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Article

Urbanization Characteristics of Central Asian Since the New Century

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Abstract: Central Asian countries are the core area of the Belt and Road Initiative (BRI). However, as part of the former Soviet Union, we know very little about the urbanization processes in these countries after the collapse of the Soviet Union. Here, we used land-cover type data, vegetation index data, and grided population data to quantify the urban expansion, urban population changes, and urban environment changes across cities in Central Asia from 2000 to 2020, and took Xinjiang, China as a reference. We found that the urbanization in the study area was uneven. Specifically, the urban expansion in Xinjiang, China and Kazakhstan was faster than in the other countries in the study area. Due to rapid urban expansion, the urban population density in Xinjiang, China decreased from 2000 to 2020, but rapid urban population growth was maintained. Consequently, the pressure of the urban population growth in Xinjiang, China was less than that in the other countries of Central Asia. On average, more than 35% of the urban built-up areas underwent significant greening in Xinjiang, China and more than 3.63 million residents directly benefited from the increase in urban greenness. Although the urban greenness in Xinjiang increased significantly, the overall greenness was still lower than in most cities in Central Asia. In the future, the cities in Xinjiang, China should continue to promote the construction of urban ecological civilization and strengthen their role as a link in the Belt and Road Initiative.

Keywords: urbanization; urban greenspace; urban population; Xinjiang; central Asian countries

1. Introduction

As a heart of the “world island” ^[1], Central Asia links Asia and Europe. After the collapse of the Soviet Union, the five Central Asian countries, i.e., Kazakhstan, Kyrgyzstan, Tajikistan, Turkmenistan and Uzbekistan, have experienced different degrees of economic recession and have been actively exploring a path suitable for their own economic development for the last 30 years ^[2]. In 2013, China proposed the Belt and Road Initiative (BRI), which provides a historic opportunity for the development of the region ^[3]. BRI is the largest infrastructure investment in history, it is expected to boost the economic development of participating countries ^[4] and provide important funds for infrastructure construction ^[5]. In addition, Xinjiang, China, as the core area of the Silk Road Economic Belt, connects the northwestern provinces of China and the five countries in Central Asia, making it an important link in the BRI ^[6].

In the BRI, cities are the main carriers of transportation and logistics, social and economic development, and they are the most important natural complexes ^[7]. Understanding the urbanization characteristics of BRI and quantifying the urbanization process will provide scientific data to support the implementation of the BRI. At present, a large number of studies have used land use data ^[4] and nighttime light data ^[8] to analyze the urban expansion and morphological changes in Xinjiang and Central Asia. Several previous studies have also examined the water resources ^[9] and sustainable development of cities in this region ^[10]. According to the World Bank's income classification ^[11], Xinjiang, China, Kazakhstan, and Turkmenistan are countries/regions with upper-middle-income and Uzbekistan, Kyrgyzstan, and Tajikistan are countries with lower-middle-income. Although they are located in the same arid climate zone, but their levels of urban development are uneven ^[12].

Nevertheless, as developing countries, the data for the cities in the five Central Asian countries are always insufficient ^[13], and the current quantitative research on the cities in Central Asia is still not sufficient ^[12, 14].

This study was designed to address the questions that follow. (1) As a part of the former Soviet Union, how have the cities in Central Asia developed in recent decades? (2) What are the differences in the urbanization characteristics among countries or regions? To address both of these questions, we took Xinjiang, China as a reference, and use the five Central Asian countries as our study area and used remote sensing data and demographic data from 2000 to 2020 to quantify the population changes of 168 cities in the study area. This comprehensive and systematic study provides scientific data in data limited region, and supports sustainable development of the cities in our study area ^[14].

