

Article

Quantification of Spatial Association between Commercial and Residential Spaces in Beijing Using Urban Big Data

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Abstract: Commercial and residential spaces are two core types of geographical objects in urban areas. However, these two types of spaces are not independent of each other. Spatial associations exist between them, and a thorough understanding of this spatial association is of great significance for improving the efficiency of urban spatial allocation and realizing scientific spatial planning and governance. Thus, in this paper, the spatial association between commercial and residential spaces in Beijing is quantified with GIS spatial analysis of the average nearest neighbor distance, kernel density, spatial correlation, and honeycomb grid analysis. Point-of-interest (POI) big data of the commercial and residential spaces is used in the quantification since this big data represents a comprehensive sampling of these two spaces. The results show that the spatial distributions of commercial and residential spaces are highly correlated, maintaining a relatively close consumption spatial association. However, the degrees of association between different commercial formats and residential spaces vary, presenting the spatial association characteristics of “integration of daily consumption and separation of nondaily consumption”. The commercial formats of catering services, recreation and leisure services, specialty stores, and agricultural markets are strongly associated with the residential spaces. However, the development of frequently used commercial formats of daily consumption such as living services, convenience stores, and supermarkets appears to lag behind the development of residential spaces. In addition, large-scale comprehensive and specialized commercial formats such as shopping malls, home appliances and electronics stores, and home building materials markets are lagging behind the residential spaces over a wide range. This paper is expected to provide development suggestions for the transformation of urban commercial and residential spaces and the construction of “people-oriented” smart cities.

Keywords: commercial space; residential space; spatial structure; spatial association; urban POI data



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1. Introduction

As one of the most important types of places for human activities, urban spaces are composed of various subfunctional spaces. Of these functional spaces, commercial and residential spaces are the two most active and core components of urban spaces, and they have been a focus of urban geography research [1,2]. Along with economic and social development, commercial and residential spaces have gradually become two closely related physical spaces among urban functional spaces, with the characteristics of causal symbiosis, spatial dependence, and close association [3,4]. Thus, research on the spatial structure of commercial and residential spaces and their association can not only effectively explore the organization mode and mechanism of urban commercial and residential spaces, but also help to optimize the organization structure of urban spaces, to improve the allocation efficiency of urban spaces, and to realize scientific spatial planning and governance [5].

Currently, with the scale of urban built-up areas expanding from large cities to megacities, the number of urban populations increasing from millions to tens of millions, and the

consumption behavior of residents transforming from physical consumption to “entity-online” dual consumption, urban commercial and residential spaces have also changed accordingly [6–10]. On the one hand, the marketization reform of land and housing and the accelerated development of suburbanization have led to the spatial restructuring of commercial and residential spaces [11]. On the other hand, the rapid development of e-commerce and online shopping has fundamentally affected the attributes, format, and scale of original urban commercial spaces [12], which further drives the spatial association between commercial and residential spaces to self-adjust, upgrade, and transform. In addition, a mismatch exists between commercial and residential spaces from the perspective of their grade, scale, format and spatial distribution [2]. Therefore, it is necessary to investigate and to further clarify the spatial structure of commercial and residential spaces and their association relationship in the current study of urban spatial organization.

Currently, the research on commercial and residential spaces in urban areas mainly focuses on their respective spatial characteristics, evolutionary features and influence on urban spaces [13–21]. There are also a large number of associated research fields [22,23]. For example, in housing research, the accessibility to commercial sites is usually taken as an explanatory variable [22]. In addition, a huge number of studies have focused on mixed land use (commercial, residential, and industrial) [23]. Only a few scholars have studied the spatial relationship between commercial and residential spaces [2,24–29]. For instance, Zhou et al. [2] studied the relationship and evolution mechanism of commercial and residential spaces in Guangzhou based on residents’ consumption behavior. In addition, a mismatch between commercial and residential spaces can be found in different cities from the perspective of residential suburbanization, real estate development, and daily shopping travel distance [24–28]. Xue et al. [28] studied the relationship between residential prices and retail spatial heterogeneity. Previous studies have constructively explored the relationship between commercial and residential spaces. By investigating the changes in the retail service industry in New York City, Meltzer and Schuetz [29] found that the retail stores in low-income and minority communities are few and homogeneous. However, most of the existing studies are based on limited consumption behavior data from questionnaire surveys or their statistical data. These consumption behavior data can be used only for a certain small area, while large numbers of commercial and residential spaces are widely distributed in urban areas. Moreover, from the perspective of commercial formats, the commercial formats have become richer due to the impact of online consumption [30]. The spatial association between different formats of commercial spaces and residential spaces needs to be investigated and explored in depth.

Recently, with the development of big data acquisition and analysis technology, point-of-interest (POI) data in geographical information science have gradually become an important basis for explaining urban spatial organization and structure [31]. Because this type of data can sample all the geographic entities on an electronic map, these urban POI data can effectively reflect the spatial and attribute information of geographic entities [32] and considerably widen the channels of urban spatial organization research [32–35]. At present, the number of commercial service facilities and residential communities in cities is growing extremely fast. Meanwhile, the spatial distribution of commercial and residential spaces has evolved from monocentric to polycentric or even decentered patterns [36]. This change gives the POI data of commercial and residential spaces, with their comprehensive sampling characteristics, an obvious advantage over traditional limited consumption behavior data. POI data provide a new perspective for quantifying the spatial associations between commercial and residential spaces in urban areas.

In this paper, with the POI data of commercial service facilities and residential communities from electronic maps, the spatial structure of commercial and residential spaces and their association relationships are investigated by using GIS spatial analysis of the average nearest neighbor distance, kernel density, and spatial correlation. Honeycomb grid analysis is also applied to identify the associated types of commercial and residential spaces. The specific questions to be addressed are: (1) For the commercial and residential

spaces, who goes first in the accelerated development of suburbanization? (2) Which kind of commercial space would contribute to the development of residential space? Which kind of commercial space would be attracted by the development of residential space? Further, this study can provide implications to develop a commercial–residential balanced region, which should be the key issue in polycentric urban development.

2. Materials and Methods

2.1. Materials

2.1.1. Study Area

In this research, the city of Beijing in China is selected to investigate the spatial association between commercial and residential spaces. Beijing, with a long history of commercial development and well-developed commercial service functions, is one of the top ten commercial cities in China. In recent years, the commercial space of Beijing has gradually transformed from a monocentric pyramidal hierarchy to a polycentric flattened hierarchy [36]. Meanwhile, in the context of market-oriented transformation and rapid urbanization, the differentiation of social class in residential spaces in Beijing is becoming increasingly significant. Thus, the spatial association between commercial and residential spaces has been undergoing profound transformation. Therefore, Beijing is a representative city for exploring the spatial association between commercial and residential spaces. Beijing consists of 16 districts (Figure 1) and 4 zones divided according to Beijing Urban Master Plan (2016–2035), i.e., the core area, the central area, the inner suburb, and the outer suburb. In addition, four grades of commercial centers can also be found in Beijing, i.e., the regional level, municipal level, district level, and community level (Figure 1).

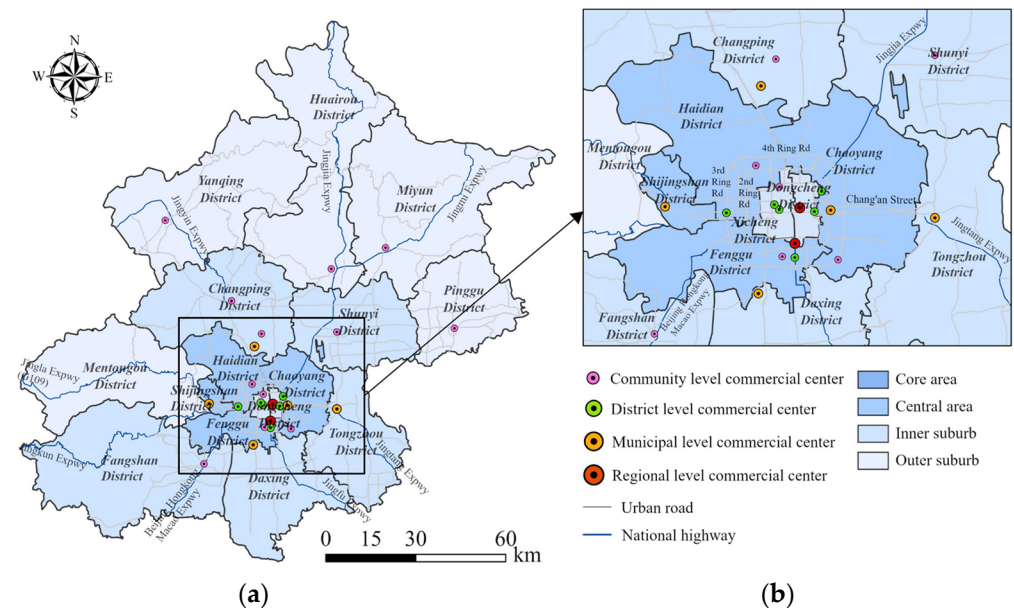


Figure 1. The study area of Beijing: (a) the 16 administrative districts and 4 overall planning zones in Beijing, and (b) the core and central area of Beijing, with different grades of commercial centers.

