

Article

Urban Land Use Efficiency and Coordination in China

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Abstract: Due to the focused pursuit of economic growth in the process of the large-scale urban development of China, the phenomena of low land use efficiency and discordance of land use induce unwanted economic, social, and environmental costs. This paper presents a comprehensive study of urban land use efficiency and of the degree of land use coordination of 33 cities in China, using theoretical analysis, data envelopment analysis, principal component analysis, the coordination coefficient method, and four-quadrant analysis. The findings of this study suggest a gradually increasing proportion of land use efficiency from eastern to central and western regions of China, coinciding with China's pattern of socioeconomic development. No correlation was found between high levels of urban land use efficiency and the degree of land use coordination; however, a significant correlation was found between low land use efficiency and low degrees of land use coordination. Rational land use planning and policy design can effectively improve both urban land use efficiency and coordination.

Keywords: urban land use; land use efficiency; land use coordination; correlation

1. Introduction

Ever since the onset of urban development, numerous problems have emerged that can be attributed to limited land resources. These problems are traffic congestion, insufficient water and electricity supplies, deterioration of living environment and interpersonal relationships, and reduced social stability [1–3]. Improving urban land use efficiency is commonly recognized to be of paramount importance in balancing the protection of farmland and the accommodation of the socio-economic development. This is particularly true for a country like China with its intense human–land relationship and rapid ongoing urbanization [4,5]. Careful evaluation of land use capabilities is required to deal with the issues that cities currently face within the limited land area and to provide a reliable basis for the improved use of urban land [6,7]. Land use efficiency is an important indicator for the level of urban land use [8]; however, the complexity of cities should not be overlooked. In particular, the coexistence and coordination of economic, social, and environmental systems within cities should be taken into account [9–11]. Assessing how these systems are being coordinated in light of the pursuit of land use efficiency can further provide insight into urban land use [12].

Urban land use has been investigated from many theoretical perspectives, including ecology, regional economic, social behavior, and political economics [13,14]. In recent years, scholars from various countries have conducted extensive research into urban land use. Macedo [15], Kironde [16], Terry and Raoul [17], and Jessica [18] focused on urban land policies, Pauleit et al. [19], Mouri and Aisaki [20], and Mirzaei et al. [21] studied the impact of urban land use on the environment, while Lin and Ho [22], and Sanjiv and Venkatesh [23] examined the urban land use status of different countries and regions. After studying the externalities of land use [24,25] and issues related to urban expansion, Irwin and Bockstael [26] and other scholars indicated the existence of a “dangerous” mode of land

development and predicted negative effects from urban land expansion. Pouriyeh et al. [27] conducted a case study of Yazd city to explore the impact of ecological factors, including the intensive use of land in the process of urban development. Herold et al. [28] studied the application of the spatial measurement method in the process of urban land use change. This method is widely utilized for urban land use measurements, and Aguilera et al. have studied how this method can be applied to assess urban land use [28–30].

Studies in the U.S. have measured land use efficiency in terms of developmental density, population density, and/or employment density [31,32]. However, in China, researchers have developed indicators to both measure and monitor land use efficiency. Key indices include investment intensity and economic output per land unit [4]. Fang [33] argued that objective and effective evaluation of land use efficiency requires the establishment of a scientific and effective land use evaluation indexing system, and presented principles as well as additional ideas for building such an indexing system. Wang et al. [34] used a national survey to analyze institutional setups for rural residential land use, thus assessing the effectiveness of existing regulations, and to evaluate the efficiencies of rural residential land use. Xie [35] analyzed spatial differences of urban industrial land use efficiencies in the six main economic zones of China based on the slacks-based measure (SBM) model. In this study, we analyzed land use efficiency from an economic perspective, which refers to effects of resource allocation and economic activities that show the extent to which the respective resource or labor value has been realized. The efficiency index framework is acquired at an early stage by selecting from an existing index that can reflect land use efficiency. This framework is then used to compute efficiency scores via Data Envelopment Analysis (DEA).

Many studies have investigated the relationships and coordination among economic, social, and environmental systems in city development [36]. Feiock and Stream [37] examined investment variations at the state level in the U.S. and found that administrative arrangements for environmental regulations have the potential to enhance economic development. Yang [38] calculated indexes for environmental quality, economic development, and degree of coordinated development in Guangzhou, China, to investigate both urban environment and economy. Differences in the studies above can be found in the creative study of the relationship between urban land use efficiency and coordination in China.

In this study we used urban land use efficiency and the degree of coordination as a measure of urban land use. Efficiency assesses the proportion of both inputs and outputs in the process of urban land use. Decreased input with increased output indicates higher efficiency, reflecting a high level of urban land use. The degree of land use coordination measures whether the development of all systems and elements of urban land use is reasonable and consistent. Improved coordination among subsystems reflects a higher level of urban land use [39,40]. The aim of this study was to discover a correlation between land use efficiency and land use coordination through a comprehensive evaluation of both parameters. The remainder of the paper is organized into three sections: the “Methodology” section describes the study region and approaches of evaluation and sample selection; the “Results and discussion” section provides a presentation and an analysis of the acquired results; and the “Concluding remarks” section offers a demonstrative conclusion to the study and policy implications of the findings.