2. Data and methods

2.1. Data

At present, several sets of land cover type datasets with spatial resolutions ranging from 10 m ^[15] to 500 m ^[16] are available for free use by the public worldwide. Since land use data with a 10 m resolution is usually only available for one period, it is not possible to monitor land use changes using these data. In contrast, the 500 m resolution land use data are too coarse to monitor urban details. In this study, we used the GlobeLand30 data with a 30 m resolution to extract the distribution of the built-up urban areas (BUAs). The GlobeLand30 dataset currently contains three periods of data for 2000, 2010, and 2020 with 10 land-cover types ^[17]. The GlobeLand30 data are mainly generated from 30-meter multispectral images. The dataset is available for free globally and can be downloaded from <http://www.globallandcover.com/>. The urban areas were extracted using the artificial surfaces classification (Code 80) in the GlobeLand30 dataset.

The vegetation index were extracted from the MOD13Q1 version 6 product (~250 m) ^[18]. The datasets were generated with 16 days temporal resolution with a 250 m spatial resolution, including two primary vegetation layers and quality layers. Due to the greater sensitivity of the Enhanced vegetation index (EVI) relative to the Normalized Difference Vegetation Index (NDVI) in the high-biomass regions ^[19], we used the EVI datasets to monitor the greenness changes in the urban BUA. These datasets are available from website (<https://e4ftl01.cr.usgs.gov>).

The population data used in this study were grided population data with a spatial resolution of 100 meters obtained from the WorldPop website ^[20]. The population data of WorldPop has been used in research investigation and policy making in countries worldwide (<https://www.worldpop.org/>). The 100 m resolution grided population datasets were downloaded separately for each country and were merged for further analysis.

2.2. Methods

2.2.1. Built-up area expansion

By using the Globeland30 datasets, the distributions of the urban built-up areas in study area in 2000, 2010, and 2020 were extracted. Since the 30-meter resolution was too fine to describe the urban boundaries, the urban built-up area data were resampled to a 300-meter resolution. Finally, the pixels with more than 50% of the urban BUA were identified as urban pixels. According to the urban pixels at a 300-meter resolution in 2020, the BUA pixels adjacent to each other were merged into a single patch. Finally, we detected 272 urban patches larger than 10 km² in our study area.

2.2.2. Urban population growth

The total population of each urban patch was calculated using overlay analysis of the BUA and gridded population data with a 100-meter resolution from WorldPop^[20]. According to the gridded 2020 population data, there were a total of 168 urban patches with populations of more than 10,000 and an urban built-up area greater than 10 km² in our study area. In this study, we took these 168

cities and towns with relatively large populations and relatively large urban built-up areas as the research objects (hereinafter referred to as cities).

In general, the process of urbanization is inevitably accompanied by population growth and expansion of the BUA. If the urban population growth rate (R_{pop}) and the BUA expansion rate (R_{bua}) are coordinated, the urban population growth pressure during the urbanization will be relatively small. In this study, we calculated the ratio (R_{pb}) of R_{pop} to R_{bua} for the 168 cities [21]:

$$R_{pop} = \frac{P_{2020} - P_{2000}}{P_{2020}} \quad (1)$$

$$R_{bua} = \frac{B_{2020} - B_{2000}}{B_{2020}} \quad (2)$$

$$R_{pb} = \frac{R_{pop}}{R_{bua}} \quad (3)$$

where P_{2020} and P_{2000} are the populations of the urban BUA in 2020 and 2000, respectively. B_{2020} and B_{2000} are the urban BUA in 2020 and 2000, respectively. If R_{pb} is greater than 1, the R_{pop} is higher than R_{bua} , and the population growth faster than the urban expansion since the advent of the new century.

2.2.3. Urban greening in built-up areas

Urban greening is an important part of a city [22, 23]. Specifically, in arid and semi-arid regions, urban greening plays a very important role in maintaining the sustainable development of cities [24, 25]. Here, we used the EVI from the MOD13Q1 dataset as the greenness indicator. Then, the maximum greenness of each urban BUA pixel was generated from maximum EVI value (E_{max}) in 23 phases. By using E_{max} , it was possible to avoid interference from the vegetation phenology and represent the best state of urban greening. Finally, the Mann-Kendall method [12, 26] was used to calculate the trend of E_{max} for each urban pixel from 2001 to 2020.