2.1.2. Data

The POI data of commercial and residential spaces are obtained from the 2018 Baidu electronic map and the Lianjia.com data, respectively. A total of 235,549 valid commercial POI data points and 7618 valid residential POI data points were obtained in this research (Tables 1 and 2 and Figure 2). The POI data of commercial spaces include information such as the merchant name, address, latitude and longitude coordinates, and commercial formats. According to the “Retail Format Classification” national standard in China and the POI data attributes, the commercial formats in Beijing are divided into 10 categories (Table 1). The residential POI data include the name of the residential district, housing

price, building age, and latitude and longitude. To ensure the minimum within-group difference and the maximum between-group difference [37], the Jenks natural breaks algorithm is applied to divide the housing prices into three groups of residential space: high-grade (106,600–280,100 RMB/m²), mid-grade (61,300–106,500 RMB/m²), and low-grade (10,000–61,200 RMB/m²).

Table 1. Classification of commercial formats.

Commercial Format	POI Subclasses	Number	Proportion
Catering services	Chinese restaurants, fast food restaurants, tea houses, cafes, pastry bakeries, foreign restaurants, hot pot restaurants	97,198	41.26%
Recreation and leisure services	Karaoke television (KTV), bowling alleys, fishing parks, cinemas, ski resorts, golf courses, fitness centers, game centers, skating rinks, agritainment, theaters, amusement parks, multipurpose sports stadiums	23,361	9.92%
Living services	Beauty salons, bath and massage facilities, logistics and express delivery facilities, laundries, telecommunication business offices, photography and printing facilities, post offices	43,144	18.32%
Convenience stores	/	15,624	6.63%
Supermarkets	/	7202	3.06%
Shopping malls	/	882	0.37%
Home appliance and electronics stores	Comprehensive home appliance stores	457	0.19%
Home building materials markets	Fabric markets, lamp and porcelain markets, comprehensive home building material markets	301	0.13%
Specialty stores	Personal product stores, sporting goods stores, cultural goods stores, clothing stores, shoe stores, hat and leather goods stores	38,838	16.49%
Agricultural markets	/	8542	3.63%

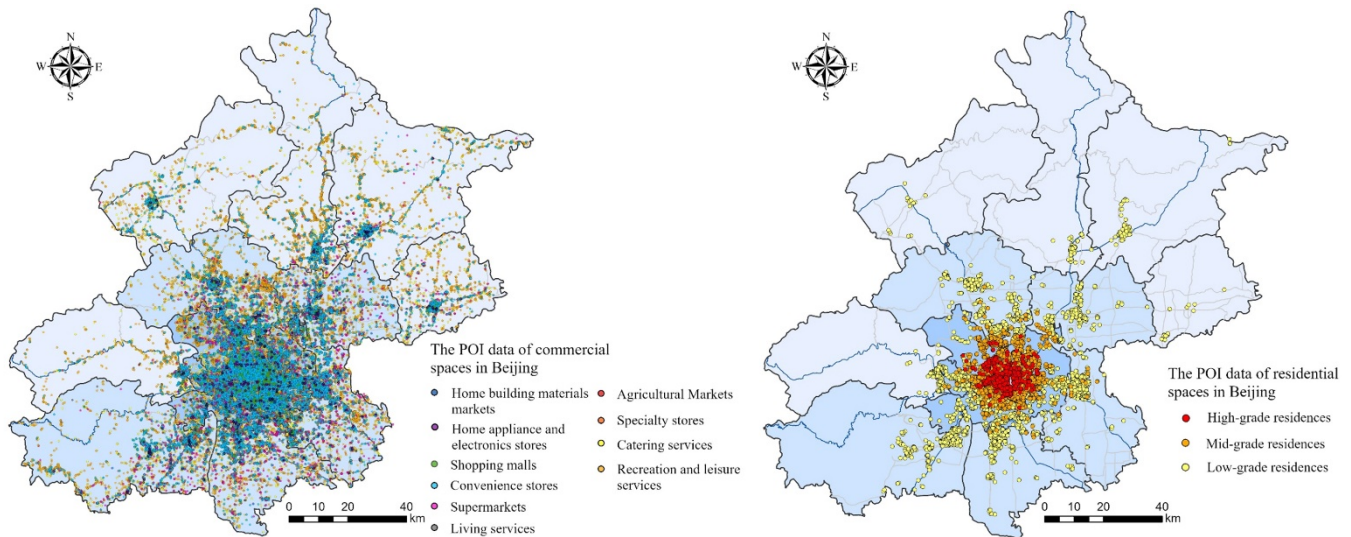


Figure 2. The POI data of commercial and residential spaces in Beijing.

Table 2. The POI data of commercial and residential spaces in Beijing.

	Administrative District	Number of Commercial Space POIs								Number of Residential Space POIs				
		Catering Services	Recreation and Leisure Services	Living Services	Convenience Stores	Supermarkets	Shopping Malls	Home Appliance and Electronics Stores	Home Building Materials Markets	Specialty Stores	Agricultural Markets	High-Grade Residences	Mid-Grade Residences	Low-Grade Residences
Core area	Dongcheng	5386	841	1761	634	228	61	16	2	2908	284	206	377	27
	Xicheng	5387	856	1951	756	244	54	23	0	3585	455	440	361	19
Central area	Haidian	12,227	2538	4746	1285	707	101	66	31	5331	873	145	928	169
	Chaoyang	25,213	4577	11,616	2620	1432	238	75	100	8738	1659	43	772	653
	Fengtai	7819	1241	4197	1150	586	73	54	48	3700	1032	5	172	681
Inner suburb	Shijingshan	1860	376	1057	290	128	13	9	7	946	153	0	14	181
	Tongzhou	6150	1219	3113	1263	659	46	47	13	2320	610	0	13	463
	Daxing	7230	1262	3275	1201	711	58	34	13	2428	563	0	10	383
	Shunyi	4725	1079	2447	1113	594	58	25	20	1491	563	0	21	238
	Changping	8840	2150	3947	1582	751	81	24	25	3032	772	0	16	583
Outer suburb	Fangshan	4223	1348	1724	932	466	42	40	14	1467	474	0	0	363
	Mentougou	850	451	384	206	48	10	2	1	237	104	0	2	180
	Yanqing	1352	923	450	438	68	12	4	12	412	110	0	0	15
	Huairou	2023	1933	817	854	173	12	8	4	595	287	0	0	52
	Miyun	2449	1701	989	865	243	12	18	5	988	344	0	0	67
	Pinggu	1464	866	670	435	164	11	12	6	660	259	0	0	19
Total		97,198	23,361	43,144	15,624	7202	882	457	301	38,838	8542	839	2686	4093

2.2. Methods

2.2.1. Average Nearest Neighbor Distance

The average nearest neighbor distance is used to measure the macroscopic clustering degree of commercial spaces (or a certain commercial format) and residential spaces of the whole study area by calculating the distance between each element and its nearest neighboring element. The nearest neighbor ratio R is the ratio of the observed average distance and the expected average distance [16], calculated as follows:

$$R = \frac{\bar{D}_O}{\bar{D}_E} \quad (1)$$

$$\bar{D}_O = \frac{\sum_{i=1}^n d_i}{n}, \bar{D}_E = \frac{0.5}{\sqrt{n/A}} \quad (2)$$

where \bar{D}_O is the observed average distance between each element and their nearest neighboring elements, \bar{D}_E is the expected average distance in a random pattern, d_i is the distance of the nearest neighboring element pairs, n is the number of elements in the region, and A is the area of the envelope of all elements. When the value of R is less than 1, the elements are considered spatially agglomerated. The smaller the value of R is, the greater the degree of clustering. When the value of R is 0, the elements are completely agglomerated. When the value of R is 1, the elements are randomly distributed. When the value of R is greater than 1, the elements tend to be spatially evenly distributed.

2.2.2. Kernel Density Estimation

Kernel density estimation is applied to explore the spatial clustering of commercial and residential spaces in a specific region by investigating their spatial variation in POI data density [38]. The quartic polynomial kernel density function was used in this study, as follows:

$$\hat{\lambda}_h(p) = \sum_{i=1}^n \frac{3}{\pi h^4} \left(1 - \frac{(p - p_i)^2}{h^2}\right)^2 \quad (3)$$

where p is the position of the point to be estimated, h is the radius with p as the center of the circle, $\hat{\lambda}_h(p)$ is the value of kernel density estimation in position p , and p_i is the i 'th commercial or residential POI with p as the center and h as the radius of the circle. The value of h affects the smoothness of the density estimation. The area with the highest estimated kernel density is the core clustering area of commercial or residential spaces, followed by the secondary core clustering.

2.2.3. Spatial Correlation

Spatial correlation analysis is applied to conduct a comparative study of the raster images of commercial spaces (or a certain commercial format) and residential spaces to portray the macroscopic correlation characteristics of commercial and residential spaces of the whole study area [39], and is calculated as follows:

$$Corr_{ij} = \frac{\sum_{k=1}^N (Z_{ik} - \mu_i)(Z_{jk} - \mu_j)}{(N - 1)\delta_i\delta_j} \quad (4)$$

where $Corr_{ij}$ is the correlation coefficient, Z is the image element value, i, j is the raster layer, μ is the mean value of image elements, N is the number of image elements, k is the specific image element, and δ is the standard deviation. $Corr_{ij}$ can be used to measure the interdependency relationship between two raster layers, which ranges from -1 to 1 . The more $|Corr_{ij}|$ tends toward 1 , the higher the degree of correlation, and the less $|Corr_{ij}|$ converges to 0 , the lower the degree of correlation.