2. Materials and Methods

2.1. Evaluation System

According to systematic theory, the related factors and their interactions in urban land use are an extensive system. In the specific analysis and evaluation process, the entire system can be decomposed into social, economic, and environmental subsystems [41]. Our analysis therefore focused on these subsystems of selected evaluation indexes [42].

2.2. Evaluation of Urban Land Use Efficiency

We employed data envelopment analysis (DEA) to evaluate urban land use efficiency. This method of analysis not only overcomes the limitations of a subjective assessment, but also tests inputs and outputs of efficiency, rendering it well suited for application in this study. We used the DEAP 2.1 software to conduct an efficiency analysis. We conducted the efficiency evaluation in terms of input and output indicators that we extracted from the entire process of urban land use, rather than specifically focusing on economic, social, and environmental subsystems.

In the system of urban land use, capital represents fixed asset investments of the society as a whole as well as its regenerative capacity. Therefore, we used the following input indicators: acreage of land, total year-end population, and total fixed assets investment (Table 1).

To measure the level of economic output in relation to urban land use, we used the regional gross domestic product (GDP). We assessed the social development using the people's standard of living, urban infrastructure, and the level of educational development, corresponding to the following output indicators: the Engel coefficient, the actual year-end acreage of road within a municipal district, and the number of kindergartens and ordinary schools (including ordinary primary schools, middle schools, colleges, and universities). To measure environmental development in urban land use, we assessed pollution control, represented by the "three wastes" output value, and the green level, which is the green coverage within a built-up area. Table 1 lists the selected input and output indicators of urban land use efficiency in this study.

Table 1. Urban land use efficiency evaluation indicators.

Input Indicators	Output Indicators
Land area	Gross Regional Product
Total year-end population	Engel coefficient
	Municipal districts end urban road area
Total fixed assets investment	Number of kindergartens and ordinary schools (including primary schools, middle schools, colleges and universities)
	Green coverage within built-up area

2.3. Evaluation of Urban Land Use Coordination

We employed the coordination coefficient method to study the degree of land use coordination. This method is widely applicable, features simple and clear calculations and intuitive results, and facilitates comparisons within time or spatial dimensions [43]. The principles of this method are as follows:

The data are Z standardized, and then the main components that reflect more than 80% of the system information are extracted, via principal component analysis. If there are two or more main components, their corresponding variance contribution is utilized as weight, added to the main components, and divided by their total variance contribution. Given the resulting main component of the system (extracted at a specific point in time), the level of development metric function value $F_i(t, x)$ can be obtained via calculation of $F_1(t, x)$ and $F_2(t, y)$.

We utilized Equation (1) to calculate the coordination degree between the systems S1 and S2.

$$G(t) = \{[F_1(t, x) \cdot F_2(t, x)]/[0.5F_1(t, x) + 0.5F_2(t, x)]^2\}^k \quad (1)$$

where $G(t)$ is the coordination degree between S1 and S2 within year t ; $F_1(t, x)$ is the development level metric of S1 within year t ; $F_2(t, y)$ is the development level metric of S2 within year t ; and, k is the distinguished coefficient.

When the $G(t)$ value calculated from Equation (1) exceeds a specific value, the coordination level is higher, which indicates the existence of a higher degree of coordination between S1 and S2 within year t .

We evaluated this coordination based on economic, social, and environmental aspects. The resulting indexing system of urban land use coordination evaluation is shown in Table 2.

Table 2. Urban land use coordination degree evaluation indicators.

Category	Indicator Name	Unit
Economic	Regional GDP	Billion RMB/km ²
	Regional total fixed assets investment	Billion RMB/km ²
	Regional general budget revenue	Billion RMB/km ²
Society	Engel coefficient	-
	The number of regional kindergartens and ordinary schools	Units/km ²
	The number of regional public libraries	Units/km ²
	The number of regional health institutions	Units/km ²
Environment	Sewage treatment rate	%
	Life garbage treatment rate	%
	Comprehensive utilization rate of industrial solid waste	%
	Green coverage of built-up area	%

2.4. Evaluation Coordinate System of Urban Land Use Efficiency and Coordination

We employed the Boston matrix method to study urban land use efficiency and coordination, as this method can classify samples according to individual performance in two aspects. We created a coordinate system with land use efficiency as the x -axis and land use coordination as the y -axis and plotted the corresponding data of our research subjects. A boundary value was set for both efficiency level and coordination degree (i.e., the coordinate origin (X_0, Y_0)). When the efficiency level (coordination degree) increased beyond this value, the efficiency (coordination) of urban land use is high, and vice versa. This coordinate system is shown in Figure 1.

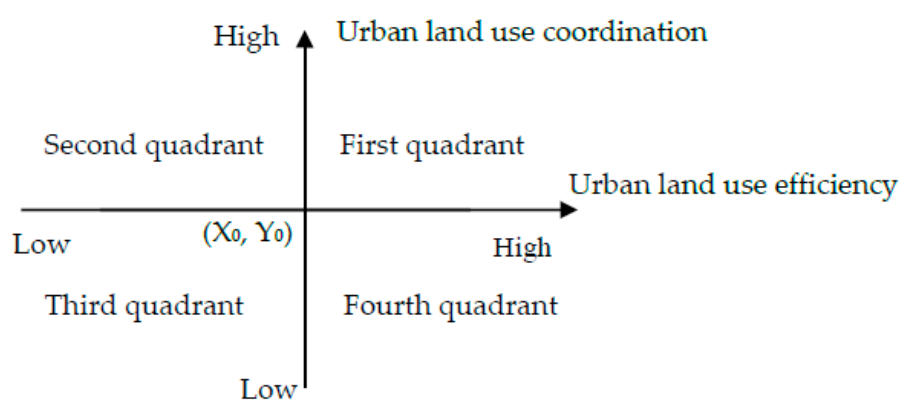


Figure 1. Evaluation coordinate system of urban land use efficiency and coordination.