We have a hypothesis that if the E_{max} of an urban pixel increased significantly from 2001 to 2020 ($P < 0.05$), some green spaces [27], new parks, and/or green roof spaces [28] may have been constructed in that pixel, or, the street trees experienced growth. By taking the BUA in 2020 as the latest urban boundary and the BUA in 2000 as the old urban boundary, we calculated the ratios (R_{EVI}) of the greening of the BUAs in the old and new urban boundaries. The calculation formula is as follows [12]:

$$R_{EVI} = \frac{BUA_{green_year}}{BUA_{year}} \quad (4)$$

For 2000, BUA_{green_2000} is the total area of the greening pixels in the old urban boundary; BUA_{2000} is the total BUA within the old urban boundary, and R_{EVI} is the ratio of the significantly greened BUAs in the old city. For 2020, BUA_{green_2020} is the total area of greening pixels in the new urban boundary; BUA_{2020} is the total BUA within the new urban boundary, and R_{EVI} is the ratio of the significantly greened BUAs in the new city.

3. Research results

3.1. Urban expansion

Five of the 168 cities experienced a BUA expansion of over 50 km² from 2000 to 2020 during the study period (purple points in Figure 1a): Urumqi (134 km²), Nur-Sultan (82 km²), Almaty (55.4 km²), Atyrau (54.5 km²), and Yining (52.5 km²). According to the statistical results for the 168 cities, Kazakhstan (2,403 km²) and Uzbekistan (2,187 km²) had the large total BUA in 2020 (Figure 1b). The total BUA in Xinjiang, China was only 1,118 km², which was less than half those of the above two countries. In addition, the total BUAs in Kyrgyzstan, Turkmenistan, and Tajikistan were 378 km², 326 km², and 252 km², respectively. In Central Asia, Kazakhstan, which contained 37 cities, had a total

BUA expansion of 609 km² (Figure 1b). Followed by Xinjiang, China, with a total BUA expansion of its 29 cities reaching 578 km². In Uzbekistan, there were 74 cities located along the Amu Darya and Syr Darya. However, the total BUA expansion area was only 372 km² in Uzbekistan. In addition, the total BUA expansions in Turkmenistan, Kyrgyzstan, and Tajikistan were only 70 km², 57 km², and 36 km², respectively (Figure 1b).