2.2.4. Honeycomb Grid Analysis

The spatial association relationship between commercial and residential spaces can be analyzed from two perspectives: macroscopic overall association degree and microscopic local association type. The macroscopic association degree measures the overall spatial proximity of the commercial and residential spaces. The local association type is the spatial combination mode of the commercial and residential spaces in a specific region. The honeycomb grid method can both measure the macroscopic association degree and local association type between commercial and residential spaces [40]. In this study, a hexagonal honeycomb grid is laid out in the study area to analyze the spatial association between commercial and residential spaces within each hexagon. The specific indicators of the honeycomb grid analysis are inclusion rate of commercial space (*IRCS*) and ratio of commercial and residential space (*RCRS*).

(1) Through multiple experiments, honeycombs with 500 m and 1000 m radii are laid out in the study area. *IRCS* is obtained by counting the number of honeycombs with residential communities with commercial service facilities in Beijing (Figure 3a), and is calculated as follows:

$$IRCS = \frac{n_{cr}}{n_r} \times 100\% \quad (5)$$

where n_r is the number of honeycombs containing residential communities and n_{cr} is the number of residential honeycombs containing commercial service facilities. *IRCS* is used to characterize the degree of integration between commercial spaces (of a certain commercial format) and residential spaces. The higher the *IRCS* value is, the higher the probability that there are commercial and residential spaces in the same honeycomb grid, the higher the degree of integration of commercial and residential spaces, and the closer their association. When a low association exists on a small scale (500 m radius), whereas a high association exists on a medium scale (1000 m radius), the commercial space is both attractive and repulsive to residential spaces. Specifically, attractiveness is manifested as a consumption association, while the repulsive force is manifested as a negative environmental externality of the commercial space.

a Inclusion rate of commercial space (*IRCS*)

b Ratio of commercial and residential space (*RCRS*)

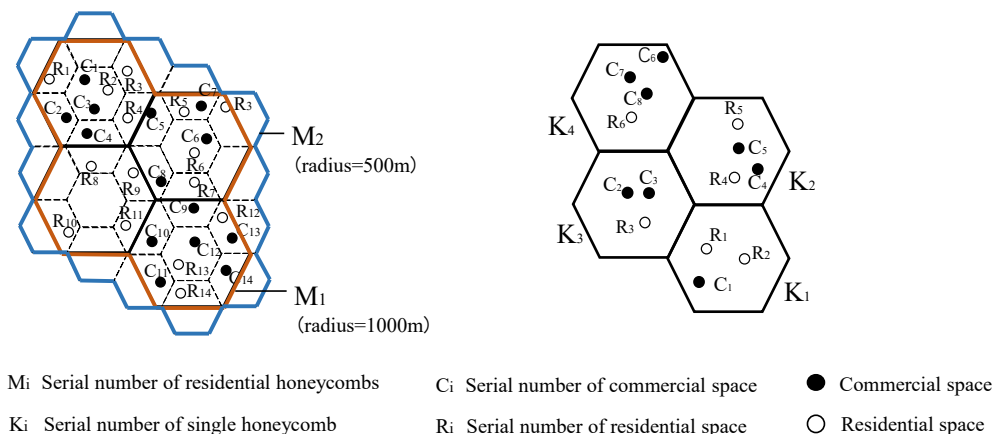


Figure 3. Indicator description of the honeycomb grid analysis (a,b).

(2) A 1000 m radius honeycomb is applied, and the numbers of residential communities and commercial service facilities (of a certain commercial format) within a single honeycomb are counted. Then, the *RCRS* in a single honeycomb can be calculated as follows:

$$RCRS = \frac{N_c + N_r}{N_r} \quad (6)$$

where N_c is the number of commercial service facilities (of a certain commercial format) in a single honeycomb and N_r is the number of residential communities in a single honeycomb.

As shown in Figure 3b, the RCRSs of honeycombs K1, K2, K3, and K4 appear in the order of $K4 > K3 > K2 > K1$. RCRS is used to classify the association type of commercial and residential spaces.

3. Spatial Structure of Commercial and Residential Spaces

3.1. Spatial Distribution Characteristics

The distribution of each commercial format in Beijing from the inside out can be found in Figure 4. The proportions of catering services, specialty stores, and shopping malls in all commercial formats of each circle gradually decrease in the outward direction. The proportions of living services and supermarkets appear to increase first and then decrease, with living services accounting for the highest proportion in the central area and supermarkets accounting for the highest proportion in the inner suburb. The proportion of convenience stores shows a trend of first decreasing and then increasing, with the lowest proportion in the central area. The proportions of recreation and leisure services and agricultural markets show a trend of increasing in the outward direction, with the highest proportions in the outer suburb. In general, the commercial spaces of catering services, living services, and specialty stores need to be near the living and shopping areas of customers, and these three commercial formats have the widest spatial distribution within the city. Commercial formats with very high market demand (e.g., convenience stores, supermarkets, and agricultural markets) are also distributed widely. Recreation and leisure services, which include a variety of subdivided commercial formats, present a particularly obvious trend of suburbanization. Commercial formats present in small numbers with large service areas (e.g., shopping malls, home building materials markets, and home appliance and electronics stores) are distributed dispersedly.

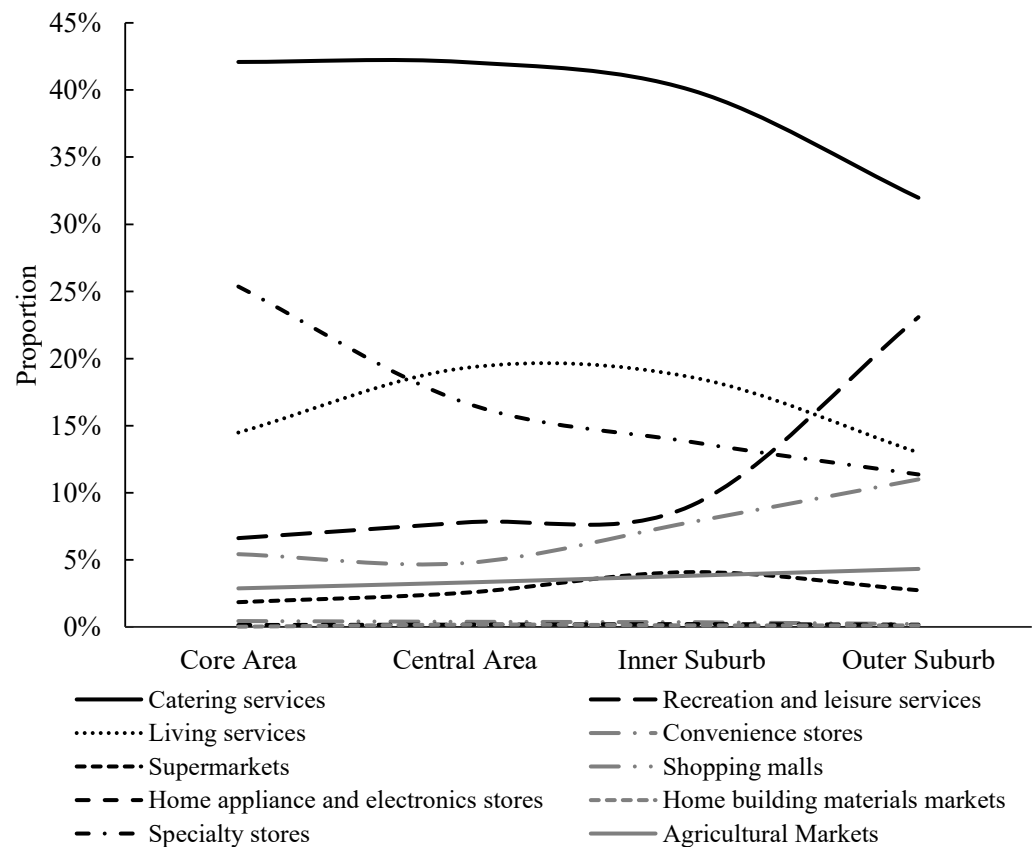


Figure 4. Spatial circular distribution of the various commercial formats in Beijing.

The nearest neighbor ratios of commercial and residential spaces are 0.2001 and 0.2767, respectively (Table 3), indicating that the commercial and residential spaces are characterized by obvious clustering distributions. In addition, the nearest neighbor ratio of each

commercial format is less than 1, with aggregation degrees from high to low of specialty stores > catering services > home building materials markets > living services > agricultural markets > shopping malls > convenience stores > home appliance and electronics stores > supermarkets > recreation and leisure services. The nearest neighbor ratios of specialty stores, catering services, home building materials markets, living services, and agricultural markets are less than the average ratio of all formats. This demonstrates that commercial formats with a high frequency of daily consumption have a high degree of spatial clustering. These commercial formats appear either close to residents' needs (e.g., catering services and living services) or with specialized characteristics (e.g., specialty stores, home building materials markets, and agricultural markets). In contrast, other commercial formats, such as durable goods consumption (e.g., home appliance and electronics stores), extensive daily consumption (e.g., supermarkets and convenience stores), diversified demand and strong experiential consumption (e.g., recreation and leisure services) and comprehensive consumption (e.g., shopping malls), present relatively weak degrees of spatial clustering.

Table 3. Average nearest neighbor analysis of commercial spaces in Beijing.