2.5. Sample Selection

The samples in our study include 33 provincial capital cities, municipalities, and municipalities with independent planning status in China. The distribution of these cities basically covers the entire Chinese mainland; therefore, these representative cities can objectively reflect the basic situation of land use in China. The socio-economic data employed in this study were assembled from the China Statistical Yearbook and the region Area Statistical Yearbooks. All data of socio-economic indices were confirmed by the Chinese National Bureau of Statistics in 2015, which reflects real values of 2014.

We divided the cities in our sample into three geographical regions as follows: 17 eastern cities include Beijing, Tianjin, Harbin, Shanghai, Nanjing, Hangzhou, Fuzhou, Xiamen, Shenzhen, Haikou, Qingdao, Ningbo, Dalian, Changchun, Jinan, Shijiazhuang, and Shenyang; 6 central cities include Hefei, Nanchang, Wuhan, Taiyuan, Zhengzhou, and Changsha; and 10 western cities include Nanning, Chongqing, Lanzhou, Xi'ning, Yinchuan, Xi'an, Urumqi, Guiyang, Chengdu, and Hohhot.

3. Results

3.1. Urban Land Use Efficiency in China

The DEA method was utilized to analyze the urban land use efficiency in China. The results are shown in Table 3.

Table 3. Urban land use efficiency data envelopment analysis.

City	Firm	CRSTE	VRSTE	Scale	
Beijing	1	0.629	1	0.629	drs
Tianjin	2	0.506	1	0.506	drs
Shijiazhuang	3	0.534	0.802	0.666	drs
Taiyuan	4	0.739	0.741	0.997	drs
Hohhot	5	0.592	0.78	0.759	irs
Shenyang	6	0.393	0.394	0.999	drs
Dalian	7	0.571	0.598	0.954	drs
Changchun	8	0.611	0.676	0.905	drs
Harbin	9	0.602	0.719	0.837	drs
Shanghai	10	1	1	1	-
Nanjing	11	0.594	1	0.594	drs
Hangzhou	12	0.394	0.408	0.965	drs
Ningbo	13	0.548	0.648	0.846	drs
Hefei	14	0.413	0.428	0.965	drs
Fuzhou	15	0.585	0.729	0.803	drs
Xiamen	16	1	1	1	-
Nanchang	17	0.616	0.709	0.869	drs
Jinan	18	0.776	0.882	0.879	drs
Qingdao	19	0.699	1	0.699	drs
Zhengzhou	20	0.576	0.878	0.656	drs
Wuhan	21	0.411	0.426	0.964	drs
Changsha	22	0.642	0.812	0.791	drs
Shenzhen	23	1	1	1	-
Nanning	24	0.819	1	0.819	drs
Haikou	25	1	1	1	-
Chongqing	26	0.691	1	0.691	drs
Chengdu	27	0.507	0.933	0.544	drs
Guiyang	28	0.551	0.565	0.976	drs
Xi'an	29	0.555	0.837	0.664	drs
Lanzhou	30	0.815	0.847	0.962	drs
Xi'ning	31	0.745	0.816	0.913	irs
Yinchuan	32	0.624	0.877	0.712	irs
Urumqi	33	0.851	0.859	0.991	irs

Notes: CRSTE refers to the technical efficiency of CRS DEA (Constant Returns to Scale); VRSTE refers to the technical efficiency of VRS DEA (Variable Returns to Scale); Scale = scale efficiency = CRSTE/VRSTE; "drs" refers to diminishing returns to scale; "irs" refers to increasing returns to scale; "-" refers to constant returns to scale.

We analyzed these results focusing of Constant Returns to Scale (CRS) and Variable Returns to Scale (VRS). When CRS are constant, technical efficiency (TE) was considered, which stands for comprehensive efficiency. Technical efficiency of individual decision-making units is plotted as a column chart in Figure 2.

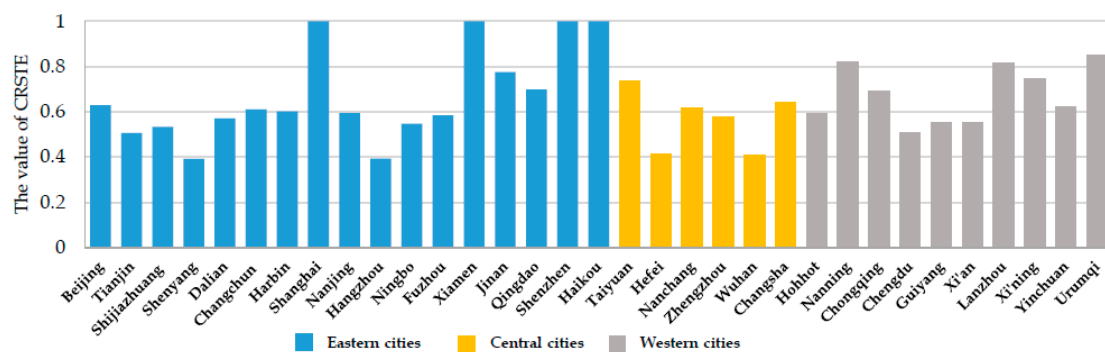


Figure 2. Comprehensive efficiency (CRSTE) of the urban land use for each city.

