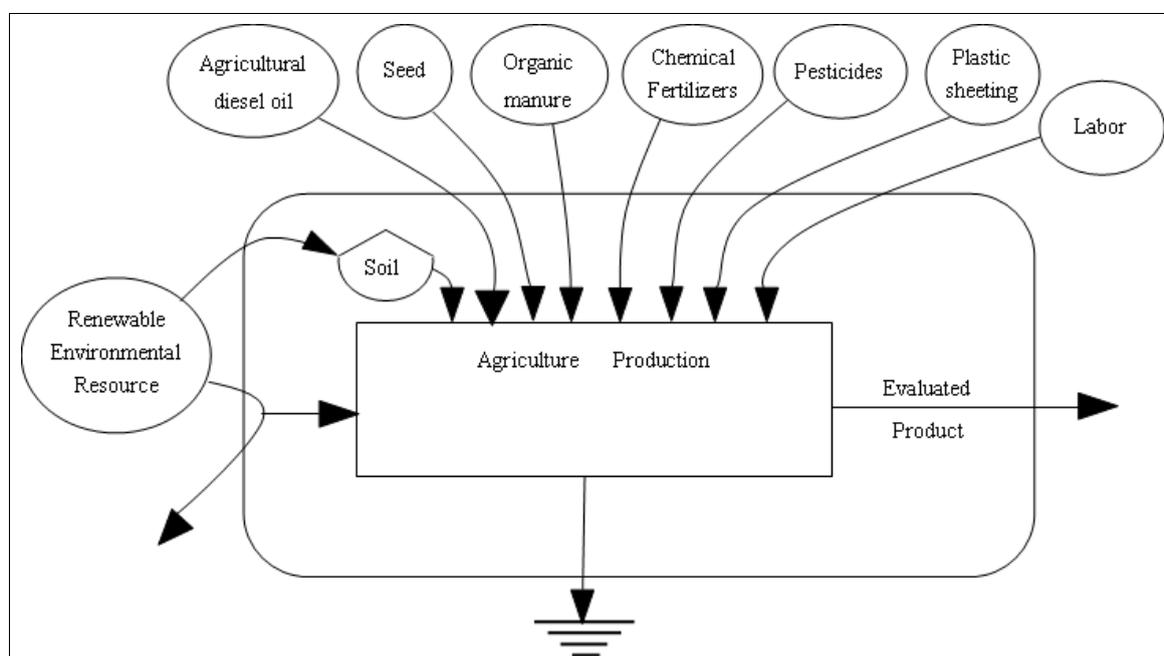
problem of spatiotemporal differences [18,19]. It can eliminate the defects of the monetary evaluation by evaluating the arable land-use intensity from an eco-centric perspective, thereby excluding any anthropocentric views on agricultural production [20]. In recent years, scholars have begun to apply the theory of emergy to analyze arable land eco-economic systems [21,22]. Thus, using emergy analysis to investigate the quality of the land-use intensity, this study analyzes the spatiotemporal patterns, regional disparities and driving factors of arable land-use intensity in China. The results will contribute toward the sustainable management of arable land in China.

2. Materials and Methods

2.1. System Investigated

In this study, we will investigate the agricultural production system in China during the time period of 1999–2008 to examine the spatiotemporal pattern and driving forces of arable land-use intensity. The agricultural production system in China is a semi-natural ecosystem under human control. The main input items of the agricultural production system in China include chemical fertilizers, pesticides, plastic sheeting, organic fertilizers, seeds and labor. In particular, modern agriculture in China focuses on the input of industrial auxiliary goods and is not aware of the values of environmental resources. To evaluate a system, first, a system diagram is drawn to organize the evaluation and account for all inputs and outflows [16]. The energy diagram of the agricultural production system in China can enable us to understand more easily the basic structure and the internal and external relationships between the ecological stream flows of the agricultural production system. According to related studies and the characteristics of agricultural production in China [23–26], we have drawn the energy diagram of inputs for agricultural production in China (Figure 1).

Figure 1. Energy diagram of inputs for agricultural production in China.



According to the level of economic development and the status of national policy support, China is generally divided into four major regions, *i.e.*, the developed eastern region (Eastern region), the rising middle region (Central region), the developing western region (Western region) and the revitalizing northeast district (Northeastern region). When measuring the spatial disparity, this study considers the above four regions in China.

2.2. Data

Based on the accessibility of data, this study divides the input items of arable land in China into the five categories, including renewable resources (sun, rain and wind), non-renewable resources (soil), industrial auxiliary goods (chemical fertilizers, pesticides, plastic sheeting, *etc.*), organic goods (organic fertilizers, seeds) and others (labor, *etc.*). The data pertaining to the nitrogen, phosphate, potash and compound fertilizers, pesticides, plastic sheeting and diesel are sourced from the China Rural Statistical Yearbook. The data on the labor force and the area of arable land are sourced from the China Statistics Yearbook of Labor and the China Statistical Yearbook of Land and Resources, respectively. The data on the amount of labor per hectare are derived from the 2012 National Agricultural Costs and Returns Assembly. The time scale of the data spans 10 years (1999–2008), and spatially, we consider China's 31 provinces (municipalities and autonomous regions). Hong Kong, Macao and Taiwan have been excluded from the scope of this study, due to data unavailability.