Commercial Formats	Average Nearest Neighbor Distance (m)	Nearest Neighbor Ratio	Z Score	p-Value	Clustering Degree Ranking
Specialty stores	51.25	0.142	−323.64	0.00	1
Catering services	48.50	0.194	−480.88	0.00	2
Home building materials markets	551.20	0.204	−26.41	0.00	3
Living services	70.06	0.224	−104.12	0.00	4
Agricultural markets	182.28	0.234	−135.40	0.00	5
Shopping malls	549.16	0.284	−40.70	0.00	6
Convenience stores	183.92	0.301	−167.23	0.00	7
Home appliance and electronics stores	890.6	0.303	−28.52	0.00	8
Supermarkets	299.09	0.340	−107.13	0.00	9
Recreation and leisure services	182.47	0.358	−187.58	0.00	10

3.2. Spatial Clustering Characteristics

The overall commercial space in Beijing presents a spatial distribution pattern of “strong monocentric in the core area—contiguously agglomerated in the central area—scattered polycentric in the inner and outer suburbs” (Figure 5). Specifically, some of the regional level, municipal level, and district level commercial centers represent the core clustering areas. Several commercial centers at the district level and community level are secondary core clustering areas. The commercial space in the inner and outer suburbs was mainly concentrated in the center of each district, presenting a multicenter point-like distribution along the main roads. The clustering of residential space is smaller than that of commercial space, presenting a spatial distribution pattern of “contiguous clustering within the central area—multiple groups along the main roads in the inner suburb”. Specifically, the core clustering of residential spaces is distributed within the Third Ring Road of Beijing.

The spatial clustering of each commercial format appears differentiated (Figure 6). Specifically, the spatial distribution of catering services is similar to that of all commercial service facilities. The suburbanization of recreation and leisure services is particularly obvious and is characterized by primary and secondary high-density clustering within the central area. A large-scale scattered distribution of recreation and leisure services can also be found in the inner and outer suburbs. The spatial distribution of living services is similar to that of catering services, with the regional commercial center of Beijing as the core clustering. Convenience stores are widely distributed, and the distribution is characterized by contiguous clusters within the central area and polycentric clusters in the inner and

outer suburbs. The distribution of supermarkets is characterized by contiguous clustering in the central area, planar distribution in the inner suburb, and point-like clustering in the outer suburb. Shopping malls, present in small numbers, are widely distributed within the central area. The shopping malls are mostly located in the commercial center, and their distribution pattern is characterized by primary and secondary cores within the central area and discrete points in the inner and outer suburbs. The spatial distribution of home appliance and electronics stores is characterized by multicenter and multipoint discreteness. Home building materials markets, with a small quantity, large service areas, high product transportation costs, and strong professionalism, are concentrated in the central area and the inner suburb. The distribution of home building materials markets is characterized by a point-like distribution around the Third Ring and Fourth Ring Road. Specialty stores need to be located close to the large-scale commercial district. They are highly clustered in the regional-level and municipal-level commercial centers. Their spatial distribution pattern in the inner and outer suburbs is the same as those of catering services, living services, and shopping malls. The spatial distribution of the agricultural markets is characterized by multicentric clustering in the core area and the central area and scattered clustering in the inner and outer suburbs. Compared with other formats, the core clustering area of the agricultural markets is wider.

The high-, middle- and low-grade residential spaces present differentiated spatial patterns in Beijing (Figure 7). High-grade residential space is characterized by high clustering in the core area and a decrease in the central area. Middle-grade residential spaces are concentrated in the central area and part of the core area, presenting an annular clustering area around the core area in the Third and Fourth Ring Road of Beijing and decreasing on both sides. The low-grade residential space presents a spatial pattern of clustering along the main traffic roads from the core area outwards. A point-like clustering area is also formed in the inner and outer suburbs. In general, influenced by the residents' preference for residential space in the north of the city in Beijing, high- and middle-grade residential spaces are mostly concentrated in the north within the Fourth Ring Road. In contrast, the density of low-grade residential spaces in the south of the city is higher.

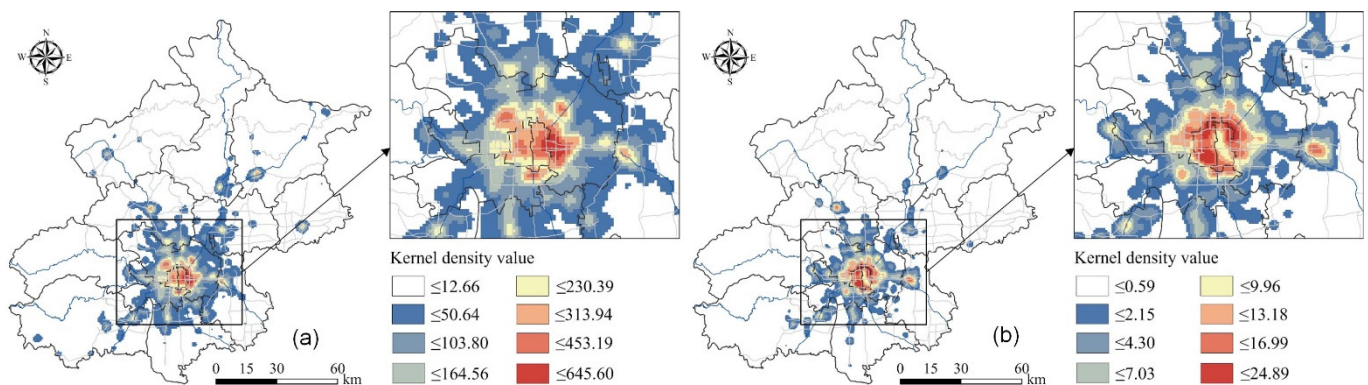


Figure 5. Estimated kernel density of commercial and residential spaces in Beijing: (a) the overall commercial spaces, and (b) the overall residential spaces.

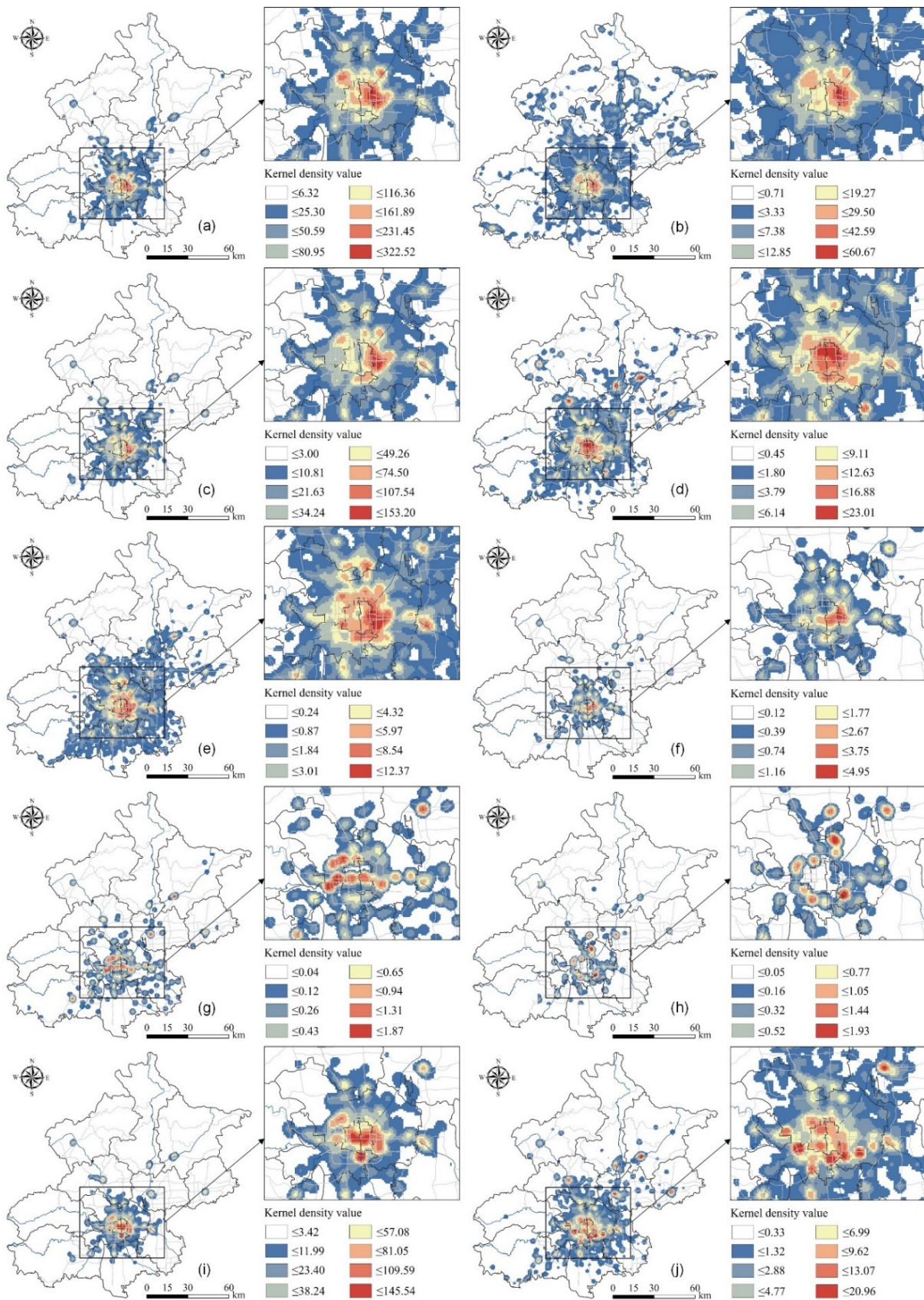


Figure 6. Estimated kernel density of various commercial formats in Beijing: (a) catering services, (b) recreation and leisure services, (c) living services, (d) convenience stores, (e) supermarkets, (f) shopping malls, (g) home appliance and electronics stores, (h) home building materials markets, (i) specialty stores, and (j) agricultural markets.

