Measures of Spatial and Demographic Disparities in Access to Urban Green Space in Harbin, China

Qian Xie and Ming Lu

School of Architecture, Harbin Institute of Technology, Key Laboratory of Cold Region Urban and Rural Human Settlement Environment Science and Technology, Ministry of Industry and Information Technology, Harbin 150006, China

Correspondence should be addressed to Ming Lu; hitlm@126.com

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1. Introduction

Urban green spaces (UGSs) are the primary places within cities that offer inhabitants opportunities to engage in recreational and physical activities [1]. Research has found that access to UGS can effectively contribute to enhancing health and well-being, especially in terms of physical activity improvement, stress relief, weight loss, mortality reduction, and prevention of chronic diseases [2–7]. Since UGS is a public environmental and natural resource that can supply urban ecosystem services and bring great health benefits to urban inhabitants [8, 9], access to UGS has become a popular research topic. Many studies consider green space accessibility a spatial or socioenvironmental justice issue [10–15]. Currently, achieving spatial and social-environmental equity has become the goal of UGS provision, which holds that people should have equitable opportunities to access and benefit from UGS independent of the location of their residence, demographic characteristics, or socioeconomic status [16–19]. However, disparities in access and uneven distribution of UGS in urban areas occur all over the world. Therefore, measuring disparity and inequity becomes a necessary task in an attempt to achieve more equitable UGS provisions within a city. Generally, the task consists of three main processes: (i) measuring accessibility to UGS, (ii) evaluating disparities, and (iii) identifying which groups that have less access to UGS.

Measuring accessibility to UGS is the fundamental step in identifying the disparities. Even though Wolch (2014)
stated that previous studies have not reached a consensus about how to measure accessibility to UGS [2], there are two primary parameters: proximity and quantity of UGS. First, UGS proximity is one of the determining factors of green space access, which is associated with the level of physical activities [20]. More UGS within a walking space positively correlates with park use [7]. A widely used approach for measuring access from the perspective of proximity is to generate a UGS service area (catchment), which is defined as an area influenced by an individual UGS [21]. The service area determines the users who can potentially access the green space based on Euclidean distance and network distance, and the interval of distance depends on the mobility of users. Given that the range of walking speed is 2–10 km/h and the average speed is 4.8 km/h [22], scholars have used a variety of threshold times and distances to define the service area (catchment) of a UGS. Boyle (1983) described how park users tend to visit a park within 500 m of their residence [23], which is approximately a seven-minute walking distance; the European Environment Agency advocates that UGSs be placed within 15 min walking distance from visitors (900–1,000 m) [24]. Similarly, some scholars indicate that 1,000 m is the critical distance between residents and parks because the average leisure walking distance in North America is 2,000 m [25, 26]. Wright Wendel claimed that the size of green space catchment should be associated with its acreage; for large green spaces such as urban parks, the service area (catchment) can be delineated as a 1,600 m buffer [27]. Additionally, Pham and Labbé noted that the critical distance for park visitation can be extended to 2,500 m when considering cycling modes [28].

The second parameter to measure accessibility is UGS quantity, which describes the amount or acreage of UGS to which an individual has access. A low quantity of UGS might result in “green space pressure,” which indicates potential need and congestion [13]. The method for measuring UGS quantity is mainly based on either “population-ratio” or “area-percentage.” However, the two-step floating catchment area (2SFCA) method, an approach that integrates both proximity and quantity parameters to measure access, was proposed by Radke and Mu (2000) [29] and named by Luo and Wang (2003) [30]. The method includes operations that generate both a UGS service area (catchment) and calculate population-ratio. In recent years, many further improvements to this approach have emerged, including enhanced 2SFCA [31], kernel 2SFCA [32], variable 2SFCA [33], and Gaussian S2FCA [34]; the primary differences between them are the considerations of “distance effect” in the function. These methods have been applied to the measures of accessibility to health care and food stores [30–32, 34, 35]. However, studies on UGS accessibility and disparity are still limited, particularly in developing countries.