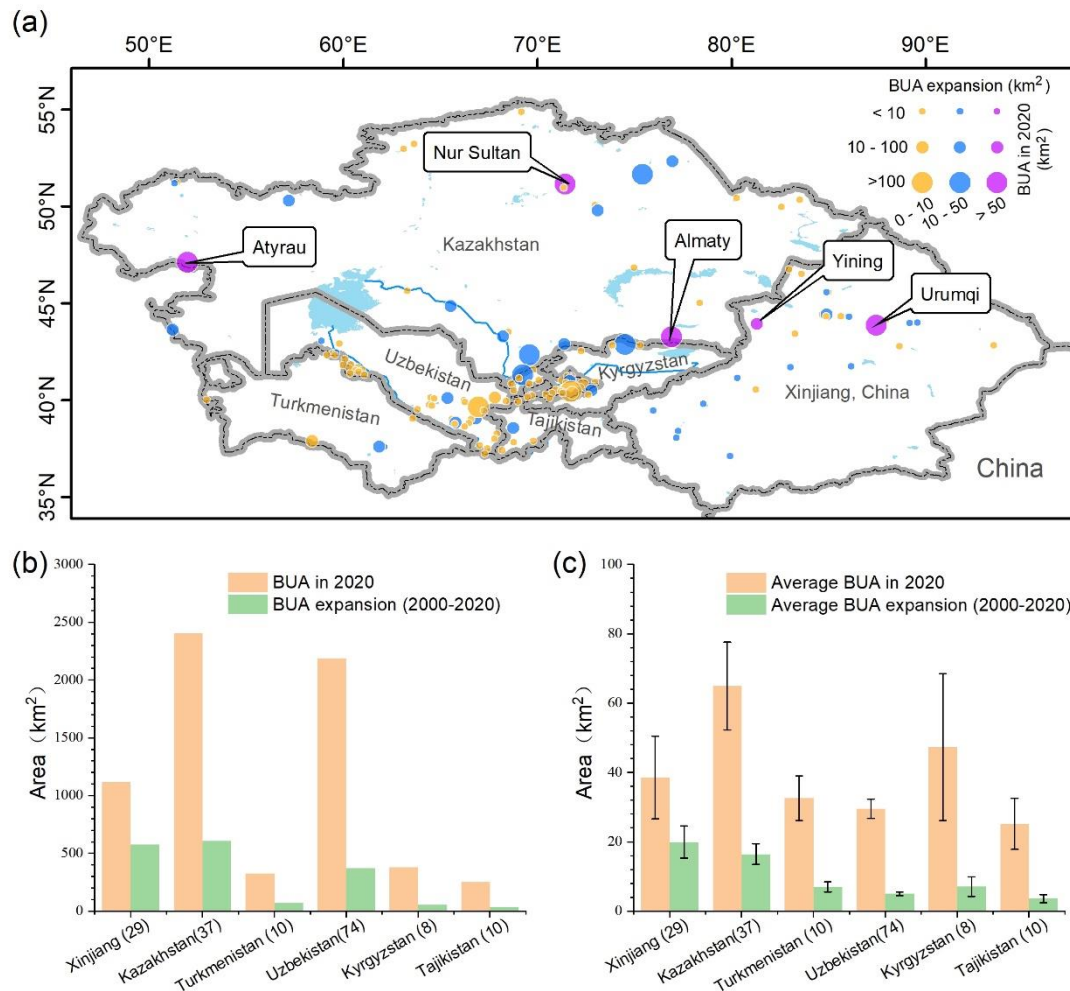


Figure 1. Distribution of cities with urban expansion in Xinjiang and the five Central Asian countries. **(a)** Spatial distribution of the 168 cities with BUA expansion (BUAE) in different levels from 2000 to 2020. The top five cities with BUAE are marked (purple points). **(b)** The bars are the total BUA in 2020 and the total BUAE from 2000 to 2020. **(c)** The bars are the average BUA in 2020 and the average BUAE from 2000 to 2020 for each city. In (c), the error bars denote the standard error of the mean (SEM) for Xinjiang and each country in Central Asia.

In 2020, the average urban BUA in the 37 cities in Kazakhstan was the largest (65 km²), followed by Kyrgyzstan (47 km²), Xinjiang, China (39 km²), Turkmenistan (33 km²), Uzbekistan (30 km²), and Tajikistan (25 km²) (Figure 1c). However, it was found that the BUA expansion rate of Xinjiang, China was the fastest in the region, reaching nearly 20 km² per city, followed by Kazakhstan, with an average BUA expansion rate of 16.5 km² per city. It should be noted that the average BUA expansions of the other four Central Asian countries did not exceed 10 km² per city (Figure 1c). In particular, Uzbekistan had the largest number of cities in the region (74 cities), but only 6 of these cities had BUA expansions of > 10 km².