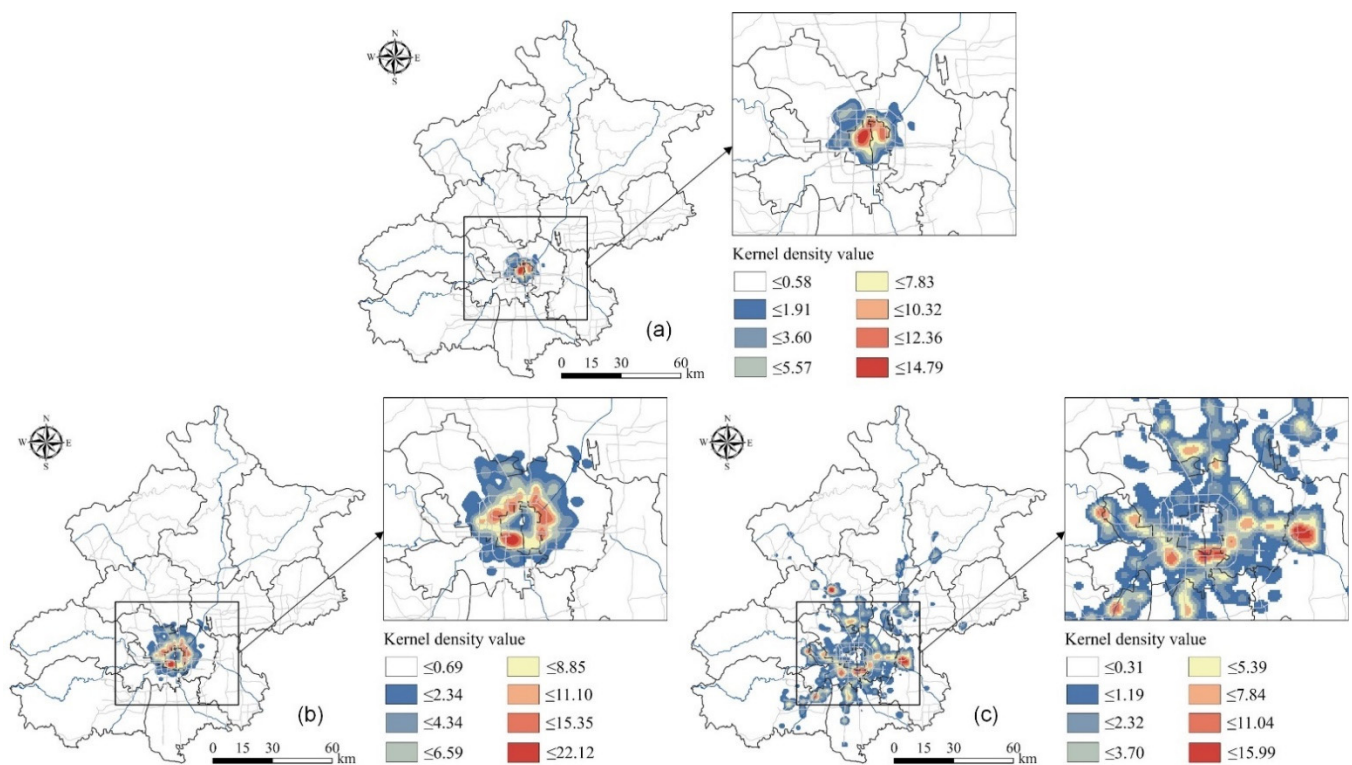


Figure 7. Estimated kernel density of various residential spaces in Beijing: (a) high-grade residential spaces, (b) middle-grade residential spaces, and (c) low-grade residential spaces.

4. Association Relationship between Commercial and Residential Spaces

4.1. Spatial Correlation of Commercial and Residential Spaces

The coefficient of spatial correlation between commercial and residential spaces in Beijing is 0.865 (Table 4). This indicates that there is a close spatial correlation between them and that the distribution of commerce is highly attached to that of residences. Obvious differences exist between the spatial correlation coefficients of each commercial format and residences, ranging from high to low: catering services, living services, specialty stores, recreation and leisure services, supermarkets, convenience stores, shopping malls, agricultural markets, home appliance and electronics stores, and home building materials markets. With a high frequency of daily consumption and residents' demands of close proximity, catering services (0.864) and living services (0.832) have the highest spatial correlation coefficients with residential spaces. Their spatial clustering patterns are closer to that of residential spaces (Figure 5). Taking household consumption as the target market, supermarkets (0.812) have a relatively high correlation with residential spaces. In comparison, the correlation between convenience stores (0.787) and residential spaces is smaller, which is mainly attributed to the slow development of foreign-invested convenience stores in Beijing before 2016. With a small quantity and scattered distribution, shopping malls (0.763) present a low correlation with residential spaces. To facilitate the transportation and trading of agricultural products, some large agricultural markets are located in the inner suburb, resulting in a low correlation (0.721) with residential spaces. The commercial formats with the sale of durable goods such as home appliances and electronics stores (0.612) and home building materials markets (0.365) are mostly scattered in areas far from residential areas, and their correlation coefficients are the lowest.

Of the high-, middle- and low-grade residential spaces, the middle-grade presents the highest spatial correlation with catering services (0.794), recreation and leisure services (0.826), living services (0.742), supermarkets (0.715), shopping malls (0.710), home appliance and electronics stores (0.561), and specialty stores (0.723). High-grade residential spaces have the highest spatial correlation with convenience stores (0.742). Low-grade residential

spaces have the highest spatial correlation with home building materials markets (0.381) and agricultural markets (0.616), and their spatial correlations with other commercial formats are relatively low. This result indicates that the commercial formats of daily consumption and shopping are mostly distributed around the Third and Fourth Ring Road of Beijing and have the highest spatial correlation with the middle-grade residential spaces in this area (Figure 7). Moreover, convenience stores meet the needs of the residents in high-grade residential spaces. With a large demand for usable areas and low environmental requirements, large-scale markets such as home building materials markets and agricultural markets are closely related to low-grade residential spaces.

Table 4. Spatial correlation analysis of commercial and residential spaces in Beijing.

Commercial Formats	All Residences	High-Grade Residences	Mid-Grade Residences	Low-Grade Residences
All commercial formats	0.865	0.953	0.804	0.599
Catering services	0.864	0.651	0.794	0.559
Recreation and leisure services	0.814	0.595	0.826	0.495
Living services	0.832	0.506	0.742	0.640
Convenience stores	0.787	0.742	0.689	0.595
Supermarkets	0.812	0.519	0.715	0.645
Shopping malls	0.763	0.532	0.710	0.498
Home appliance and electronics stores	0.612	0.488	0.561	0.465
Home building materials markets	0.365	0.039	0.276	0.381
Specialty stores	0.818	0.690	0.723	0.543
Agricultural markets	0.721	0.575	0.608	0.616

4.2. Association Degree of Commercial and Residential Spaces

As shown in Table 5, the inclusion rates of all commercial spaces based on honeycomb grids with radii of 1000 m and 500 m are 98.1% and 91.1%, respectively, indicating a high degree of integration and a close consumption association of the commercial and residential spaces. The association degree of each commercial format with the residential space varies considerably. Within a 500 m radius of residential space, the IRCS are arranged from high to low as follows: catering services, living services, recreation and leisure services, convenience stores, specialty stores, supermarkets, agricultural markets, shopping malls, home appliance and electronics stores, and home building materials markets. However, within a 1000 m radius of residential spaces, the IRCS values are ordered from high to low as follows: home building materials markets, home appliance and electronics stores, shopping malls, supermarkets, agricultural markets, recreation and leisure services, convenience stores, specialty stores, living services, and catering services. It can be seen that the spatial association of commercial and residential spaces is characterized by the “integration of daily consumption and separation of nondaily consumption”. Additionally, the trend of “mesoscale integration” of residences with commercial formats with a larger service radius and nondaily consumption (e.g., shopping malls, home appliance and electronics stores, and home building materials markets) is particularly obvious [40]. The trend of “mesoscale integration” is relatively obvious for residences with commercial formats occupying larger areas (e.g., supermarkets and agricultural markets). The trend of “small-scale integration” is obvious for residences with commercial formats with a high frequency of daily consumption and small usable areas (e.g., food services, living services, convenience stores, and specialty stores) [40]. The spatial integration degrees between high-grade residential spaces and all commercial formats, except for home building materials markets, are high, followed by middle-grade residential spaces. Low-grade residential spaces have the lowest spatial integration degrees with each commercial format. High-grade residential spaces are surrounded by abundant commercial formats, while the integration level of low-grade residential spaces and commercial spaces is relatively poor.

Table 5. The inclusion rate of commercial spaces based on honeycomb grids.