For variable VRS, we considered technical efficiency, which simply refers to the technology. The technical efficiency of individual decision-making units is plotted as a column chart in Figure 3.

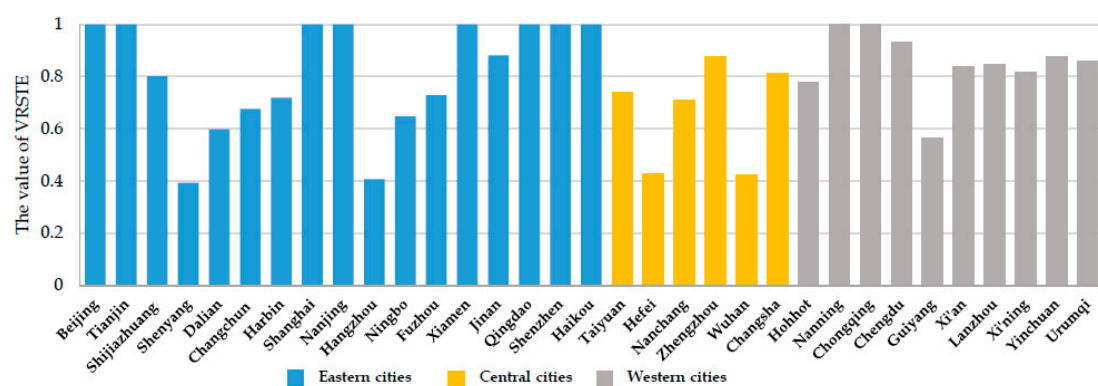


Figure 3. Technical efficiency (VRSTE) of the urban land use for each city.

The land use efficiencies of the cities in our sample vary depending on the region where the city is located. Four (23.5%) of the eastern cities have high land use efficiency (1.0), while five (29.4%) eastern cities have low land use efficiency (<0.80). None of the central cities have high land use efficiency, but two (33.3%) central cities have low land use efficiency. None of the western cities have high land use efficiency, but five (50%) western cities have low land use efficiency. This evident increase in low land use efficiency from eastern to central to western regions is in accordance with the distribution of economic development in China, indicating that lower land use efficiency coincides with areas with lower economic development more so than with areas of higher economic development. We suggest three main reasons for this phenomenon:

Cities with higher economic development have a much more reasonable industrial structure [44]. In contrast to primary and secondary industries, tertiary industries use higher land productivity per unit area, leading to higher land use efficiency in developed cities.

Cities with higher economic development place more emphasis on the relationship between economic development and social development. The theory of economic development suggests increasing social demands for better infrastructure, better education, and better medical services with the improvement of the level of economic development in a city. This guides local government investment to improve public services, which in turn increases land development intensity and land productivity per unit area [45].

Cities with higher economic development invested far more in environmental protection, which can also increase land productivity per unit area.

3.2. The Degree of Urban Land Use Coordination in China

We evaluated the coordination level of economic (S1) and social environmental systems (S2) for the urban land use of each city. The results of principal component analysis of the standardized data are shown in Tables 4 and 5. In the extraction process for the economic system, the variance contribution of the first principal component reached 90.075, revealing a good result. Therefore, the first principal component can represent each of the developmental levels of the economic system of a city. The extraction process for the social environmental system revealed a cumulative variance contribution of the first four principal components of 80.277. Therefore, the first four principal components can represent the developmental level of the social environmental system of each city.

Table 4. Total variance table for principal component analysis of the urban economic system.

Component	Initial Eigenvalues			Extraction Sums of Squared Loadings		
	Total	Variance Contribution	Cumulative Variance Contribution	Total	Variance Contribution	Cumulative Variance Contribution
1	2.702	90.075	90.075	2.702	90.075	90.075
2	0.245	8.156	98.230	-	-	-
3	0.053	1.770	100.000	-	-	-

Table 5. Total variance explained table for urban social environment system principal component analysis.

Component	Initial Eigenvalues			Extraction Sums of Squared Loadings		
	Total	Variance Contribution	Cumulative Variance Contribution	Total	Variance Contribution	Cumulative Variance Contribution
1	3.602	40.021	40.021	3.602	40.021	40.021
2	1.342	14.911	54.932	1.342	14.911	54.932
3	1.260	13.995	68.927	1.260	13.995	68.927
4	1.021	11.349	80.277	1.021	11.349	80.277
5	0.686	7.620	87.897	-	-	-
6	0.624	6.934	94.831	-	-	-
7	0.315	3.504	98.334	-	-	-
8	0.085	0.946	99.281	-	-	-
9	0.065	0.719	100.000	-	-	-

According to Equation (2), the horizontal metrics $F_1(2014, x_i)$ and $F_2(2014, y_j)$ are equal to the main factor in the extracted variance contribution-weighted average.

$$\begin{aligned} F_1(t, x) &= \sum_{i=1}^n a_i x_{it} \\ F_2(t, y) &= \sum_{j=1}^n b_j y_{jt} \end{aligned} \quad (2)$$

where a_i is the characteristic index weight of S1, which is the principal component variance contribution of S1; and b_j is the characteristic index weight of S2, which is the principal component variance contribution of S2.