There are two frequently used measures of disparity in accessibility: spatial disparity and population disparity (i.e., demographic and socioeconomic disparity). Research on spatial disparity focuses on the spatial variation of an accessibility value [36–38]. This value can be calculated based on the 2SFCA method, the implementation of which is often dependent on GIS-based mapping and visualization to identify high or low access regions. However, the 2SFCA method only calculates the value of access; the patterns of spatial variation in the accessibility value cannot always be found and tested. Therefore, this study integrates both 2SFCA method and spatial autocorrelation analysis to measure disparities in access to UGS in Harbin. Not only can these approaches calculate spatial accessibility, but spatial autocorrelation analysis can also test the distribution pattern of disparities and identify high- and low-value clusters. In contrast, studies of population disparity primarily investigate differences in access to UGSs between demographic, racial, social, and economic groups [27, 39, 40]. Racial and ethnic disparities in green space access are widely measured in western contexts because of their racial diversity. However, for countries such as China, which is highly racially homogenous, demographic disparities in UGS access are a greater concern. Therefore, this study focuses on exploring the demographic characteristics associated with accessibility.

Many similar studies that measure spatial or population disparities were conducted in developed countries, particularly in America [36]. However, very few studies have been conducted on that topic in developing countries such as China, and those that exist have focused on studying Southern Chinese cities (i.e., Xing et al. [39] and Li et al. [41]). Studies about Northern China, particularly about northeast Chinese cities, where there are less green infrastructure and greater concerns about UGS access than in the south, have rarely been conducted. Few studies about green spaces in Northeastern Chinese cities have focused on the cooling effects of green spaces [42], and there were still gaps in the research literature on disparities in UGS access. Therefore, this study will pay close attention to one of the representative cities in Northeast China, Harbin. The city has typical urban and sociodemographic characteristics of a Northeastern Chinese city including the spatial identity of industrial cities and urban aging, which makes it a representative case to conduct a survey. The aim of this study is to measure spatial and demographic disparities in UGS access in Harbin, China. There are two research objectives. First, we measured spatial accessibility to UGS using Gaussian-based 2SFCA to identify the distribution pattern of access values. Second, we investigated the correlative relationship between UGS access and demographic variables and evaluated the demographic disparities that urban inhabitants experience.

2. Data and Methods

2.1. Study Area. Harbin, the capital of Heilongjiang Province, is considered one of the political, cultural, and economic centers of Northeastern China. Since the administrative area of Harbin is very large, including two county-level cities, seven counties, and nine administrative districts, we selected the main urban area for investigation. The study area covers 94 residential tracts (jiedao) in four administrative districts (Daoli, Daowai, Nangang, and Xiangfang) and has a population of 2.86 million people according to the 2017 population census. Figure 1 presents the location of the study area; the northern part of the study
area is next to the Songhua River, known as one of China’s seven major rivers. The other three regions in the area are linked to the rural-urban fringe.

2.2. Data Source and Processing. Population data based on the 2017 Harbin Census were obtained from the Harbin Natural Resource and Planning Administration, which collects data at city, district, and residential tract (jiedao) levels. The residential tract serves as the minimum census tract and administrative division at subdistrict level, and its size generally contains several neighborhoods. The information pertaining to each residential tract includes total population, number of births per year, number of deaths per year, migration (number of inflows and outflows), and number of males and females. Based on this information, it is possible to measure the birth rate, mortality, sex ratio, immigration rate, and out-migration rate of each tract.

Meanwhile, 39 UGSs were selected in the study area based on three criteria: first, the UGS is available for various physical, social, and recreational activities; second, the UGS is accessible to the general public; third, the UGS is free of charge. The UGS data are mainly based on the Harbin land use survey conducted by the Harbin Natural Resource and Planning Administration and the vector data for the UGSs have been sourced from Baidu Maps (Baidu, Inc., 10 Shangdi 10th St., Haidian District, Beijing). The land use survey shows the distribution of all types of green spaces in the study area, which include not only the UGSs that people are able to access (such as comprehensive parks, theme parks, and recreation ground) but also the UGSs that are not accessible to the general public (such as private green space, roadside green space, and green buffer). Therefore, the Baidu Maps were used to identify the border of each selected UGS. As the size (acreage) of the green spaces is one of the crucial data in the study, to ensure the accuracy of the information from Baidu Maps, we conducted a three-week field survey in December 2019 to identify the actual border of each green space.