2.3. Methods

2.3.1. Measuring Emergy for Input Items

The “solar value” can be used to measure a variety of energy values in practical applications, since solar energy can be converted into any other form of energy. In other words, the amount of solar energy consumed in the creation of any resource, product or service is the solar value of that entity [18,27]. In this study, we use the concept of emergy as a unified measure for six inputs to arable land per hectare, and we analyze China's emergy intensity of arable land use and its structural characteristics. It should be noted that in addition to the abovementioned five inputs/investments, investment in arable land includes herbicides and other services. However, taken together, the proportions of the five kinds of inputs selected in this study are relatively larger and represent the overall trend of investment in arable land. Emergy analysis is an environmental accounting approach and is able to account for free local resources. In particular, it enables the unification of flows in various forms. Unit emergy values (UEVs) are computed based on the emergy required to generate one unit of output from a process [16]. The unit emergy values (UEVs) of the sun, rain, wind, soil, labor, agricultural diesel, N fertilizers, P fertilizers, K fertilizers, pesticides and plastic sheeting are sourced from related studies [26,27]. The translated data used in this study mainly refer to the relevant literature [27,28].

The fundamental formula for the calculation of the emergy of the inputs is as follows:

$$Em (seJ) = flow (J, g, \$) \times UEVs (seJ/J, seJ/g, seJ/\$) \quad (1)$$

The formula for labor emergy (*LE*) per hectare of arable land is as follows:

$$LE = T_l \times UEV_l \times N_l \times D_l \quad (2)$$

where LE is the labor energy per unit area of arable land, T_l is the conversion rate of labor energy, UEV_l is the UEV of labor, N_l is the quantity of labor per hectare of arable land and D_l is the average labor days per hectare of arable land.

2.3.2. Measuring Regional Disparity

Theil's T statistic was introduced by Theil to measure local income inequality [29]. The regional disparities in arable land-use intensity and its structure can be measured using Theil's T statistic in this study, because Theil's T statistic effectively uses group data and allows researchers to analyze inequalities within groups and between components of a group. Therefore, Theil's T statistic can reveal the source of inequality and its values. The following equation calculates Theil's T statistic for differences in arable land-use intensity at the provincial level:

$$T = \sum_{p=1}^n \left[\left(\frac{1}{n} \right) \times \left(\frac{y_p}{\mu_y} \right) \times \ln \left(\frac{y_p}{\mu_y} \right) \right] \quad (3)$$

where n is the number of provinces, y_p is the arable land-use intensity of province p and μ_y is the national average of arable land-use intensity. If every province has exactly the same arable land-use intensity, T will be zero; this represents perfect equality and is the minimum value of T . If one province is accountable for all arable land-use intensity, T will equal $\ln n$; this represents the utmost inequality and is the maximum value of T . The value of T is a monotonically increasing measure of inequality in the distribution of arable land-use intensity, bounded by $T \in [0, \ln n]$.

If provinces in China are classified into four types of areas, then the additive decomposability of Theil's T statistic can be estimated for China as the sum of two components: the elements between regions (T_{br}) and the elements within a region (T_{wr}).

$$T = T_{br} + T_{wr} \quad (4)$$

The index for between-region elements (T_{br}) can be used as a lower bound for the country's Theil's T statistic. The between-region component of Theil's T statistic can be written as:

$$T_{br} = \sum_{i=1}^m \left[\left(\frac{p_i}{P} \right) \times \left(\frac{y_i}{\mu} \right) \times \ln \left(\frac{y_i}{\mu} \right) \right] \quad (5)$$

where m is the number of regions, p_i is the number of the provinces in region i , P is the total number of provinces, y_i is the average of arable land-use intensity in region i and μ is the average arable land-use intensity in China.

2.3.3. Econometric Model

In this paper, we establish an empirical model to explore the factors influencing arable land-use energy intensity in the various provinces of China. The econometric model is as follows:

$$Y_{it} = \alpha_{it} + \beta_1 INC_{it} + \beta_2 IND_{it} + \beta_3 POP_{it} + \beta_4 POL_{it} + u_{it} \quad (6)$$

where i ($i = 1, \dots, 31$) and t ($t = 1999, \dots, 2008$) represent the province, i , and year t , respectively. The term, α_{it} , is a constant, and u_{it} is the random error term. Y_{it} is the energy intensity of land-use, an explanatory variable. INC_{it} is the net rural household income per capita. IND_{it} is the proportion of

nonagricultural industry. POP_{it} is the proportion of nonagricultural population, and POL_{it} represents the agricultural support policy.

According to the hypothesis of economic man, an increase in the net income of rural households per capita drives farmers to increase agricultural production and also determines the farmers' investment capacities in arable land. To remove the heteroscedasticity in the data to the extent possible, we take the log of INC_{it} . In general, positive changes in agricultural output in one year lead to adjustments in agricultural inputs in the following year. Therefore, we set INC_{it} with a time lag effect of one year.

The development of the secondary and tertiary industries has driven improvements in production equipment, capital investment and the technology for agricultural production, and the increasing demand for agricultural products has promoted agricultural intensification. The rising proportion of the nonagricultural industry has had a tremendous positive impact on the intensive use of arable land.

An important characterization of the urbanization process is the reduction in the proportion of the nonagricultural population. Thus, in this study, we choose the proportion of the nonagricultural population to characterize the level of urbanization. The rising proportion of nonagricultural population drives the demand for agricultural products and, therefore, also spurs improvements in agricultural management. In such a case, farmers would increase investments in farmland and increase the use intensity of arable land, thus producing more agricultural products and reaping greater economic benefits.

In China, agricultural support policies have had a significant impact on agricultural development. Since 2004, the Chinese government has guided this development in rural areas through *No. 1 Central Document*; for nine consecutive years, the government has maintained concerted efforts to support and promote agriculture, thus improving the overall production capacity of the sector. Therefore, POL_{it} is zero before the year 2004 and one otherwise.