3.2. Urban population change

Using the grided population data from WorldPop [20], we calculated the population changes for each city in our study area from 2000 to 2020. In 2020, there were five cities in in our study area with populations of greater than 500,000 (Figure 2a): Urumqi (3.17 million), Almaty (1.35 million), Bishkek (1.07 million), Dushanbe (980,000), and Chimkent (580,000). From 2000 to 2020, the populations of all of the 168 cities increased. In particular, 18 cities experienced population growths of greater than 100,000 people (purple points labeled with names in Figure 2a). Table 1 shows the cities with population growths of > 100,000 people and their corresponding BUA expansions. From 2000 to 2020, the three cities with the largest urban population growths were Urumqi (1.654 million), the capital of Xinjiang, China; Almaty (1.348 million), the former capital of Kazakhstan; and Dushanbe (0.435 million), the capital of Tajikistan. Eight cities in Xinjiang, China experienced a population growth of > 100,000 people (Table 1), which was comparable to the population growths of the capitals and core cities of the five Central Asian countries. It should be noted that according to the official statistics, Tashkent, the capital of Uzbekistan, was a large city with a population of 2.13 million. However, according to the dataset from WorldPop, the population of Tashkent in 2020 was only 447,000 people, which is quite different from the official number.

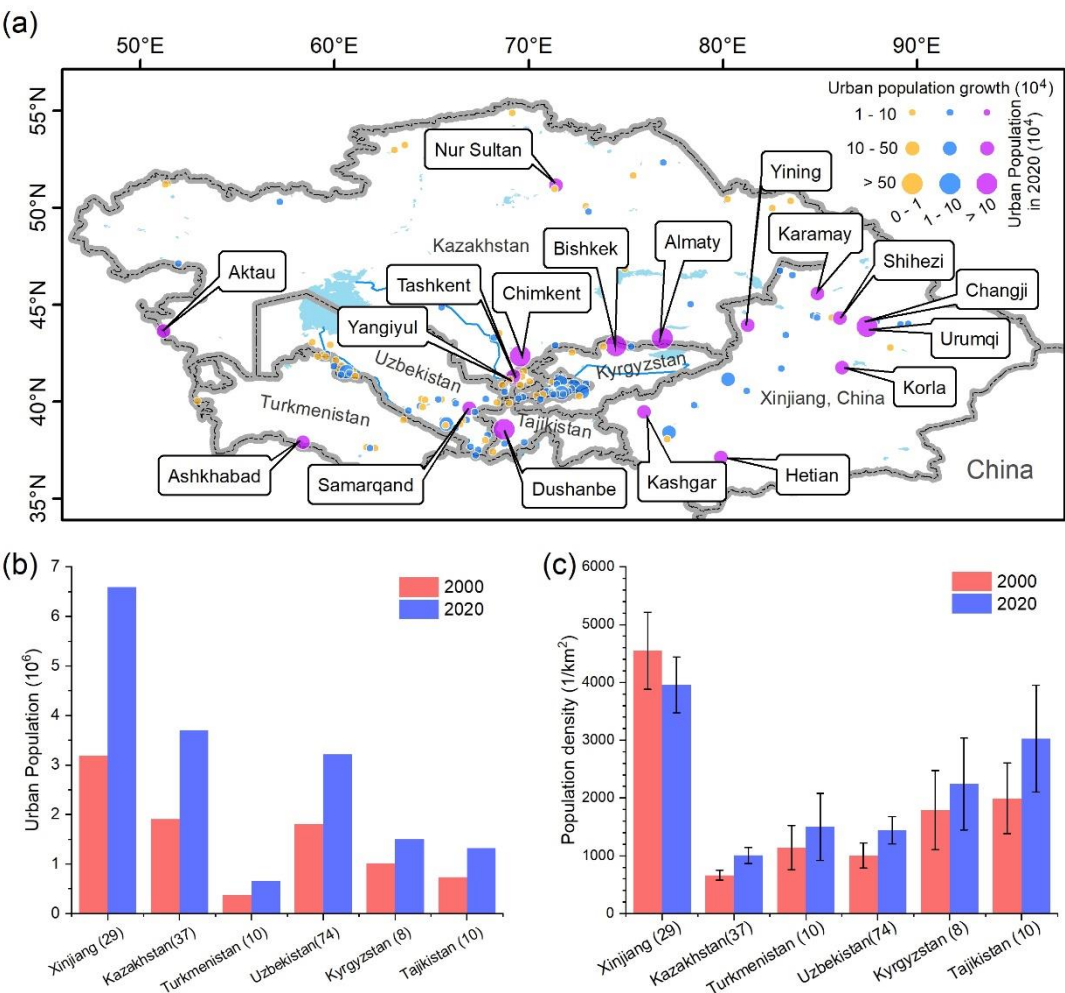


Figure 2. Distribution of cities with urban populations growth. **(a)** Spatial distribution of the 168 cities with urban population growth (UPG) in different levels from 2000 to 2020. The top 18 cities with UPGs of more than 100,000 are labeled with their names (purple points). **(b)** The bars are the total urban populations in 2000 and 2020. **(c)** The bars are the average urban population densities in 2000 and 2020 in each city. In **(c)**, the error bars denote the standard error of the mean (SEM) for Xinjiang and each country in Central Asia.