After collection, data on the UGSs and census tracts were integrated and converted to polygons using ESRI ArcGIS 10.5 (Environmental Systems Research Institute, Inc., 380 New York St., Redlands, CA). The acreage and population density of the polygons were also calculated using ArcGIS. A map of the study area was exported to present the population density of residential tracts (pop/km²) and the distribution of UGSs (Figure 2). Overall, higher population density tracts are clustered in the central urban region; peripheral regions in the west, south, and east regions of the urban area have comparatively low densities. UGSs are scattered throughout the study area with a generally random distribution pattern.

2.3. Measures of Spatial Accessibility between Demand and Supply. This study measures UGS access using the Gaussian-based 2SFCA technique, which is an intensified version of the original 2SFCA method proposed by two teams of scholars [29, 30]. This method is an integration of the
Gaussian function into the 2SFCA method, accounting for the “distance decay effect” of park services within a catchment area [36]. The distance decay effect, in this study, indicates the likelihood that the number of people visiting a UGS will decrease when the distance between them and UGSs increases. In comparison to the default position that visitation to UGS is equalized in the catchment, the distance decay function of a Gaussian-based 2SFCA method is more in accordance with the way people visit green spaces in real life.

Gaussian-based 2SFCA operations consist of two steps: calculating the supply-to-demand ratio of each UGS and measuring the accessibility value of each residential tract. In the first step, this study establishes a catchment area of green space $i$ based on threshold time ($t_0$) and searches all demand locations (population tracts) $j$ within the catchment. The population at location $j$ is weighted by a Gaussian function ($G$). The ratio $R$ of the UGS is calculated by summing all the weighted populations that fall into the catchment area. The equation can be written as follows:

$$R_i = \frac{S_i}{\sum_{j \in \{t_{ji} \leq t_0\}} G(t_{ji}, t_0) P_j},$$

where $S_i$ is the capacity of UGS supply at $i$; $P_j$ is the population at residential tracts within the UGS catchment area ($t_{ji} \leq t_0$); $t_{ji}$ is the time cost from green space $i$ (location of supply) to population location $j$ (location of demand); $t_0$ is the threshold time, which determines the size of the UGS catchment area; and $G$ is the Gaussian function, which accounts for the distance decay effect of green space service. The function can be listed as follows:

$$G(t_{ji}, t_0) = \begin{cases} e^{-\frac{1}{2} \left(\frac{t_{ji}}{t_0}\right)^2} - e^{-\frac{1}{2}}, & \text{if } t_{ji} \leq t_0, \\ 1 - e^{-\frac{1}{2}}, & \text{if } t_{ji} > t_0. \end{cases}$$

In the second step of the Gaussian 2SFCA method, we generate a catchment of demand location (population tracts) $k$ based on the time threshold ($t_0$) and search all the green spaces $l$ within the catchment area. Then, the Gaussian function ($G$) is used to weight the ratio $R$ of each UGS; the accessibility value ($A$) of the population tracts $k$ is calculated by summing all the weighted ratios $R$ within the catchment area, the equation for which is
Complexity

\[ A_k = \sum_{l \in \{t_{ij} \leq t_0\}} G(t_{ij}, t_0) R_l, \]  

(3)

where \( l \) is the UGS that falls within the catchment of demand location \( k \), \( R \) is the ratio of green space \( l \), \( G \) is the Gaussian function as described in (2), and the accessibility value \( A_k \) denotes the amount of UGS (in square meters) per capita.