Commercial Formats	All Residences			High-Grade Residences		Middle-Grade Residences		Low-Grade Residences	
	500 m	1000 m	500–1000 m Growth Rate	500 m	1000 m	500 m	1000 m	500 m	1000 m
All commercial formats	91.1%	98.1%	7.6%	99.1%	100.0%	97.4%	99.6%	89.8%	97.9%
Catering services	84.0%	95.5%	13.7%	99.1%	100.0%	93.4%	98.9%	81.7%	95.3%
Recreation and leisure services	73.4%	91.6%	24.9%	97.6%	100.0%	90.8%	98.5%	69.0%	91.1%
Living services	75.2%	88.2%	17.3%	96.7%	100.0%	89.1%	97.1%	72.1%	87.6%
Convenience stores	71.0%	88.4%	24.5%	93.4%	99.0%	84.0%	96.4%	67.6%	88.0%
Supermarkets	57.3%	79.1%	37.9%	82.9%	96.1%	73.1%	92.3%	54.3%	78.3%
Shopping malls	16.8%	32.2%	91.1%	34.6%	65.7%	27.2%	54.0%	15.3%	31.8%
Home appliance and electronics stores	7.9%	18.7%	136.8%	11.9%	35.3%	10.3%	28.8%	7.7%	18.8%
Home building materials markets	4.2%	11.6%	176.2%	2.8%	11.8%	3.7%	13.9%	4.5%	12.2%
Specialty stores	62.5%	77.5%	24.0%	92.4%	98.0%	79.9%	92.0%	58.7%	77.0%
Agricultural markets	52.7%	72.4%	37.2%	83.9%	96.1%	67.7%	88.3%	49.7%	71.3%

4.3. Association Type of Commercial and Residential Spaces

Due to the obvious differences in the service ranges and residents' daily demands of each commercial format, the mean \pm standard deviation of the RCRS in a single honeycomb is used as the criterion to define the association type of commercial and residential spaces. Three association types of commercial-lagging-residential, commercial-residential-coordinated, and commercial-advanced-residential are classified (Table 6).

Table 6. Classification standard of association relationships between commercial and residential spaces.

Commercial Formats	Commercial-Lagging-Residential	Commercial-Residential-Coordinated	Commercial-Advanced-Residential
All commercial formats	1.00–9.49	9.49–72.68	72.68–873.00
Catering services	1.00–5.28	5.28–30.20	30.20–416.00
Recreation and leisure services	1.00–1.89	1.89–5.91	5.91–77.00
Living services	1.00–2.84	2.84–15.14	15.14–209.00
Convenience stores	1.00–1.87	1.87–5.96	5.96–45.00
Supermarkets	1.00–1.48	1.48–3.08	3.08–19.00
Shopping malls	1.00–1.02	1.02–1.25	1.25–8.00
Home appliance and electronics stores	1.00–1.01	1.01–1.14	1.14–8.50
Home building materials markets	1.00–1.01	1.01–1.13	1.13–6.00
Specialty stores	1.00–1.60	1.60–12.64	12.64–169.00
Agricultural markets	1.00–1.15	1.15–4.59	4.59–64.50

The overall association relationship between commercial and residential spaces in Beijing is dominated by the coordinated type (accounting for 56.20%), with the lagging type accounting for 29.46% and the advanced type accounting for 14.34% (Table 7). Specifically, the coordinated-type association is mainly distributed within the Fourth Ring Road of Beijing. The advanced-type association is concentrated within the central area (around the commercial centers within the Fourth Ring Road and near the traffic arteries outside the Fourth Ring Road) and in the centers of the suburban districts. The lagging-type association is widely distributed in the inner and outer suburbs (Figure 8).

Table 7. The proportion of each association type of commercial and residential spaces.

Commercial Formats	Commercial-Lagging-Residential	Commercial-Residential-Coordinated	Commercial-Advanced-Residential
All commercial formats	29.46%	56.20%	14.34%
Catering services	37.34%	47.80%	14.86%
Recreation and leisure services	31.91%	52.45%	15.63%
Living services	47.16%	49.61%	14.34%
Convenience stores	39.28%	44.57%	16.15%
Supermarkets	44.57%	38.37%	17.05%
Shopping malls	68.60%	18.35%	13.05%
Home appliance and electronics stores	81.27%	9.04%	9.69%
Home building materials markets	88.37%	3.88%	7.75%
Specialty stores	36.18%	52.07%	11.76%
Agricultural markets	31.91%	56.46%	11.63%

Significant spatial heterogeneity exists in the association relationships of different commercial formats and residential spaces in Beijing. (1) For catering services, the coordinated type, lagging type, and advanced type account for 47.80%, 37.34%, and 14.86%, respectively. Specifically, the coordinated type and advanced type are mainly concentrated within the central area. Outside of the Fourth Ring Road of Beijing, the advanced type is also distributed along the main roads and spreads to district centers in the inner and outer suburbs. Affected by the clustered distribution of government agencies, the lagging type is distributed within the Second Ring Road of Beijing. (2) For recreation and leisure services, the coordinated type, lagging type, and advanced type account for 52.45%, 31.91%, and 15.63%, respectively. The advanced type is mainly concentrated outside of the Fourth Ring Road of Beijing, and several honeycombs of the advanced type are located in the center of each district in the inner and outer suburbs. (3) For living services, the proportion of coordinated type (49.61%) and lagging type (47.16%) is approximately the same. The area within the Fourth Ring Road is dominated by the coordinated type. Some lagging type exists in the core area. The distribution of advanced type is similar to that of catering services. (4) For convenience stores, the coordinated type accounts for a relatively large proportion (44.57%). The proportion of the lagging type in the northern part within the Fourth Ring Road is relatively large. More advanced type exists for convenience stores in the southern part of the city than in the northern part of the city. (5) For supermarkets, the lagging type accounts for a relatively large proportion, 44.57%, mainly concentrated in the inner and outer suburbs and the Xicheng District within the core area. The coordinated type (38.37%) is mainly concentrated in the central area and the inner suburb, and the advanced type (17.05%) is concentrated in the district-level and community-level commercial centers. (6) For shopping malls, the lagging type accounts for 68.60%. The coordinated type (18.35%) and the advanced type (13.05%) are concentrated within the Fourth Ring Road. Specifically, the advanced type is concentrated in the regional level and municipal level commercial centers. Although the lagging type accounts for the largest proportion in the inner and outer suburbs, one or two honeycombs of the advanced type are still formed in the center of each district in the inner and outer suburbs. (7) For home appliances and electronics stores, the lagging type (accounting for 81.27%) is the most widely distributed in Beijing. The advanced type is scattered as a dotted layout mainly distributed in the areas around Shijingshan Road and in the centers of each district. (8) For the home building materials markets, the lagging type accounts for 88.37%. The coordinated type (accounting for 3.88%) is scattered in the central area. The advanced type (accounting for 7.75%) is concentrated in the southern part of the central area around the Fourth Ring Road. Additionally, one or two honeycombs of advanced type are formed in the center of each district in the inner and outer suburbs. (9) For specialty stores, the coordinated type (accounting for 52.07%) is widely distributed in the central area and part of the inner suburb. The advanced type (accounting for 11.76%) is scattered in the vicinity of the main roads. Because some spe-

cialty stores for cultural and creative products are clustered in the Palace Museum and other tourist attractions, the advanced type is also concentrated in these areas. (10) For agricultural markets, the coordinated type with residential space (accounting for 56.46%) is widely distributed. The advanced type (accounting for 11.63%) is mostly distributed in the range from the Fourth Ring Road to the border of the central area and is also concentrated in the southern region of the city.

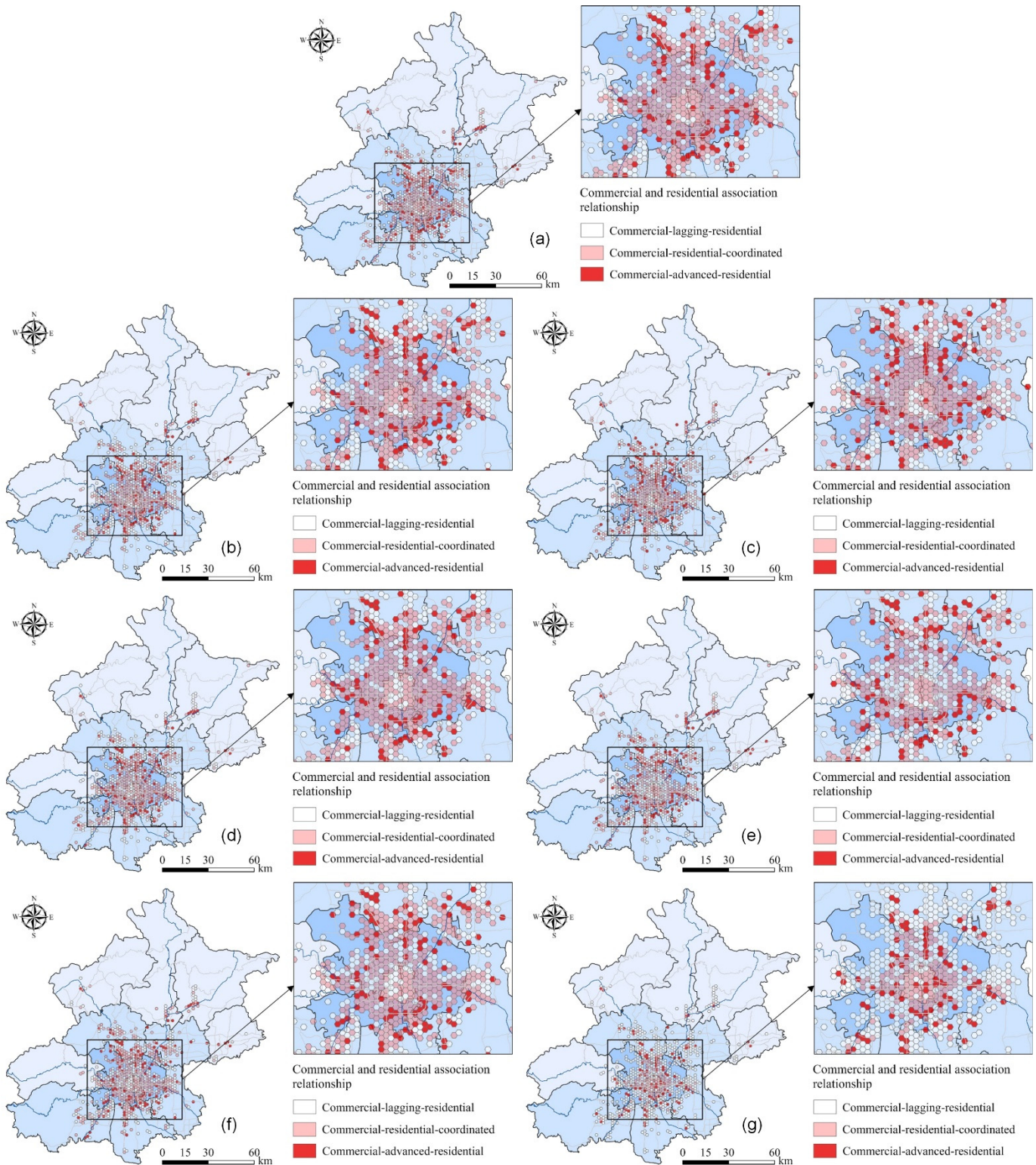


Figure 8. Cont.