This study considers the panel data of 31 provinces from 1999 to 2008. We conducted a regression analysis of arable land-use intensity with Eviews7.0 software. First, we conducted the F-test analysis for the abovementioned econometric model, to ascertain whether it supports the variable intercept model. Then, we use the Hausman test to determine whether the fixed effects model or the random effects model is more suitable. The results of the Hausman test showed that the model supports the fixed effects model. Finally, the least squares dummy variable method (LSDV) was used to estimate the fixed effects model.

3. Results and Discussion

3.1. Change in Emery Intensity of Arable Land-Use at the Country Level

According to Formulas (1) and (2), we can obtain the emery flows of agricultural production in China during the time period of 1999–2008. Considering the limited space, we only list the emery flow of China in 1999 (see Table 1).

Table 1. Emergy flows of agricultural production of China in 1999.

Item	Flow	Flow units	UEV	UEV units	Emergy	Emergy units
Renewable environmental flows:						
1	Sun	3.56×10^{22}	J	1.00	sej/J	3.56×10^{22} sej
2	Rain	19.12×10^{18}	J	1.82×10^4	sej/J	3.48×10^{22} sej
3	Wind	22.84×10^{18}	J	1.58×10^3	sej/J	3.61×10^{22} sej
Non-renewable environmental flows:						
4	Soil	1.68×10^{18}	J	1.70×10^5	sej/J	28.63×10^{22} sej
Non-renewable industrial flows:						
5	N fertilizer	2180.90×10^{10}	g	3.80×10^9	sej/g	8.29×10^{22} sej
6	P fertilizer	697.80×10^{10}	g	3.90×10^9	sej/g	2.72×10^{22} sej
7	K fertilizer	365.60×10^{10}	g	1.10×10^9	sej/g	0.40×10^{22} sej
8	Chemical manure	880.00×10^{18}	g	2.80×10^9	sej/g	2.46×10^{22} sej
9	Pesticides	132.16×10^{10}	g	1.62×10^9	sej/g	0.21×10^{22} sej
10	Plastic sheeting	125.87×10^{10}	g	3.80×10^8	sej/g	0.05×10^{22} sej
11	Agricultural diesel oil	1354.30×10^{10}	g	6.60×10^4	sej/J	3.87×10^{22} sej
Renewable organic flows:						
12	Organic manure	192.98×10^{13}	g	4.54×10^6	sej/g	0.88×10^{22} sej
13	Seed	133.93×10^{16}	g	3.36×10^6	sej/J	4.50×10^{22} sej
Other flows:						
14	Labor	3.58×10^8	J	3.80×10^5	sej/J	5.51×10^{22} sej
Total *						47.23×10^{22} sej

Notes: UEV = unit energy value; * To avoid double counting, we consider only the largest of renewable environmental flows (sun, wind, rain).

According to the emergy flows of agricultural production in China, we obtain the emergy intensity of arable land use during the time period of 1999–2008 (see Table 2). From Table 2, we can see that the emergy intensity of arable land use increased from 47.23×10^{14} sej/ha in 1999 to 50.10×10^{14} sej/ha in 2008, an increase of 6.07% (see Table 2). The internal structures of arable land-use intensity reveal that labor input declined year-by-year, and agricultural diesel, fertilizers, pesticides and plastic sheeting exhibited increasing trends from 1999 to 2008. The largest increase was recorded for chemical manure.

From Table 2, we can see that for all the years studied, the emergy of fertilizers continued to represent the largest input factor in China's agricultural sector. Although the proportions of fertilizer inputs fluctuated, they always exceeded 52% of the total input factors. The emergy of fertilizers input per unit area of arable land gradually increased through time, increasing by 29.26% from 1999 to 2008. Agricultural diesel oil increased by 47.67%, thus becoming the fastest-growing input factor of agricultural growth in China during this period.

Table 2. Input structures per unit of arable land in China from 1999 to 2008.

Input items	1999	2000	2001	2002	2003	2004	2005	2006	2007	2008	
Renewable environmental resources	2.69	2.65	2.70	2.68	2.50	2.63	2.64	2.75	2.99	2.99	
Non-renewable environmental resources	22.16	22.13	22.14	22.14	22.13	22.16	22.12	22.12	22.13	22.13	
Industrial Goods	Chemical manure	10.74	10.83	11.10	11.40	11.79	12.41	12.75	13.17	13.59	13.88
	Pesticides	0.17	0.18	0.16	0.17	0.17	0.24	0.27	0.20	0.22	0.22
	Plastic sheeting	0.04	0.04	0.04	0.05	0.05	0.05	0.05	0.06	0.06	0.06
	Agricultural diesel oil	3.00	3.13	3.33	3.42	3.65	4.25	4.45	4.51	4.74	4.43
Organic materials	Organic manure	0.68	0.69	0.63	0.64	0.65	0.65	0.66	0.63	0.63	0.63
	Seed	3.48	3.49	3.49	3.49	3.48	3.48	3.48	3.49	3.48	3.48
Others	Labor	4.26	4.16	4.05	4.02	3.92	3.41	3.16	2.74	2.48	2.27
Total *	47.23	47.29	47.63	48.00	48.35	49.28	49.57	49.68	50.33	50.10	

Note: * To avoid double counting, we consider only the largest of renewable environmental flows (sun, wind, rain).