Catchment size is a crucial parameter in the Gaussian 2SFCAs method for determining whether a residential tract can be considered near (accessible) a UGS. The threshold time (or threshold distance) \( t_0 \) between a UGS and a residential tract is the key variable for shaping the catchment size, which is defined by the mobility of park users. Mobility is differentiated from transportation modes, including walking, biking, driving, and public transit. In terms of accessibility to UGSs, active transportation (walking and cycling) is recommended because of the benefits they provide for one’s health and the environment [36]. Thus, in this study, we established catchment size and threshold time based on walking and cycling modes.

This study selected four threshold walking times to define four types of UGS catchment sizes (CSs): CS\(_1\) (\( t_0 = 7 \) min), CS\(_2\) (\( t_0 = 15 \) min), CS\(_3\) (\( t_0 = 20 \) min), and CS\(_4\) (\( t_0 = 30 \) min). Different catchment sizes represent different mobility patterns. CS\(_1\) and CS\(_2\) (range of time cost from 0 to 15 min) represent regular walking distance to access a UGS, and CS\(_3\) and CS\(_4\), where the range of time cost is greater than 15 min, represent a long walking or cycling distance.

3. Results

3.1. Spatial Disparities in Accessibility. This study measures spatial disparities in accessibility across the four catchment sizes using a Gaussian 2SFCAs. The results of the accessibility values were visualized using the exported ArcGIS maps (Figure 3), for which we employed a quantile method to classify the value. The four maps show UGS accessibility corresponding to CS\(_1\), CS\(_2\), CS\(_3\), and CS\(_4\). The values of accessibility signify the quantity of UGS (area in m\(^2\)) per capita, which varies from 0 to 82.39 throughout the study area.

According to Figure 3(a), when \( t_0 = 7 \) min, the majority of residential areas have a low accessibility to UGS. In contrast, a few residential tracts with great accessibility are scattered in the central and western urban areas, some of which have considerably high accessibility values (over 80 m\(^2\)/pop.). When \( t_0 = 15 \) min (Figure 3(b)), a few residential tracts show an increase in accessibility, and residential areas with comparatively higher accessibility values tend to gather in the southwest. However, the vast majority of regions remain access-disadvantaged; when \( t_0 = 20 \) min (Figure 3(c)), more tracts have improved levels of accessibility. Residential areas with relatively greater access nearly form a ring around the central city; when \( t_0 = 30 \) min (Figure 3(d)), approximately half of the residential tracts show a clear improvement in access to UGS; however, the other half remain much less accessible. Tracts with comparatively greater access surround the central city in a cyclic annular, and high-access tracts are located mainly in western and southern suburbs.

3.2. Testing the Spatial Distribution Pattern of Accessibility. Although Figure 3 provides information about access disparities in the study area, the tendency and pattern of the distribution remain unclear and untested. Therefore, in this section, we use spatial autocorrelation to evaluate the distribution of accessibility, which can determine whether selected objects correlate with the surrounding objects in value [43]. Moran’s I is a commonly used approach to test global spatial autocorrelation [44, 45]. Positive and negative Moran’s I indexes correspond, respectively, to a clustered and a dispersed distribution of objects, except when the test accepts the null hypothesis (\( H_0 \)), which indicates that the distribution pattern is random. The z-score and \( p \) value in the test are used to determine whether the alternative hypothesis (\( H_1 \)) should be accepted.

This study uses Moran’s I to test the distribution of accessibility to a UGS (Table 1). The results show that \( H_1 \) is rejected when \( t_0 \) equals 7 min, 15 min, or 20 min, which indicates that the distribution of accessibility is random in these three cases. Only when \( t_0 \) equals 30 min is the result statistically significant. Because of positive Moran’s I index, the accessibility associated with the residential tracts tends to be a clustered distribution.

Global measures of spatial autocorrelation (Moran’s I) investigate the distribution of the access value as a whole within the study area (at the global level) and are not able to provide information about local spatial dependencies. Therefore, this study undertakes hot spot analysis (Getis-Ord \( G^*_I \)), a local spatial autocorrelation tool offered by Getis and Ord [46], to further identify local clusters in the study area. We conducted an analysis of the distribution of accessibility when \( t_0 \) equals 30 min, which is statistically significant in the measure of spatial autocorrelation. In the Getis-Ord \( G^*_I \) method, a z-score is calculated for each residential tract, and a significant z-score with a high positive value indicates that local clusters of tracts are associated with high-access values (hot spot). In contrast, a significant z-score with a low and negative score value denotes a clustering of low accessibility values (cold spot). Based on the z-score, the statistical significance is divided into three confidence levels (90%, 95%, and 99% confidence).