Table 1. Ranking of the top 18 cities with the largest population growth from 2000 to 2020.

Rank	City Name	Region	BUA in 2020 (km ²)	BUAE (km ²)	Urban population in 2000 (×10 ⁴)	Urban population in 2020 (×10 ⁴)	Urban population growth
1	Urumqi	Xinjiang, China	362	134	151.3	316.7	165.4
2	Almaty	Kazakhstan	286	55	75.5	134.8	59.3
3	Dushanbe	Tajikistan	89	13	54.6	98.1	43.5
4	Bishkek	Kyrgyzstan	191	24	73.3	106.9	33.6
5	Chimkent	Kazakhstan	158	26	24.3	57.5	33.2
6	Kashgar	Xinjiang, China	50	36	16.5	41.2	24.7
7	Ashkhabad	Turkmenistan	69	7	24.5	44.8	20.4
8	Nur Sultan	Kazakhstan	159	82	4.4	24.0	19.6
9	Korla	Xinjiang, China	49	35	14.0	32.8	18.9
10	Yining	Xinjiang, China	46	20	20.9	39.4	18.5
11	Shihezi	Xinjiang, China	142	9	13.0	31.0	18.0
12	Samarqand	Uzbekistan	384	16	14.2	29.1	14.9
13	Tashkent*	Uzbekistan	38	31	32.0	44.7	12.7
14	Hetian	Xinjiang, China	49	28	6.8	18.5	11.7
15	Karamay	Xinjiang, China	21	5	5.0	16.7	11.7
16	Yangiyul	Uzbekistan	29	11	26.7	38.3	11.5
17	Changji	Xinjiang, China	63	47	20.8	31.4	10.5
18	Aktau	Kazakhstan	76	52	1.1	11.1	10.0

* The demographic data for Tashkent City are quite different from the official statistics (<https://worldpopulationreview.com/countries/uzbekistan-population>).

Although Xinjiang, China only contained 29 cities in this study, the total urban population of Xinjiang increased from 3.19 million in 2000 to 6.59 million in 2020, which far exceeded those of the Central Asian countries. Among the five Central Asian countries, the total urban population of the 37 cities in Kazakhstan increased from 1.9 million in 2000 to 3.7 million in 2020. In Uzbekistan, the urban population of the 74 cities increased from 1.81 million to 3.22 million from 2000 to 2020. In contrast, the urban population increments of the other three countries did not exceed one million. In particular, Turkmenistan had the smallest urban population growth among the five Central Asian countries.