Table 2 also shows that the emery of labor decreased from 4.26×10^{14} sej/ha in 1999 to 2.27×10^{14} sej/ha in 2008, a reduction of 46.65%. This reduction resulted from the modernization of agriculture and, in particular, agricultural mechanization. It is also due to the increasing opportunity cost of farming labor. The proportion of agricultural diesel is one of the important indicators of agricultural mechanization. The emery of investment in agricultural diesel increased from 3.00×10^{14} sej/ha in 1999 to 4.43×10^{14} sej/ha in 2008 (Table 2), which means that the level of agricultural mechanization in China increased from 1999 to 2008.

3.2. Change in Emery Intensity of Arable Land-Use at the Regional Level

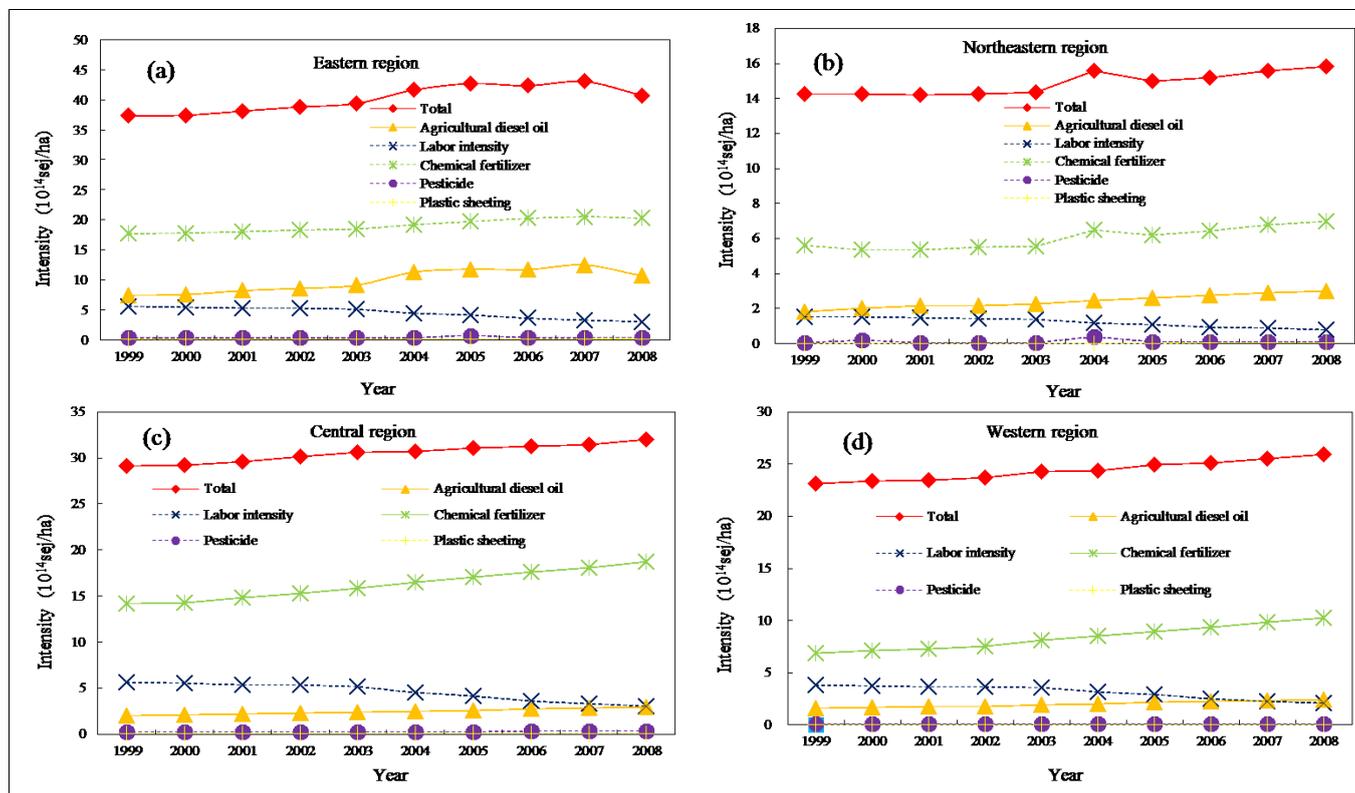
Figure 2 depicts the changes in land-use intensity and its structural features in the four regions of China from 1999 to 2008. In contrast to the northeast and western regions, land-use intensity in the eastern and central regions exceeded the national average from 1999 to 2008.

Figure 2 shows that the degree of arable land-use intensity at the regional level could be arranged in decreasing order as follows: the eastern region, the central region, the western region and the northeastern region. The difference in arable land-use intensity between the eastern and northeastern regions and the central and western regions is significant (Figure 2). In the past decade, the gaps in arable land-use intensity between the eastern and northeastern regions, the eastern and central regions, and the eastern and western regions increased by 1.69%, 10.56% and 9.69%, respectively.

The emergies of the five input items per unit area of arable land in the eastern region (including agricultural diesel, fertilizers, pesticides and plastic sheeting) were higher than the national level. The corresponding characteristics of the western region were diametrically opposite. The investment in diesel in the central region was lower than that of the national level, while for the investment in plastic sheeting flats, with the national average. The emery of fertilizer inputs per unit of arable land were the largest in the eastern region; they increased from 17.67×10^{14} sej/ha in 1999 to 20.28×10^{14} sej/ha

in 2008, an increase of 14.77%. The rapid increase in the energy intensity of arable land-use in the northeast region can be attributed to the rapid growth in chemical manure from 1999–2008.