Figure 4 presents the results of Getis-Ord \( G^*_I \) analysis, which clearly identifies cold spots and hot spots of accessibility. The significant spatial clusters of residential tracts with a low accessibility value (cold spot) are located in the center of the research area. In contrast, significant clusters of residential tracts with a high accessibility value (hot spot) were located in northern and western suburbs. The results of other regions of the study area have no statistical significance in Getis-Ord \( G^*_I \) analysis. In general, within the study area, the tracts with low accessibility to UGS tend to cluster in the middle of the city, while high-accessibility areas tend to be located on the urban fringes.
3.3. Evaluating Demographic Disparities in Accessibility. In this section, we attempt to evaluate the demographic disparities by exploring the statistical relationship between demographic characteristics of residential tracts and accessibility to UGS. Bivariate correlation was undertaken to examine accessibility\(^{a-d}\) \((t_0 = 7, 15, 20, \text{ and } 30 \text{ min})\) with each of the eight demographic variables, including population of children under 17 (%), population of elderly individuals over 60 (%), birth rate (%), mortality rate (%), sex ratio, rate of immigration, rate of out-migration, and population density.

Table 2 presents the results of bivariate correlation analysis. First, the sex ratio of residential tracts shows a significant negative correlation with accessibility\(^d\) at a level of 0.01. Because of the way sex ratio is calculated in this instance (females divided by males), the female ratio is

![Figure 3: Spatial accessibility to urban green spaces in Harbin \((t_0 = 7 \text{ min}; t_0 = 15 \text{ min}; t_0 = 20 \text{ min}; t_0 = 30 \text{ min})\).](image)
inversely proportional to accessibility. Moreover, we found that birth rate is positively associated with accessibility, with significant levels of 0.05, 0.05, and 0.01, respectively, which indicates that as birth rate increases, accessibility increase. In contrast, accessibility are all negatively correlated with population density, and the significance levels are 0.05, 0.01, and 0.01, respectively. In addition, other demographic variables related to age and migration, including population of children under 17, population of elderly above 60, rate of immigration, and rate of out-migration, do not have a statistically significant relationship with any of the indicators of accessibility.

4. Discussion

This study used a Gaussian-based 2SFCA method to measure spatial access to UGSs based on four threshold walking times (7, 15, 20, and 30 min). (i) A 7 min walk (approximately 500 m) is defined as “walking distance” by some researchers. However, this study observed that, within a 7 min catchment area, the available amount of UGS for the majority of the residential areas was under 3 m$^2$ per person, much lower than the 9 m$^2$ per person advocated by the WHO. (ii) A 15 min walk (approximately 1,000 m) is considered the critical walking distance to access a park because when people want to reach a

<table>
<thead>
<tr>
<th>Accessibility to UGS</th>
<th>Moran’s I index</th>
<th>Z-score</th>
<th>p value</th>
<th>Distribution</th>
</tr>
</thead>
<tbody>
<tr>
<td>Accessibility, CS1, $t_0 = 7$ min</td>
<td>0.00</td>
<td>0.66</td>
<td>0.51</td>
<td>Random ($H_0$)</td>
</tr>
<tr>
<td>Accessibility, CS2, $t_0 = 15$ min</td>
<td>0.03</td>
<td>1.61</td>
<td>0.11</td>
<td>Random ($H_0$)</td>
</tr>
<tr>
<td>Accessibility, CS3, $t_0 = 20$ min</td>
<td>0.03</td>
<td>1.44</td>
<td>0.15</td>
<td>Random ($H_0$)</td>
</tr>
<tr>
<td>Accessibility, CS4, $t_0 = 30$ min</td>
<td>0.14</td>
<td>4.81</td>
<td>0.00</td>
<td>Clustered ($H_A$)</td>
</tr>
</tbody>
</table>