We utilized Equation (2) to determine the degree of coordination between the economic system and the social environmental system G (2014), as shown in Table 6.

The degree of land use coordination of the sampled cities varied according to the region, a given city is located in and its land use efficiency. Five (29.4%) of the eastern cities have relatively high land use coordination degrees (>0.9), while two (11.8%) eastern cities have especially low land use coordination degrees (<-1.0). Two (33.3%) central cities have relatively high land use coordination

degrees (>0.9), while four (66.7%) central cities have especially low land use coordination degrees (<-1.0). Two (20.0%) western cities have high land use coordination degrees (>0.9), while two (20.0%) eastern cities have particularly low land use coordination degrees (<-1.0).

Table 6. The coordination degree of the economic subsystem and the social environmental subsystem.

City	G (2014)	City	G (2014)
Taiyuan	0.996	Shanghai	0.639
Beijing	0.994	Shijiazhuang	0.582
Hohhot	0.986	Fuzhou	0.490
Harbin	0.971	Urumqi	0.413
Changchun	0.969	Chengdu	0.160
Yinchuan	0.949	Xi'ning	-0.049
Nanjing	0.945	Jinan	-0.256
Shenzhen	0.925	Haikou	-0.279
Hefei	0.918	Tianjin	-0.310
Xiamen	0.891	Wuhan	-0.697
Hangzhou	0.881	Xi'an	-1.179
Guiyang	0.845	Nanchang	-1.417
Shenyang	0.840	Qingdao	-2.752
Chongqing	0.783	Zhengzhou	-10.718
Changsha	0.748	Nanning	-13.653
Dalian	0.729	Ningbo	-3740.488
Lanzhou	0.671	-	-

3.3. Urban Land Use Efficiency and Coordination in China

We plotted the values for each city on a coordinate system with urban land use efficiency on the x -axis and coordination degree on the y -axis. The origin was defined for efficiency and coordination as values of 0.99 and 0.9, respectively. We detected a wide range of coordination values in our sample; consequently, to arrive at clearer results, we omitted those cities with a coordination value below -1.0 . The resulting plot is displayed in Figure 4.

Based on the above analysis, the cities of our sample can be divided into four categories according to the occupied quadrant of the coordinate system. One (5.9%) eastern city (Shenzhen) is the only city which features both high efficiency and well-coordinated development (with coordinates of 1, >0.9). Three (17.6%) eastern cities have highly efficient but poorly coordinated development (with coordinates of 1, <0.9), including Shanghai, Xiamen, and Haikou. Four (23.5%) eastern cities have inefficient but well-coordinated development (with coordinates of <1 , >0.9), including Beijing, Changchun, Harbin, and Nanjing. Two (33.3%) central cities have inefficient but well-coordinated development (with coordinates of <1 , >0.9), including Taiyuan and Hefei. Two (20.0%) western cities have inefficient but well-coordinated development (with coordinates of <1 , >0.9), including Hohhot and Yinchuan. Nine (52.9%) eastern cities have inefficient and poorly coordinated development (with coordinates of <1 , <0.9), including Tianjin, Shijiazhuang, Shenyang, Dalian, Hangzhou, Ningbo, Fuzhou, Jinan, and Qingdao. Four (66.7%) central cities have inefficient and poorly coordinated development (with coordinates of <1 , <0.9), including Nanchang, Zhengzhou, Wuhan, and Changsha. Eight (80%) western cities have inefficient and poorly coordinated development (with coordinates of <1 , <0.9), including Nanning, Chongqing, Chengdu, Guiyang, Xi'an, Lanzhou, Xi'ning, and Urumqi.

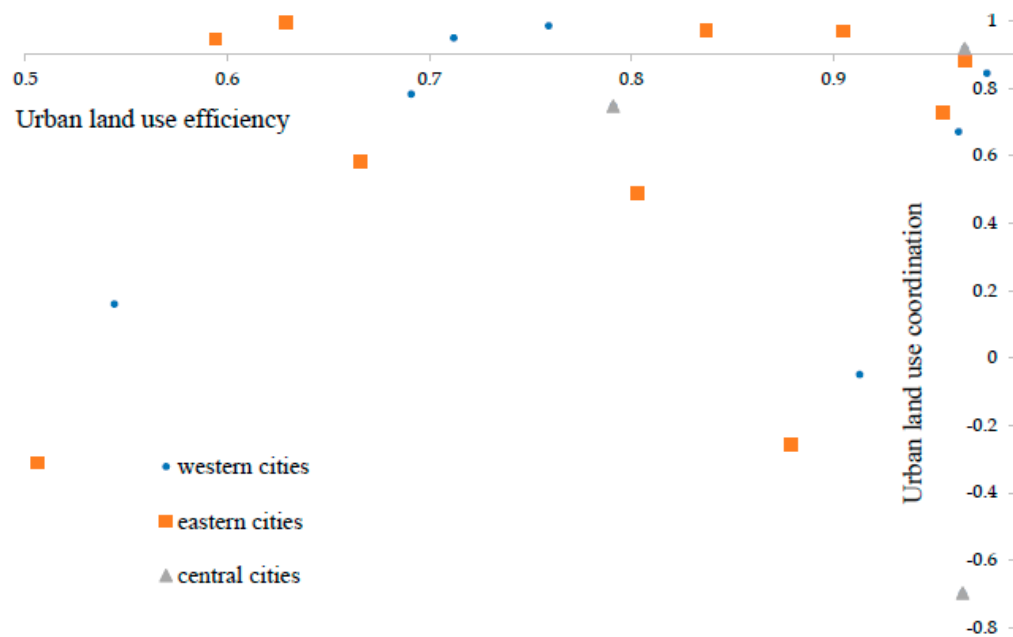


Figure 4. Coordinate system of urban land use efficiency and coordination.