Figure 2. Changes in energy intensity of arable land-use and its internal structures in China from 1999 to 2008.



Although fertilizers form the largest component of agricultural inputs across all four regions, the amounts and percentage changes vary for each region. In terms of fertilizer inputs, the eastern region recorded the largest amount and the slowest growth, while the western region showed small input amounts and the largest increase, with the middle and northeastern regions falling somewhere in between. The trend of continued high dependence on fertilizer inputs was contained to some degree in the eastern region, but it could still be observed in the central and western regions.

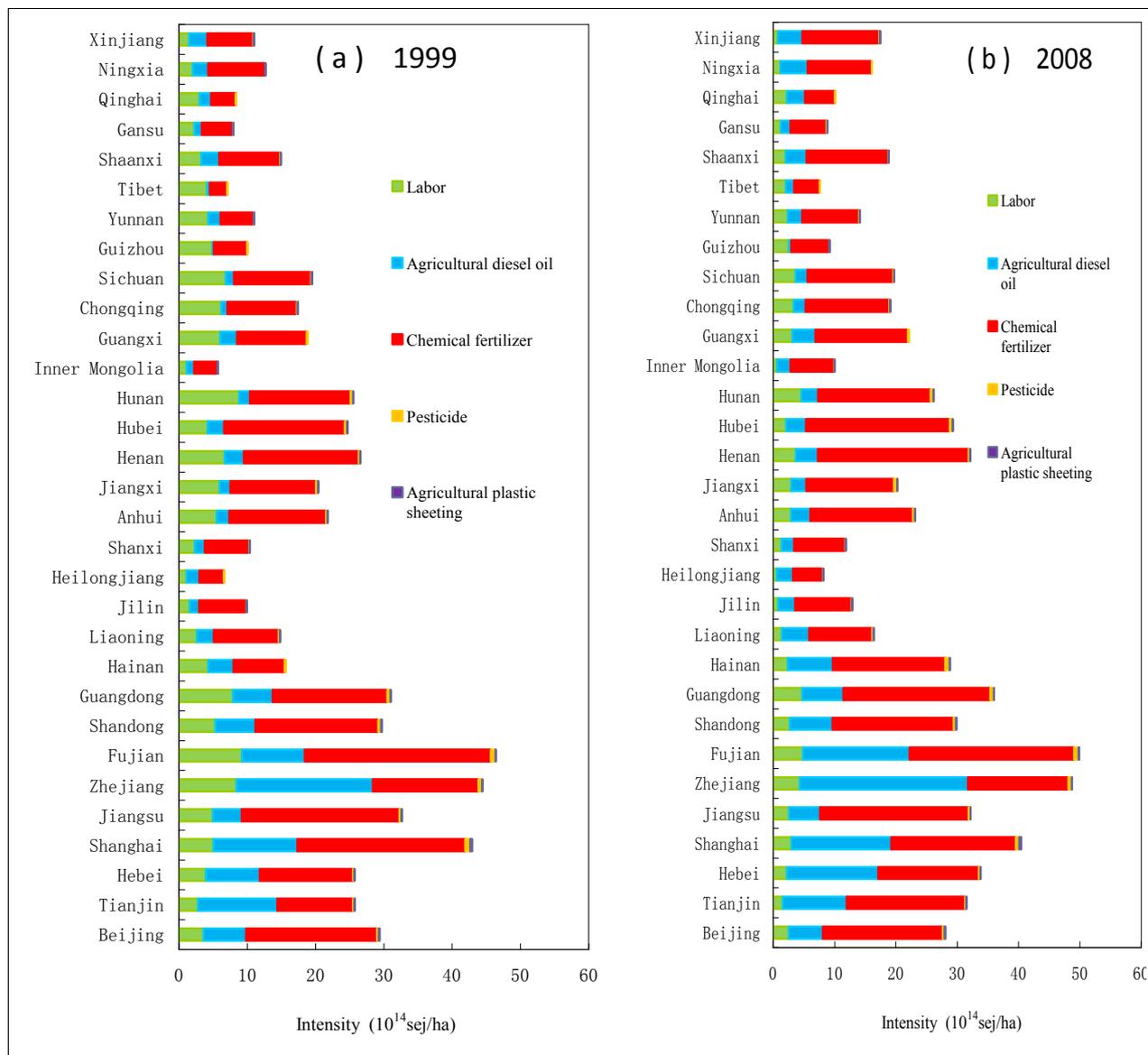
The past 10 years have witnessed inevitable large-scale rural labor force migration, which plays a major role in China's social and economic development. The northeast region saw the largest reduction in labor (49.03%), followed by the eastern (46.32%) and central regions (46.64%). The labor decline in the western region was relatively small, but it reached 45.97%. In terms of labor proportion, the proportion of energy of labor in the western region decreased the most from 1999 to 2008 with a reduction of 17.47%. In 2008, the labor force was no longer one of the largest input factors in the agriculture of the eastern, northeastern and western regions.

3.3. Change in Arable Land-Use Intensity at the Provincial Level

Figure 3 shows the structural features of the energy intensity of arable land use at the provincial level in 1999 and 2008. With the exception of the small decrease in Shanghai, the use intensities of

arable land for the remaining 30 provinces of China increase. Figure 3 shows that provinces with a low level of land-use intensity were mainly concentrated in the western and northeastern regions.

Figure 3. Internal structures of the emergy intensity of arable land-use at the provincial level in China in 1999 and 2008.