**Figure 4:** Hot spot analysis (Getis-Ord $G_{ij}^*$) of accessibility ($t_0 = 30$ min) associated with residential tracts.
UGS at a greater distance, they choose transportation modes other than walking. However, even when threshold time is increased to 15 min, only a few residential tracts can achieve a 3.5 m² of UGS per person. The 7 and 15 min catchments accord with walking distance to a UGS; therefore, in general, UGSs in Harbin urban areas are not walking friendly. The supply of green space must be improved in terms of amount and acreage, especially within a short walking distance from citizens’ residences. The catchment sizes of 20 min and 30 min accord with the transportation modes of long-distance walking, running, or cycling. (iii) A 20 min catchment (approximately 1,300 m) has been defined as “cycling distance” by previous researchers [51]. In this study, even though a number of the residential tracts have experienced an increase in access value, more than half of the tracts still have disadvantages in terms of accessibility to UGS. (iv) When it comes to a 30 min catchment, which is nearly the critical distance for cycling, half of the study area has much less access to UGS. Generally speaking, for active transportation modes, such as walking and cycling, many areas of Harbin contain very limited amounts of green space.

The spatial distribution of accessibility can reflect the uneven or unequal distribution of UGS, and this urban problem can be found in many cities worldwide [52]. In this study, accessibility is randomly distributed when the threshold time is less than 20 min. This might be because of the very small number of residential tracts with high accessibility, which are generally distributed in the city with no clear pattern. In contrast, the number of tracts with high-value accessibility dramatically increases when the threshold time is 30 min. At this point, the distribution of accessibility is clustered according to the results of Moran’s I test, and hot spot analysis also indicates that low accessibility tracts are located in the center of the study area, while high accessibility tracts tend to cluster on urban fringes, especially in northern and southern suburbs. On the contrary, the middle of the city has a higher density than the surrounding suburbs according to the population density map (Figure 2). This indicates an uneven distribution and a disproportionate supply of UGS for urban inhabitants.

Previous studies have ascertained that many socio-demographic characteristics are associated with park use [53] and disparities in the supply of UGS occur with demographic and racial differences [54]. This study also explored the relationship between demographic indicators and accessibility to UGS. Because of the dissimilarity between the research method and urban context, the results in this paper differ from those in the others [55, 56].

(i) In terms of gender, very few studies examine the correlation between sex ratio and green space access. One such study was conducted in Brisbane, Australia, which found no significant relationships between the two [55]. However, in the case of Harbin, we found that gender is correlated with accessibility (a 30 min walk) to UGS, where the proportion of females is negatively associated with green space access. This indicates that, for long walking or cycling distances, females have more disadvantages in accessing green spaces than males in the city.

(ii) Previous studies have mentioned the relationship between lack of access and mortality [56]. However, significant correlations between the two were not observed in this study. In contrast, the birth rate, the demographic indicator that relevant studies rarely pay attention to, is found to have a significant positive correlation with UGS accessibility in Harbin, which reflects that, for high birth rate areas, people tend to have greater green space access than in other