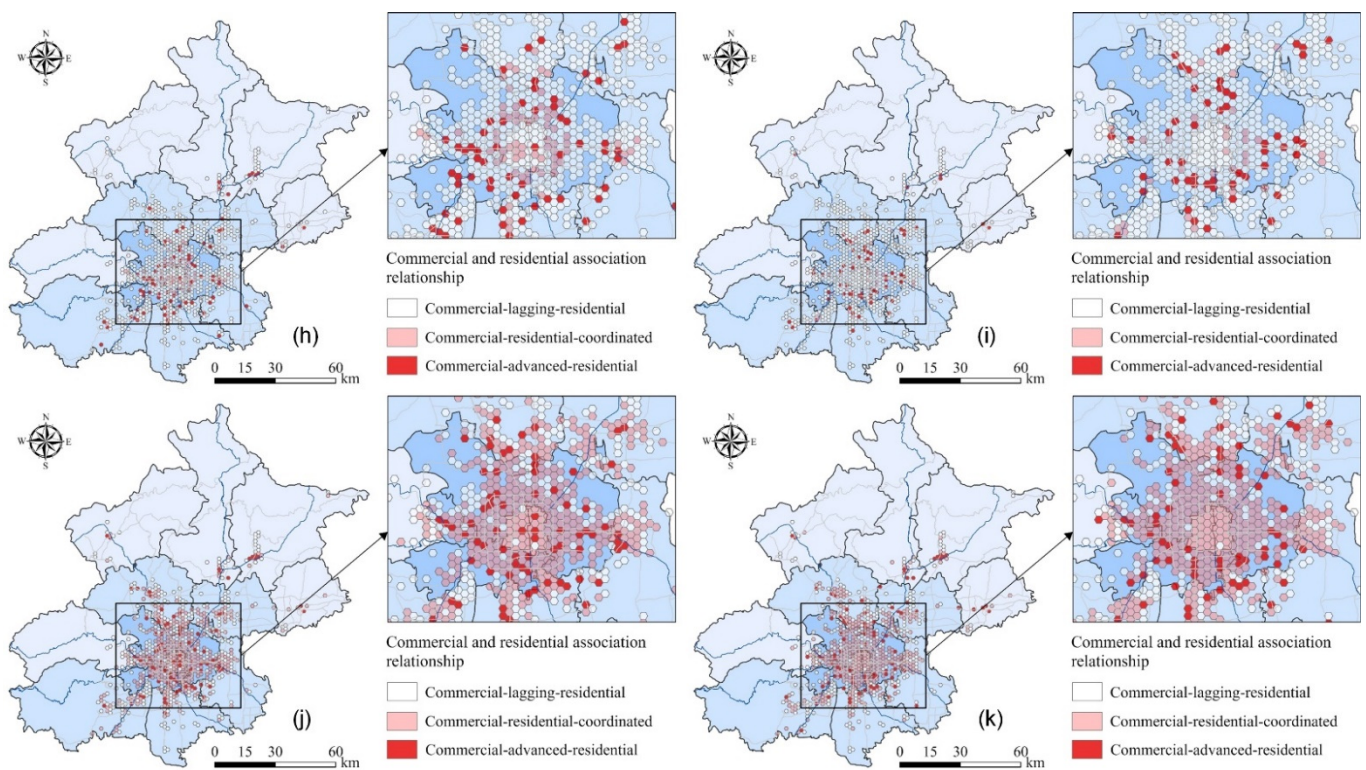


Figure 8. The distribution of association relationships between commercial and residential spaces in Beijing: (a) the overall commercial spaces, (b) catering services, (c) recreation and leisure services, (d) living services, (e) convenience stores, (f) supermarkets, (g) shopping malls, (h) home appliance and electronics stores, (i) home building materials markets, (j) specialty stores, and (k) agricultural markets.

5. Discussion and Conclusions

(1) The commercial spaces in Beijing are characterized by a pattern of “strong mono-center in the core area, contiguous clustering in the central area, scattered polycenters in the inner and outer suburbs”. However, the residential spaces are characterized by a pattern of “contiguous clustering within the central area, multicluster distribution along the main traffic in the inner suburb”. The differences in consumer groups, consumption frequencies, and service scopes of different commercial formats lead to differentiated clustering patterns. The commercial formats with a high frequency of daily consumption, resident demands of close proximity (catering services and living services), and specialized goods (specialty stores, home building materials markets, and agricultural markets) present a relatively high degree of spatial clustering. In comparison, commercial formats with durable goods (home appliance and electronics stores), daily extensive consumption (supermarkets and convenience stores), diversified demands and strong experiential consumption (recreation and leisure services), and comprehensive consumption (shopping malls) exhibit a relatively weak clustering. The high-, middle- and low-grade residential spaces present differentiated spatial patterns. Most of the high- and middle-grade residential spaces are concentrated in the northern part of the city within the Fourth Ring Road, while low-grade residential spaces are mostly concentrated in the southern part of the city, with a higher density.

(2) The distribution of commercial spaces in Beijing is strongly related to that of residential spaces. These two types of spaces are highly integrated, maintaining a relatively close consumption spatial association. However, there are obvious differences in the association degrees of different commercial formats and residential spaces, showing a characteristic of “integration of daily consumption and separation of nondaily consumption”. Specifically, the commercial formats of nondaily consumption (e.g., shopping malls, home appliance and electronics stores, and home building materials markets), which occupy large areas,

sell products with high durability, low consumption frequency, and large service scope, and they have a low degree of association with residential spaces. These commercial formats present a particularly obvious trend of “mesoscale integration” with residential spaces. Commercial formats with high daily consumption frequency occupying small areas (e.g., catering services, living services, and specialty stores) are highly related to residential spaces, presenting an obvious trend of “small-scale integration”. For the residential spaces, except for the home building materials markets and agricultural markets, the degree of spatial association between the commercial spaces and high- and middle-grade residential spaces is higher than that of low-grade residential spaces. High-grade residential spaces are surrounded by abundant commercial formats. The integration level of low-grade residential spaces and commercial spaces is relatively poor.

(3) Significant spatial heterogeneity exists in the association relationship between commercial and residential spaces in Beijing. The overall association relationship is dominated by the commercial-residential-coordinated type within the Fourth Ring Road. The commercial-advanced-residential type is concentrated in the central area and each district center in the inner and outer suburbs. The commercial-lagging-residential type is widely distributed in the inner and outer suburbs. The association relationship between commercial and residential spaces presents a significant difference in the northern and southern regions of Beijing divided by Chang’an Street. The commercial formats (e.g., convenience stores, supermarkets, shopping malls, and home building material markets) lagging residential spaces are mainly distributed in the northern region of Beijing. In the southern region of the city, the proportion of commercial-advanced-residential spaces is relatively high. For the various commercial formats, catering services, recreation and leisure services, specialty stores, and agricultural markets have a relatively good association relationship with residential spaces. Commercial formats with a high frequency of daily consumption (e.g., living services, convenience stores, and supermarkets) lag behind residential spaces. However, large-scale comprehensive and specialized commercial formats (e.g., shopping malls, home appliances and electronics stores, and home building materials markets) appear to lag behind the residential spaces in a wide area. It can be concluded that the commercial space follows the residential space at the initial stage of suburbanization, and then the residential space follows the commercial space. The development of residential space would attract commercial formats such as catering services, recreation and leisure services, specialty stores, and agricultural markets. The development of commercial formats with a high frequency of daily consumption (e.g., living services, convenience stores, and supermarkets) would contribute to the development of residential space.

(4) Based on urban big data, this paper investigates the relationship between commercial and residential spaces from the perspective of the spatial association of elements, which enriches the current research on the matching degree of commercial and residential spaces reflected by residents’ consumption travel time and distance. Currently, in the “entity-online” dual consumption era, online consumption facilitates the diverse needs of physical consumption, which has a very large impact on physical commercial formats. As a result, the goals and intentions of residents’ physical consumption have changed accordingly. Residents tend to choose experiential consumption that enhances their sense of well-being, and commercial formats that focus on physical experience still have very large market potential. The spatial heterogeneity of the association relationship between various commercial formats and residential spaces in Beijing not only reflects the dislocation of commercial and residential spaces in some areas, but it also represents the adjustment of, and changes in, physical commercial formats under the impact of online consumption. Taking the current changes in residents’ consumption preferences into account, and combining the spatial pattern of the association relationship between commercial and residential spaces, it is imperative to strengthen the layouts of daily consumption commercial formats (e.g., convenience stores and agricultural markets) and physical experience formats (e.g., recreation and leisure services) that lag behind the residential space. This paper can

help to promote the coordinated development of commercial and residential spaces in the context of the new era.