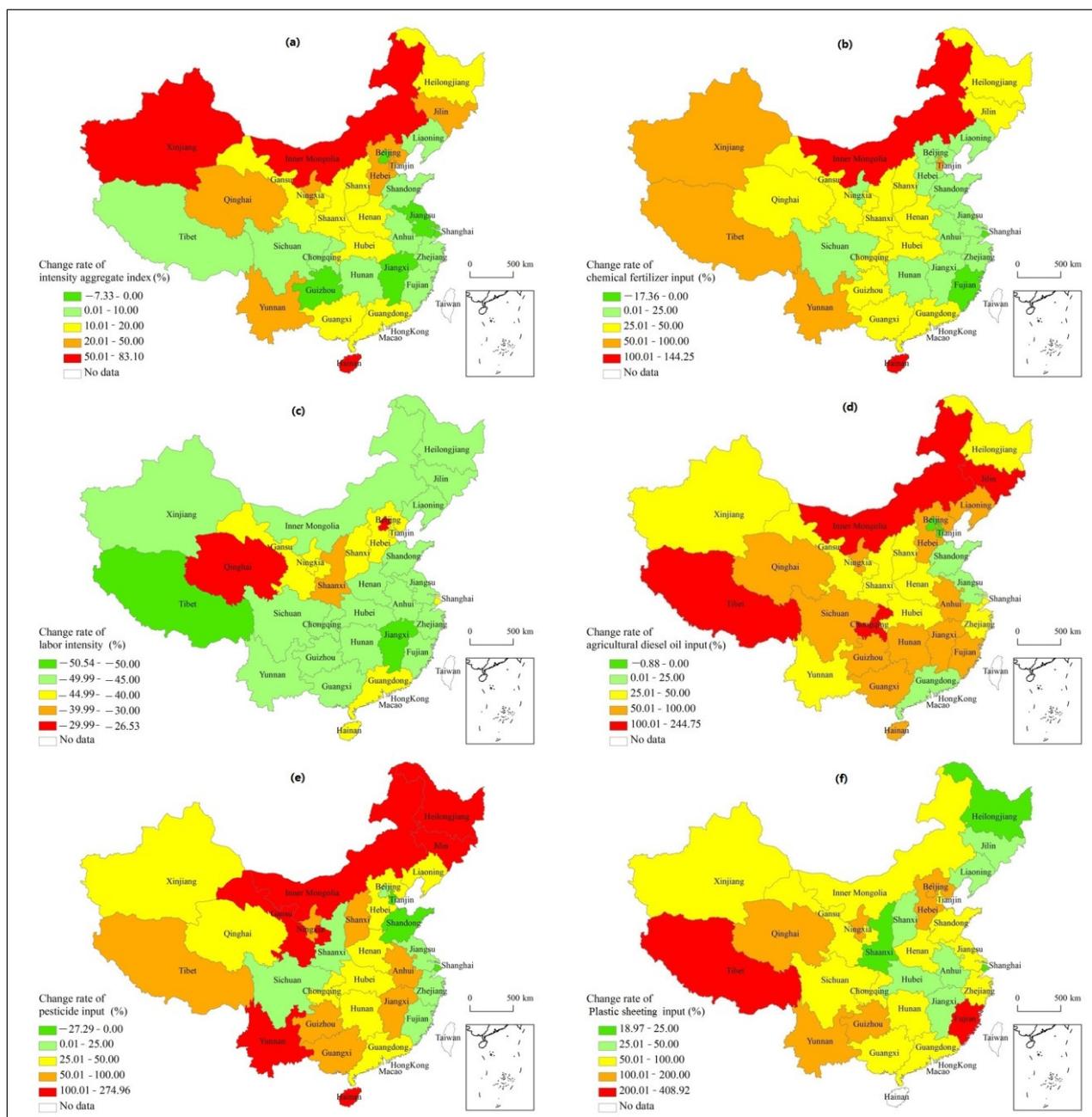
The land-use intensities of the provinces of the western and northeastern regions were low, while the growth rates of the land-use intensity increased (see Figure 4). Notably, the land-use intensity in Inner Mongolia recorded the largest increase (95%) of all provinces.

Fertilizers formed the largest component of agricultural inputs, with the proportion exceeding 50% in most of the provinces. There were enormous gaps in fertilizer input per unit area among different provinces. For instance, the gaps in the emergy of fertilizer input per unit area between the Fujian and Tibet provinces in 1999 and 2008 were 24.72×10^{14} sej/ha and 22.63×10^{14} sej/ha, respectively. The amount of fertilizer input per unit area in Inner Mongolia and Fujian provinces more than doubled in the past 10 years. The amounts of chemical fertilizer per unit area in Fujian province and Shanghai

recorded negative growth rates from 1999 to 2008. The increase in the amounts of fertilizer inputs showed an obvious spatial gradient, with the increases declining gradually from west to east.

Labor input per unit area decreased to varying degrees, ranging from 27 to 51%. The input amounts of pesticides and plastic sheeting were much lower than those of chemical fertilizer, diesel and labor. The provinces with higher pesticide inputs were mainly concentrated in the eastern and central regions.