We also calculated the population densities of each city in 2000 and 2020. It was found that the average population density in Xinjiang, China decreased from 4,548/km² to 3,957/km² (Figure 2c). In contrast, the average urban population densities of the cities in Central Asian countries increased. In Tajikistan, the average urban population density increased from 1,996/km² to 3,029/km², which was the fastest among the countries in Central Asia. Based on the huge urban population increase in Xinjiang, China (Figure 2b) and the considerable BUA expansion (Figure 1c), it can be concluded that the decrease in the urban population density in Xinjiang was mainly due to the rapid urban expansion. This result indicates that even though the urban population in Xinjiang, China has grown rapidly in the past 20 years, the urban population pressure per unit area has decreased, which has

created conditions favorable to improving the living environment. These results are basically consistent with the results of previous studies [12, 29].

Urban population growth is a key driving force in urbanization. In the study area, the 168 cities were ranked based on the ratio (R_{pb}) of the urban population growth rate to the BUA expansion rate. The results for the countries are illustrated as different colors in Figure 3. Overall, it was found that the R_{pb} values of 145 cities (86%) were greater than 1. This means that the urban expansion in these cities lagged behind the population growth from 2000 to 2020. In particular, the R_{pb} values of 72 cities in Uzbekistan were > 1 , accounting for 97% (72 out of 74) of the cities analyzed in this country. In contrast, in Xinjiang, China, the R_{pb} values of 23 cities were < 1 , indicating that the BUA expansion of the majority of the cities in Xinjiang met the needs of the urban population growth.

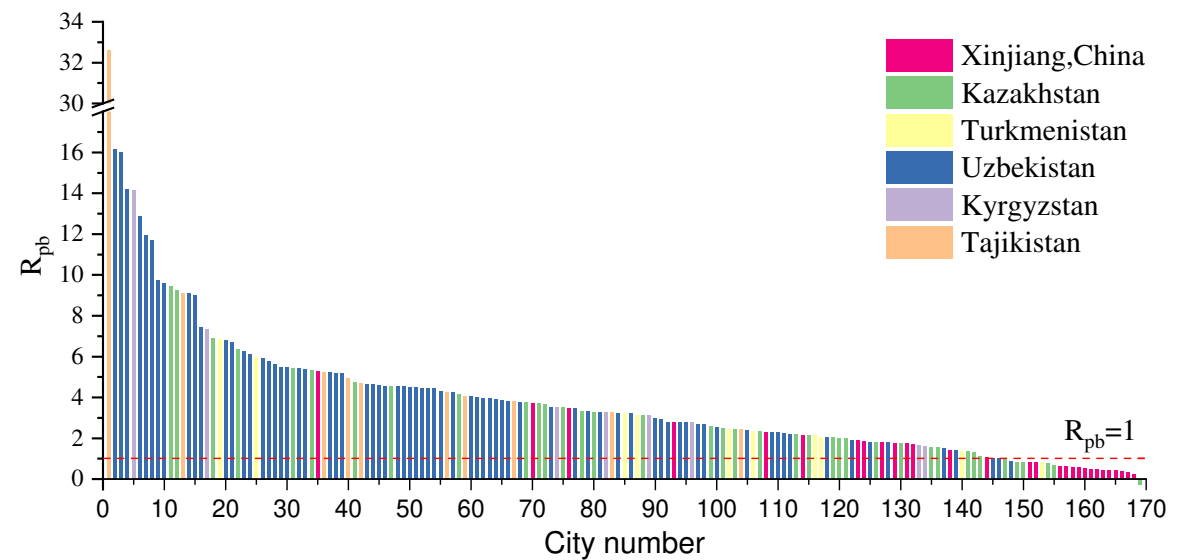


Figure 3. The R_{PB} values refer the ratio of the urban population growth rate to the urban expansion rate, for the 168 cities from 2000 to 2020.