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References

- Bell, D. The hospitable city: Social relations in commercial spaces. *Prog. Hum. Geogr.* **2007**, *31*, 7–22. [[CrossRef](#)]
- Zhou, S.; Lin, G.; Yan, X. The relationship among consumer's travel behavior, urban commercial and residential spatial structure in Guangzhou, China. *Acta Geogr. Sin.* **2008**, *63*, 395–404.
- Kobashi, T.; Choi, Y.; Hirano, Y.; Yamagata, Y.; Say, K. Rapid rise of decarbonization potentials of photovoltaics plus electric vehicles in residential houses over commercial districts. *Appl. Energy* **2022**, *306*, 118142. [[CrossRef](#)]
- Oosterbaan, C.; Arku, G.; Asiedu, A.B. Conversion of residential units to commercial spaces in Accra, Ghana: A policy dilemma. *Int. Plan. Stud.* **2012**, *17*, 45–66. [[CrossRef](#)]
- Francke, M.K.; Van de Minne, A. The hierarchical repeat sales model for granular commercial real estate and residential price indices. *J. Real Estate Financ.* **2017**, *55*, 511–532. [[CrossRef](#)]
- Zhang, W.; Li, W.; Zhang, C.; Hanink, D.M.; Liu, Y.; Zhai, R. Analyzing horizontal and vertical urban expansions in three East Asian megacities with the SS-coMCRF model. *Landsc. Urban Plan.* **2018**, *177*, 114–127. [[CrossRef](#)]
- Na, J.; Zhu, J.; Zheng, J.; Di, S.; Ding, H.; Ma, L. Cellular automata based land-use change simulation considering spatio-temporal influence heterogeneity of light rail transit construction: A case in Nanjing, China. *ISPRS Int. J. Geo-Inf.* **2021**, *10*, 308. [[CrossRef](#)]
- Zhao, J.; Xiao, Y.; Sun, S.; Sang, W.; Axmacher, J.C. Does China's increasing coupling of 'urban population' and 'urban area' growth indicators reflect a growing social and economic sustainability? *J. Environ. Manag.* **2022**, *301*, 113932. [[CrossRef](#)] [[PubMed](#)]
- Weltevreden, J.W.J. Substitution or Complementarity? How the Internet Changes City Centre Shopping. *J. Retail. Consum. Serv.* **2007**, *14*, 192–207. [[CrossRef](#)]
- Farag, S.; Schwanen, T.; Dijst, M.; Faber, J. Shopping online and/or in store? A structural equation model of the relationships between e-shopping and in-store shopping. *Transp. Res. A Policy Pract.* **2007**, *41*, 125–141. [[CrossRef](#)]
- Zeng, W.; Li, L.; Huang, Y. Industrial collaborative agglomeration, marketization, and green innovation: Evidence from China's provincial panel data. *J. Clean. Prod.* **2021**, *279*, 123598. [[CrossRef](#)]
- Liu, X.; Zhen, F.; Zhang, M.; Xi, G. Research review of online shopping impact on personal travel and urban retail space and implications. *Prog. Geogr.* **2015**, *34*, 48–54.
- Pang, R.; Zhao, Z.; Wang, W.; Yang, X. The spatial layout of residence since the reform of housing system in Changchun. *Sci. Geogr. Sin.* **2013**, *33*, 435–442.
- Jiang, L.; Feng, C. The study of residential differentiation in Changsha based on the social-spatial perspective. *Econ. Geogr.* **2015**, *35*, 78–86.
- Zhou, C.; Luo, R.; Dai, D. Evolution and mechanism of the residential spatial structure from 2000 to 2010 in Guangzhou. *Geogr. Res.* **2015**, *34*, 1109–1124.
- Shen, Q.; Li, C.; Liu, Z.; Hu, S.; Liu, Q. Spatial pattern of public service facilities space and living space in Changchun. *Geogr. Res.* **2018**, *37*, 2249–2258.
- Zhou, S.; Hao, X.; Liu, L. Validation of spatial decay law caused by urban commercial center's mutual attraction in polycentric city: Spatio-temporal data mining of floating cars' GPS data in Shenzhen. *Acta Geogr. Sin.* **2014**, *69*, 1810–1820.
- Li, Y.; Chen, X. Spatio-temporal evolution and spatial agglomeration of Harbin commercial center. *Geogr. Res.* **2017**, *36*, 1377–1385.
- Wang, D.; Wang, C.; Xie, D.; Zhong, W.; Wu, M.; Zhu, W.; Zhou, J.; Li, Y. Comparison of retail trade areas of retail centers with different hierarchical levels: A case study of east Nanjing Road, Wujiaochang, Anshan Road in Shanghai. *Urban Plan. Forum* **2015**, *223*, 50–60.
- Wu, K.; Zhang, H.; Wang, Y.; Wu, Q.; Ye, Y. Identify of the multiple types of commercial center in Guangzhou and its spatial pattern. *Prog. Geogr.* **2016**, *35*, 963–974.

21. Wang, F.; Gao, X.; Xu, Z. Identification and classification of urban commercial districts at block scale. *Geogr. Res.* **2015**, *34*, 1125–1134.
22. Gobillon, L.; Selod, H.; Zenou, Y. The mechanisms of spatial mismatch. *Urban Stud.* **2007**, *44*, 2401–2427. [[CrossRef](#)]
23. Zhou, J.; Wang, Y.; Cao, G.; Wang, S. Jobs-housing balance and development zones in China: A case study of Suzhou industry park. *Urban Geogr.* **2017**, *38*, 363–380. [[CrossRef](#)]
24. Lin, G. The relationship between spatial structure of commercial types and real estate in urban central district—A case study of Yuexiu in Guangzhou city. *Areal Res. Dev.* **2009**, *28*, 49–56.
25. Lin, G. Characteristic and transformation of consumer space in a context of residential suburbanization—A case study for Guangzhou. *Sci. Geogr. Sin.* **2009**, *29*, 353–359.
26. Ye, Q.; Cao, S.; Nie, C. Research on the correlativity of urban residential and commercial spatial structure evolution based on GIS—Case of Changsha. *Econ. Geogr.* **2012**, *32*, 65–70.
27. Wu, D.; Zhou, S. The matching pattern of housing-shopping space based on residents' daily shopping behaviors: A case study of communities in Guangzhou, China. *Sci. Geogr. Sin.* **2017**, *37*, 228–235.
28. Xue, B.; Xiao, X.; Li, J.; Xie, X.; Lu, C.; Ren, W. POI-based spatial correlation of the residences and retail industry in Shenyang city. *Sci. Geogr. Sin.* **2019**, *39*, 442–449.
29. Meltzer, R.; Schuetz, J. Bodegas or Bagel Shops? Neighborhood Differences in Retail and Household Services. *Econ. Dev. Q.* **2012**, *26*, 73–94. [[CrossRef](#)]
30. Domínguez, C.S. Retailing establishments: A competitive analysis of commercial formats from the consumers' profiles and perceptions. *J. Retail. Consum. Serv.* **2007**, *14*, 297–308. [[CrossRef](#)]
31. Jiang, S.; Zhang, H.; Wang, H.; Zhou, L.; Tang, G. Using restaurant POI data to explore regional structure of food culture based on cuisine preference. *ISPRS Int. J. Geo-Inf.* **2021**, *10*, 38. [[CrossRef](#)]
32. Chen, W.; Liu, L.; Liang, Y. Retail center recognition and spatial aggregating feature analysis of retail formats in Guangzhou based on POI data. *Geogr. Res.* **2016**, *35*, 703–716.
33. Hao, F.; Wang, S.; Feng, Z.; Yu, T.; Ma, L. Spatial pattern and its industrial of commercial space in Changchun based on POI data. *Geogr. Res.* **2018**, *37*, 366–378.
34. Lin, G.; Song, P.; Li, R.; Yang, F. The progress and prospect of commercial geography in the context of consumption society. *Hum. Geogr.* **2019**, *34*, 80–89.
35. Gao, Y.; Yang, Q.; Liang, L.; Zhao, Y. Spatial pattern and influencing factors of retailing industries in Xi'an based on POI data. *Sci. Geogr. Sin.* **2020**, *40*, 710–719.
36. Liu, Z.; Liu, S. Polycentric development and the role of urban polycentric planning in China's mega cities: An examination of Beijing's metropolitan area. *Sustainability* **2018**, *10*, 1588. [[CrossRef](#)]
37. Do Carvalho Monteiro, R.L.; Pereira, V.; Costa, H.G. A multicriteria approach to the human development index classification. *Soc. Indic. Res.* **2018**, *136*, 417–438. [[CrossRef](#)]
38. Mitchell, A. *The ESRI Guide to GIS Analysis, Volume 2: Spatial Measurements and Statistics*; ESRI Press: Redlands, CA, USA, 2005.
39. Getis, A.; Ord, J.K. The analysis of spatial association by use of distance statistics. *Geogr. Anal.* **1992**, *24*, 189–206. [[CrossRef](#)]
40. Wang, D.; Fang, B.; Li, X. Evolution and mechanism analysis of residential-industrial spatial association in Yangzhou city in recent 100 years. *Geogr. Res.* **2020**, *39*, 1295–1310.