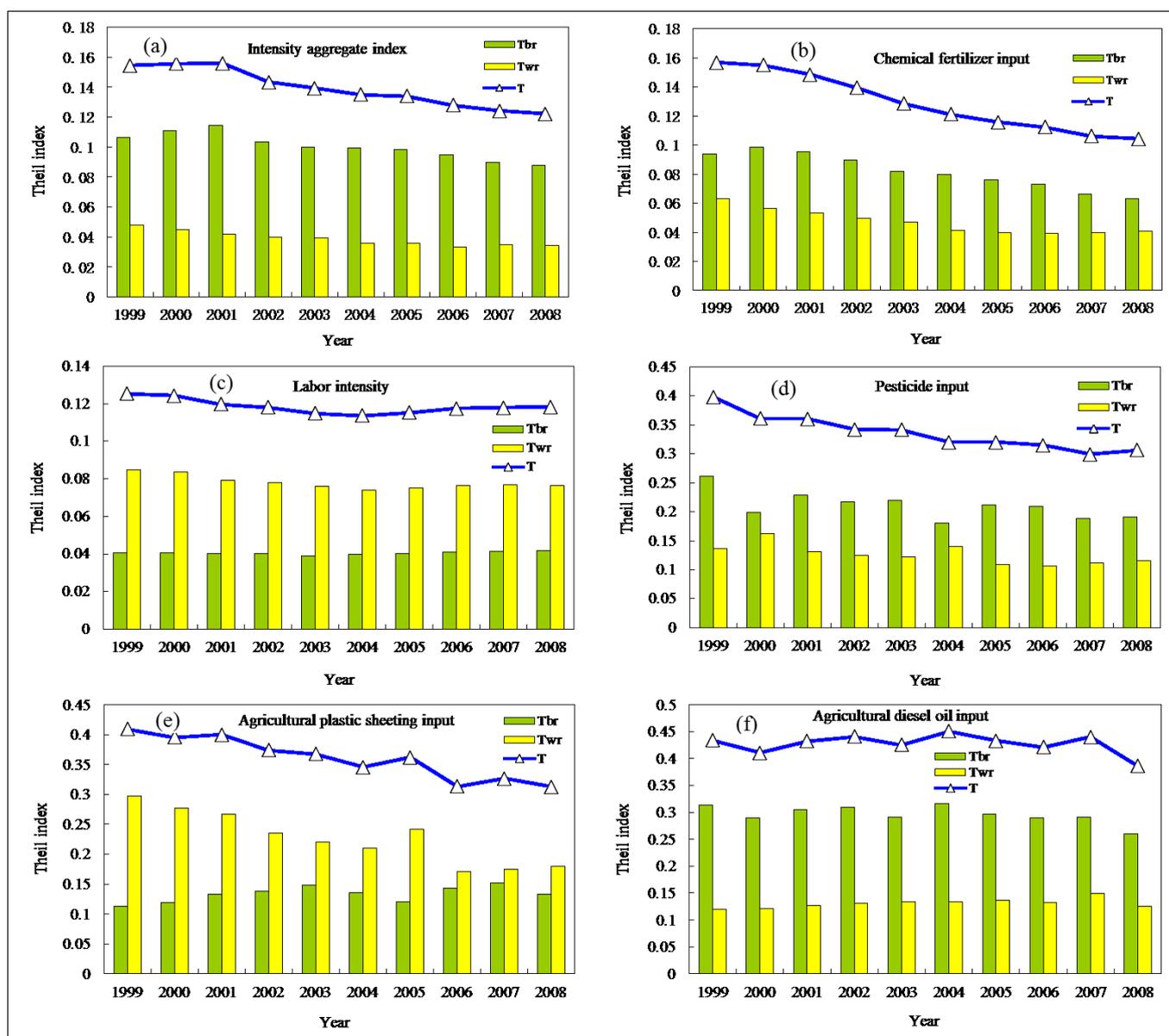
Figure 4. Degrees of change rates in the energy intensity of arable land-use and their internal structures for each province in China from 1999 to 2008.



3.4. Regional Disparities in Arable Land-Use Intensity

The Theil index of arable land-use intensity and its internal structure (including inputs for labor, diesel, fertilizers, pesticides and plastic sheeting) from 1999 to 2008 are shown in Figure 5. It shows a decreasing trend in the degree of arable land-use intensity for all inputs. The disparities in arable land-use intensities arise mainly from the differences between the eastern, northeastern, central and western regions.

Figure 5. Theil indexes of arable land-use intensity and their internal structures in China from 1999 to 2008.



From the viewpoint of the internal structure of arable land-use intensity in China, the highest regional disparities were recorded for plastic sheeting, pesticides and diesel per unit area. Their Theil indexes ranged from 0.35 to 0.45 (Figure 5). Figure 5 also shows that the Theil index of labor input remained steady at approximately 0.12 from 1999 to 2008, which means that the disparities in labor

inputs were at a minimum. The disparities in labor and plastic sheeting per unit area of arable land arose mainly from the internal differences among the four regions.

3.5. Driving Forces of Arable Land-Use Intensity

Table 3 shows the results of the econometric models for arable land-use intensity and their internal structures. The three models pass the significant F-test. The results support the variable intercept model. The results of the Hausman test show that the model supports the fixed effects model. The adjusted *R*-squared indicates that the three econometric regression models of arable land-use intensity and their internal components fit the given context better.

Table 3. Results of the econometric models for arable land-use intensity and its internal structure.