4. Discussion and Conclusions

This study aimed to analyze urban land use efficiency and coordination of China from an economic perspective. The results revealed a gradually increasing proportion of land use efficiency from eastern to central to western regions in China, coinciding with China's patterns of socioeconomic development. Regions with lower economic development have lower land use efficiency. Moreover, we found a strong correlation between low land use efficiency and low land use coordination. There are nine cities which have low land use coordination degrees among twelve cities with low land use efficiency. Furthermore, our results indicate no significant correlation between high levels of urban land use efficiency and land use coordination in our sample. Due to the limited nature of the data, this study does not address the impact of land use types on land use efficiency. Moreover, the development direction of a city is relatively small, with no distinction between resource-based cities or service-oriented cities, etc., which may have an impact on urban land use. Therefore, in future research, factors such as land use type and urban development planning should be taken into account to obtain a more accurate evaluation of land use coordination.

This paper analyzed urban land use efficiency and coordination in China based on a sample of 33 provincial capital cities, municipalities, and municipalities with independent planning status in China. For this, we combined theoretical analysis, data envelopment analysis, principal component analysis, the coordination coefficient method, and four-quadrant analysis. We decomposed the entire system into social, economic, and environmental subsystems via theoretical analysis and employed data envelopment analysis (DEA) to evaluate urban land use efficiency. The results revealed a gradually increasing proportion of low land use efficiency from eastern to central to western regions of China, coinciding with China's patterns of socioeconomic development (see Table 3). We employed the coordination coefficient method to study the degree of land use coordination and calculated the land use coordination degree of 33 sample cities (see Table 6). We employed the Boston matrix method to study urban land use efficiency and coordination (see Figure 4) and evaluated the coordination level of the economic (S1) and social environmental systems (S2) for the urban land use of each city. The results of a principal component analysis of the standardized data are shown in Tables 4 and 5.

As engines of national economic growth, the local economies of Beijing, Shanghai, Shenzhen, and other core cities in China are directly related to and affect the country as a whole and the regional economic development. These cities are under great pressure to increase economic

development and to improve noneconomic urban conditions to keep pace with the fast-growing economy. Therefore, although these cities feature high economic development, reflected in high land use efficiency, such development is generally not well coordinated. When urban land use lacks efficiency, local governments tend to intensify their efforts to develop the urban land, thus over-emphasizing economic returns, while neglecting other aspects of city construction. Such practices lead to ever-present problems, e.g., lack of public service, environmental pollution, and low land use coordination.

The policy implications are clear. First, the Chinese government should strengthen the scientific basis of urban land use planning and improve its enforcement. Rational urban land use planning is vital to the effective operation of a city [46]. Coordinated economic, social, and environmental urban development should also be considered. However, under China's existing fiscal decentralization policy, the central government's assessment of local officials is assumed to be GDP-oriented, which inevitably leads to an emphasis on economic construction at the expense of other aspects of urban planning. Therefore, the central government and all levels of government should increase supervision of urban land use planning of local governments and guide them to optimize urban land use planning in line with the individual characteristics of a city. Furthermore, the single GDP-oriented assessment of local officials needs to change. Instead, local officials should be guided to plan for the long-term development of their city to achieve long-lasting, efficient, and coordinated development.

Second, the independence of land management agencies should be promoted and a system of dual supervision by both the government and the public should be established. One reason for the severe urban land use problem in China is that local governments are both law enforcers and offenders. Local land management agencies are subordinates of local governments and cannot effectively constrain the behavior of local governments. Thus, the vertical management functions of land management agencies should be strengthened to promote independence of such agencies, consequently weakening or eliminating intervention by the local government. China can also learn from the successful land management efforts of other countries and establish a regime with a Land Management Commissioner.

Third, suppressing the investment impulses of local governments and strengthening governmental management functions are effective ways to promote land use. One central problem that currently affects China is that local governments have diverted extensive energy and resources to the economy, while neglecting the development of social undertakings and environmental protection, resulting in difficulties in developmental coordination among these three systems [47,48]. To address this situation, the GDP-oriented developmental approach to economic development should be reformed to suppress the investment impulsion of local governments. Local governments should also adjust their allocation of fiscal expenditures to improve public services and environmental protection, thus transitioning from a production-oriented government to a service-oriented government.

Finally, organizations outside of government agencies should be encouraged to play a more prominent role in the development of social undertakings and environmental protection, thus counteracting the adverse impact of governmental absence in urban development. The government could support such efforts by offering positive publicity, funding, and tax benefits.