### Table 2: Bivariate correlation between accessibility and demographic variables.

<table>
<thead>
<tr>
<th>Demographic variables</th>
<th>Accessibility(^a) ((t_0 = 7) min)</th>
<th>Accessibility(^b) ((t_0 = 15) min)</th>
<th>Accessibility(^c) ((t_0 = 20) min)</th>
<th>Accessibility(^d) ((t_0 = 30) min)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pop. of children under 17 (%)</td>
<td>0.105</td>
<td>0.099</td>
<td>0.012</td>
<td>0.093</td>
</tr>
<tr>
<td>Pop. of elderly above 60 (%)</td>
<td>−0.213</td>
<td>−0.138</td>
<td>−0.111</td>
<td>−0.194</td>
</tr>
<tr>
<td>Gender</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sex ratio (females/male)</td>
<td>−0.315</td>
<td>−0.282</td>
<td>−0.179</td>
<td>−0.369**</td>
</tr>
<tr>
<td>Natural increase in pop.</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Birth rate (%)</td>
<td>0.336</td>
<td>0.350*</td>
<td>0.353*</td>
<td>0.437**</td>
</tr>
<tr>
<td>Mortality rate (%)</td>
<td>0.019</td>
<td>0.006</td>
<td>0.047</td>
<td>−0.077</td>
</tr>
<tr>
<td>Migration</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Rate of immigration (%)</td>
<td>−0.043</td>
<td>−0.122</td>
<td>−0.108</td>
<td>−0.130</td>
</tr>
<tr>
<td>Rate of out-migration (%)</td>
<td>0.280</td>
<td>0.277</td>
<td>0.181</td>
<td>0.190</td>
</tr>
<tr>
<td>Density</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Population density (pop./m²)</td>
<td>−0.425</td>
<td>−0.432*</td>
<td>−0.422**</td>
<td>−0.403**</td>
</tr>
</tbody>
</table>

\(^a\) Statistical significance (* \(p < 0.05\); ** \(p < 0.01\)).
areas in the city. Even though it is difficult to explain this association, we still infer that newlyweds might have an advantage in access to UGSs. In China, marriage is one of the strongest motivations for housing consumption, and newlyweds are one of the primary groups that purchase an additional house [57]. Therefore, in comparison to other social groups, newlyweds will have greater opportunities to choose houses with greater location advantages and higher UGSs accessibility. However, this is an inferential point of view by this study, and further empirical studies are needed to confirm the inference.

(iii) The correlation test demonstrates that density has a negative association with accessibility when the threshold time is greater than or equal to 15 min. This indicates that more populated regions have less access to UGS in the study area, which is consistent with the information from hot spot analysis above. High-density areas in Harbin are mainly found in the central parts of the city, where there is a very limited amount of space for UGS provision. In comparison, low density area in suburbs provides more space for building UGSs, and the people there can access more UGSs than those who live in the middle of the city. However, this situation is not unique, as many studies have indicated that denser urban areas have fewer green spaces [11, 58]. Since areas with higher density reflect a greater need for UGS [59, 60], the result yields a problem of spatial disparity between population demand and UGS supply.

(iv) This study found no significant correlation between age and access to UGS, which indicates that residential tracts with a high proportion of elderly or children do not take more advantage of their access to green spaces in Harbin than others do. In fact, “compensatory equity” suggests that public benefits (such as UGS) should be consistent with need [61, 62] and previous studies have identified that the elderly and children are the disadvantaged groups with the greater need of access to green spaces [63–65]. Provinces in Northeast China have experienced rapid population aging [66]. In Harbin, one of the capital cities in Northeast China, population aging is always a severe issue. According to the 2017 population census, the proportion of people over 60 is 24.9% within the study area, and in some residential tracts, this proportion is even as high as 39.5%. This indicates that seniors in Harbin have high demands for the improvement of access to UGSs. Therefore, this paper suggests that Harbin should concentrate on the needs of seniors and other high-demand groups when planning green space provision.

In conclusion, this study first measured access disparities to UGSs in urban areas of Harbin. In general, the majority of residential areas have much less walking distance UGS access. As regards cycling distance access, even if the access value is enhanced in many areas, a number of residential tracts still cannot meet WHO standards. Furthermore, the result of spatial autocorrelation tests indicates an uneven distribution and a disproportionate provision of UGS to urban residents. Second, this study evaluated the demographic disparities by exploring the statistical relationship between eight demographic indicators of residential area and UGS access with four threshold times. The results show that gender, birth rate, and population density have a significant correlation with green space access, with females in the study having less access than males, areas with high birth rates having more access to UGS, and with population density being negatively associated with UGS access. The findings identify which areas and population groups are disadvantaged, which should help advise policy-makers on future construction and provision of UGS.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

Conflicts of Interest

The authors declare that there are no conflicts of interest.

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