Variable	Dependent Variable					
	Intensity aggregate index	Chemical fertilizer	Agricultural diesel oil	Labor intensity	Agricultural plastic sheeting	Pesticide
<i>c</i>	1.2768 *** (4.8714)	−1.1316 *** (−3.3932)	−5.8807 *** (−12.3908)	7.2038 *** (23.2273)	−10.8734 *** (−8.8441)	−5.8275 *** (−6.1582)
<i>INC</i>	0.1541 *** (4.4650)	0.2634 *** (6.0001)		−0.5758 *** (−14.0986)	0.2734 * (1.6891)	0.3024 ** (2.4267)
<i>IND</i>	0.0054 * (2.5810)	0.0191 *** (7.2676)	0.0380 *** (10.1123)	−0.0169 *** (−6.8804)	0.0536 *** (5.5068)	0.0199 *** (2.6598)
<i>POP</i>	0.0036 * (2.2761)			−0.0098 *** (−5.2302)	0.0349 *** (4.6873)	
<i>POL</i>	0.0268 * (1.9964)	0.0702 *** (4.1041)	0.1265 *** (5.1994)	−0.1284 *** (−8.0744)		0.1124 ** (2.3177)
R-squared	0.9863	0.9822	0.9861	0.9857	0.9104	0.9521
Adjusted R-squared	0.9845	0.9800	0.9843	0.9839	0.8992	0.9462
Prob (F-statistic)	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Durbin-Watson stat	0.7065	0.6765	0.9868	0.6624	0.8262	1.9278

Notes: (1) *INC* represents the net income of rural households per capita; *IND* represents the proportion of the nonagricultural industry; *POP* represents the proportion of the nonagricultural population; and *POL* represents the agricultural policy; (2) The numbers within parentheses are the values of the *t*-statistic; (3) * $p < 0.1$; ** $p < 0.05$; *** $p < 0.01$.

There was a significantly positive relationship between the net income of rural household per capita and arable land-use intensity (Table 3). Similarly, the net income of rural household per capita played a positive role in the fertilizer inputs, pesticide inputs and agricultural plastic sheeting. This is mainly because the farmers are the main users of arable land and the net income of rural households per capita is directly related to the investment capacity in arable land. The higher the net income of rural households per capita, the more money farmers will invest in arable land. This result effectively confirms the hypothesis of economic man.

Table 3 shows a significant positive correlation between the proportion of nonagricultural industry and total investment and fertilizer inputs per unit of arable land. Increasing the proportion of secondary

and tertiary industries can contribute advanced production elements, including capital, technology and equipment, to arable land use. It can also increase the demand for agricultural products and prompt farmers to increase the core agricultural inputs (such as fertilizer) to boost production.

Table 3 also shows a significant positive relationship between the proportion of the nonagricultural population and agricultural plastic sheeting per unit of arable land. Assuming that all the other factors remain constant, for every 1% increase in the proportion of nonagricultural population, agricultural plastic sheeting per unit of arable land will increase by 0.035%. China is witnessing rapid urbanization, and thus, investments in agricultural plastic sheeting per unit of arable land will continue to increase, which is conducive to improvements in agricultural infrastructure.

Moreover, Table 3 indicates that the estimated coefficient of agricultural policy is positive, which means that agricultural policies have positive impacts on total input and fertilizer input. This result confirms the hypothesis of the econometric model. Therefore, the implementation of the agricultural policy plays a highly influential role in improvements of land-use intensity in China. The cause of the small differences in the inputs of fertilizers is due to the influence of agricultural policy.

Due to the object of this study and the accessibility of data, this study only considers five input items of arable land, including renewable resources, non-renewable resources, industrial auxiliary goods, organic goods and labor. Especially in terms of import services, only labor is considered. Moreover, this paper did not analyze the energy of arable land output. In a subsequent research project, we will improve the deficiencies of this study. The analysis of the energy of arable land output, certain ratios and indices (including the energy yield ratio (EYR), the environmental loading ratio (ELR) and the energy sustainability index (ESI)) should be performed to evaluate the global performance of the agricultural production process in future studies.

4. Conclusions

The energy intensity of arable land use in China showed a constantly increasing trend from 1999–2008. The provinces with high intensities of arable land use were mainly distributed in the eastern and central regions, and they recorded small increases in intensities. On the contrary, provinces with low intensities of arable land were mainly distributed in the western and northeastern regions, and these regions recorded larger increases in intensities.

An analysis of the internal structure of the energy intensity of arable land use showed that fertilizers were the largest input item in the land-use process throughout the study period. While this trend, which is highly dependent on chemical fertilizer inputs, has slowed to some extent in the eastern and northeastern regions, it grew in the central and western regions during the study period.

The regional differences in the energy intensities of arable land use from 1999 to 2008 and their internal structures exhibited a narrowing trend. The results showed large regional differences in the inputs of plastic sheeting, pesticides and diesel oil per unit of arable land. Conversely, the regional differences in the inputs of chemical fertilizers were small.

The per capita annual net income of rural household operations and the agricultural policies correlated significantly and positively with total inputs, fertilizer inputs, pesticide inputs and agricultural plastic sheeting per unit of arable land. The proportion of the nonagricultural population

had a greater impact on the agricultural plastic sheeting. The results of this study effectively support the hypothesis of economic man.

Today, chemical fertilizer and agricultural diesel oil in agriculture are the two main input components in arable land use in China. Therefore, it is imperative to investigate the effects of fertilizer application on the surrounding ecological environment, especially in the main grain-producing provinces of the eastern, central and western regions. There is also an urgent need to focus on the effects of chemical fertilizer and pesticide inputs on the ecological environment. The agricultural policy, therefore, should be amended to benefit the poor agricultural production provinces, especially the 13 major grain-producing areas.

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Author Contributions

Hualin Xie and Jinlang Zou had the original idea for the study. Jinlang Zou and Hailing Jiang was responsible for data collecting. Hualin Xie, Jinlang Zou, Ning Zhang and Yongrok Choi carried out the analyses. All the authors drafted the manuscript, and approved the final one.

Conflicts of Interest

The authors declare no conflict of interest.

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