Since the proposed Boston matrix method of urban land use efficiency and coordination of ideas is still in development, the resulting ideas and methods of this study are innovative and exploratory. To optimize the effectiveness of this relatively simple method for application to management practice, further in-depth studies should employ an improved research methodology via inter alia, independent evaluation of each city.

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References

1. López, E.; Bocco, G.; Mendoza, M.; Duhau, E. Predicting land-cover and land-use change in the urban fringe: A case in Morelia city, Mexico. *Landsc. Urban Plan.* **2001**, *55*, 271–285. [[CrossRef](#)]
2. Waddell, P. UrbanSim: Modeling urban development for land use, transportation, and environmental planning. *J. Am. Plan. Assoc.* **2002**, *68*, 297–314. [[CrossRef](#)]
3. Saba, R.S.; Abdolrassoul, S.M.; Seyed, M.M.; Ali, A.A. Sustainability through uncertainty management in urban land suitability assessment. *Environ. Eng. Sci.* **2013**, *30*, 170–178.
4. Choy, L.H.T.; Lai, Y.; Lok, W. Economic performance of industrial development on collective land in the urbanization process in China: Empirical evidence from Shenzhen. *Habitat Int.* **2013**, *40*, 184–193. [[CrossRef](#)]
5. Chen, R.; Ye, C.; Cai, Y. The impact of rural out-migration on land use transition in China: Past, present and trend. *Land Use Policy* **2014**, *40*, 101–110. [[CrossRef](#)]
6. Jean-Marie, H.; Szymon, M.; Erwin, K. The adaptive efficiency of land use planning measured by the control of urban sprawl. The cases of the Netherlands, Belgium and Poland. *Land Use Policy* **2012**, *29*, 887–898.
7. Southavilay, B.; Teruaki, N.; Shigeyoshi, T.; Tetsuo, S. Land use change and its determinant factors in northern Laos: Spatial and socio-economic analysis. *J. Agric. Sci.* **2012**, *4*, 190–204.
8. Jin, S.; Jayne, T.S. Land rental markets in Kenya: Implications for efficiency, equity, household income, and poverty. *Land Econ.* **2013**, *89*, 246–271. [[CrossRef](#)]
9. Janine, M.U.; Julian, F.Q. How to combine tradition and modernity? Regulating customary land management in Ghana. *Land Use Policy* **2008**, *25*, 198–213.
10. Thomas, G.J.; Mitsuru, S.; Dennis, L.E. Developing county-level commodity-flow models incorporating land-use characteristics and economic factors for Utah. *J. Urban Plan. Dev.* **2012**, *138*, 35–42.
11. Ruben, N.L.; Andrew, J.P.; Robert, N.S. What drives land-use change in the United States? A national analysis of landowner decisions. *Land Econ.* **2008**, *84*, 529–550.
12. Mason, G. The role of land markets in economic crises. *Am. J. Econ. Sociol.* **2009**, *68*, 854–888.
13. Bardo, J.W.; Hartman, J.J. *Urban Sociology: A Systematic Introduction*; F.E. Peacock: Itasca, IL, USA, 1982.
14. Chapin, F.S.; Kaiser, E.J. *Urban Land Use Planning*; University of Illinois Press: Champaign, IL, USA, 1970.
15. Macedo, J. Urban land policy and new land tenure paradigms: Legitimacy vs. legality in Brazilian cities. *Land Use Policy* **2008**, *25*, 259–270. [[CrossRef](#)]
16. Kironde, J. Land policy options for urban Tanzania. *Land Use Policy* **1997**, *14*, 99–117. [[CrossRef](#)]
17. Terry, V.D.; Raoul, B. Laws, people and land use: A sociological perspective on the relation between laws and land use. *Eur. Plan. Stud.* **2009**, *17*, 1797–1815.
18. Jessica, I. From squatter to settler: Applying the lessons of nineteenth century U.S. public land policy to twenty-first century land struggles in Brazil. *Ecol. Law Quart.* **2011**, *38*, 179–232.
19. Pauleit, S.; Ennos, R.; Golding, Y. Modeling the environmental impacts of urban land use and land cover change—A study in Merseyside, UK. *Landsc. Urban Plan.* **2005**, *71*, 295–310. [[CrossRef](#)]
20. Mouri, G.; Aisaki, N. Using land-use management policies to reduce the environmental impacts of livestock farming. *Ecol. Complex.* **2015**, *22*, 169–177. [[CrossRef](#)]
21. Mirzaei, M.; Solgi, E.; Salmanmahiny, A. Assessment of impacts of land use changes on surface water using L-THIA model (case study: Zayandehrud river basin). *Environ. Monit. Assess.* **2016**, *188*, 690. [[CrossRef](#)] [[PubMed](#)]
22. Lin, G.; Ho, S.P. China's land resources and land-use change: Insights from the 1996 land survey. *Land Use Policy* **2003**, *20*, 87–107. [[CrossRef](#)]
23. Sanjiv, K.; Venkatesh, M. Characterizing long-term land use/cover change in the United States from 1850 to 2000 using a nonlinear Bi-analytical Model. *R. Swed. Acad. Sci.* **2013**, *42*, 285–297.
24. Azam, H.; Bahman, J.A.; Jan, A.; Nicola, F.; Kaneyuki, N. Assessing the impacts of four land use types on the water quality of wetlands in Japan. *Water Resour. Manag.* **2013**, *27*, 2217–2229.

25. Byungil, K.; Hyounkyu, L.; Hyungbae, P.; Hyoungkwan, K. Estimation of greenhouse gas emissions from land-use changes due to road construction in the republic of Korea. *J. Constr. Eng. Manag.* **2013**, *139*, 339–346.
26. Irwin, E.G.; Bockstael, N.E. Land use externalities, open space preservation, and urban sprawl. *Reg. Sci. Urban Econ.* **2004**, *34*, 705–725. [[CrossRef](#)]
27. Pouriyeh, A.; Khorasani, N.; Lotfi, F.H. Efficiency evaluation of urban development in Yazd City, Central Iran using data envelopment analysis. *Environ. Monit. Assess.* **2016**, *188*, 618. [[CrossRef](#)] [[PubMed](#)]
28. Herold, M.; Couclelis, H.; Clarke, K.C. The role of spatial metrics in the analysis and modeling of urban land use change. *Comput. Environ. Urban* **2005**, *29*, 369–399. [[CrossRef](#)]
29. Aguilera, F.; Valenzuela, L.M.; Botequilha-Leitão, A. Landscape metrics in the analysis of urban land use patterns: A case study in a Spanish metropolitan area. *Landsc. Urban Plan.* **2011**, *99*, 226–238. [[CrossRef](#)]
30. Herold, M.; Scepan, J.; Clarke, K.C. The use of remote sensing and landscape metrics to describe structures and changes in urban land uses. *Environ. Plan. A* **2002**, *34*, 1443–1458. [[CrossRef](#)]
31. Glaeser, E.L.; Kahn, M.E. Sprawl and urban growth. In *Handbook of Regional and Urban Economics*; Elsevier: Amsterdam, The Netherlands, 2003; pp. 2481–2527.
32. Peiser, R.B. Density and urban sprawl. *Land Econ.* **1989**, *65*, 193–204. [[CrossRef](#)]
33. Fang, X. The index system and method study of land use efficiency measurement. *Syst. Eng.* **2004**, *22*, 22–26. (In Chinese)
34. Wang, H.; Wang, L.; Su, F.; Tao, R. Rural residential properties in China: Land use patterns, efficiency and prospects for reform. *Habitat Int.* **2012**, *36*, 201–209. [[CrossRef](#)]
35. Xie, H.; Wang, W. Spatiotemporal differences and convergence of urban industrial land use efficiency for China's major economic zones. *J. Geogr. Sci.* **2015**, *25*, 1183–1198. [[CrossRef](#)]
36. Chia-Nung, L.; Tsung-Yu, L. An empirical analysis of links between transport and land use. *Transport* **2013**, *166*, 105–112.
37. Feiock, R.C.; Stream, C. Environmental protection versus economic development: A false trade-off? *Public Admin. Rev.* **2001**, *61*, 313–321. [[CrossRef](#)]
38. Yang, S. A study on the forecast and regulation of coordinated development of urban environment and economy in Guangzhou. *Chin. Geogr. Sci.* **1995**, *5*, 248–256. (In Chinese) [[CrossRef](#)]
39. Zitti, M. Long-term urban growth and land use efficiency in Southern Europe: Implications for sustainable land management. *Sustainability* **2015**, *7*, 3359–3385. [[CrossRef](#)]
40. Wang, Y.; Zhou, L.H. Assessment of the coordination ability of sustainable social-ecological systems development based on a set pair analysis: A case study in Yanchi county, China. *Sustainability* **2016**, *8*, 733. [[CrossRef](#)]
41. Xie, H.; Wang, W. Exploring the spatial-temporal disparities of urban land use economic efficiency in China and its influencing factors under environmental constraints based on a sequential slacks-based model. *Sustainability* **2015**, *7*, 10171–10190. [[CrossRef](#)]
42. Gong, S.Y. Study on the coupled coordination degrees between urban land intensive use system and social economic development: A case of Shandong province. *Adv. Mater. Res.* **2013**, *731*, 4827–4830. [[CrossRef](#)]
43. Ding, S. Comprehensive Study on Urban Land Use Efficiency and Coordination. Master's Thesis, Harbin Institute of Technology, Harbin, China, 2011. (In Chinese)
44. Deng, Q.H.; Zhu, J. Chongqing's Economic Development and Its Cultivation of Sub-Regional Centers. *J. Chongqing Univ.* **2006**, *12*, 6–12. (In Chinese)
45. Hong, D.Y. A higher level of social progress leads the new economic development. *Seeking* **2016**, *7*, 52–54.
46. Sakieh, Y.; Salmanmahiny, A.; Jafarnezhad, J. Evaluating the strategy of decentralized urban land-use planning in a developing region. *Land Use Policy* **2015**, *48*, 534–551. [[CrossRef](#)]
47. Long, H.L.; Liu, Y.Q.; Li, T.T. Spatial Interlinking of Land Use Planning and Environmental Protection Planning from the Perspective of Ecological Civilization Construction. *Econ. Geogr.* **2014**, *34*, 1–8. (In Chinese)
48. Zhang, K.; Wen, Z. Review and challenges of policies of environmental protection and sustainable development in China. *J. Environ. Manag.* **2008**, *88*, 1249–1261. [[CrossRef](#)] [[PubMed](#)]