3.3. Urban greening changes

Among the 168 cities, 25 cities contained more than 30% of the urban pixels in the new urban boundaries, indicating significant greening (i.e., $REVI > 0.3$). Among the 168 cities, we screened 10 cities with BUA values of greater than 40 km² (hereinafter referred to as large cities) in 2020 and ranked them by their $REVI$ values (Table 2). The top six cities, i.e., Karamay (0.75), Korla (0.55), Aksu (0.51), Urumqi (0.5), Kuitun (0.44), and Shihezi (0.31), were all located in Xinjiang, China. This means that in Karmay, the urban greening increased significantly within 93% of the old urban boundary and 75% within the new urban boundary. In addition, the $REVI$ values of Temirtau, Karaganda, and Ekibastuz in Kazakhstan and Nukus in Uzbekistan were greater than 0.2. It should be noted that Karmay, Temirtau, and Ekibastuz are important oil and mining industrial cities in China and Kazakhstan. It is a proud achievement that they have improved their urban environment while developing their economies.

Table 2. Ranking of the top 10 large cities* with the largest $REVI$ values.

Rank	City	Region	BUA in 2020 (km ²)	BUAE (km ²)	$REVI$ of BUAs in 2020	$REVI$ of BUAs in 2000
1	Karamay	Xinjiang, China	49.1	28.4	0.75	0.93
2	Korla	Xinjiang, China	48.9	35.1	0.55	0.72
3	Aksu	Xinjiang, China	41.1	19.0	0.51	0.57
4	Urumqi	Xinjiang, China	361.8	133.9	0.50	0.61

5	Kuitun	Xinjiang, China	51.8	27.0	0.44	0.57
6	Timirtau	Kazakhstan	44.5	5.3	0.40	0.43
7	Karaganda	Kazakhstan	45.5	12.5	0.38	0.39
8	Shihezi	Xinjiang, China	45.6	20.1	0.31	0.45
9	Ekibastuz	Kazakhstan	134.9	14.9	0.28	0.28
10	Nukus	Uzbekistan	46.5	11.3	0.22	0.20

*The term large cities refers to cities with urban built-up areas of greater than 40 km² in 2020.

The urban boundaries in 2020 and 2000 were taken as the new urban boundaries and old urban boundaries, respectively. Then, we compared the average $REVI$ values within the new and old urban boundaries of the cities in our study area (Figure 4b). The results show that within both the new or old boundaries, the average $REVI$ values of the cities in Xinjiang were significantly higher than those in the cities in the five Central Asian countries. In particular, within the new urban boundaries, the average $REVI$ value of the cities in Xinjiang, China was remarkable larger than that of the other five Central Asian countries (Figure 4b), reaching an average of 0.35 ± 0.04 , followed by Turkmenistan (0.12 ± 0.04) and Kazakhstan (0.11 ± 0.016). This means that in the past 20 years, the investment in the green infrastructure in these cities was remarkable, and the improvements of the urban ecological environment were also significant.

It should be noted that in Xinjiang, China, the average $REVI$ value within the new urban boundaries was significantly lower than that in the old urban boundaries (Figure 4b). In contrast, the average $REVI$ values within the new and old urban boundaries in the five Central Asian countries did not change significantly. Except for Kazakhstan, the $REVI$ values within the new boundaries of the cities in the other four Central Asian countries were only slightly higher than those within the old boundaries of these cities. The main reason for this is that the average urban expansion in Xinjiang and Kazakhstan in the past 20 years was higher than in the other four countries. Under normal circumstances, the newly expanded urban areas were created by conversion of the surrounding farmland, grassland, and/or forests, resulting in less plant coverage in the new BUA expansion areas. Therefore, in the cities with rapid BUA expansion, the $REVI$ values within the new urban boundaries tended to be lower than those within the old urban boundaries.

Although the above analysis shows that the urban ecological environment in Xinjiang, China has improved significantly in the past 20 years, this progress does not mean that the overall ecological environment in the cities in Xinjiang has surpassed those of the cities in the other Central Asian countries. By comparing the average E_{max} values in the first three years (2001–2003) and the last three years (2018–2020), it was found that the average E_{max} value of the last three years in the old urban boundaries in Xinjiang was still lower than those in the Central Asian countries, except for Turkmenistan (Figure 4c). It should be noted that Xinjiang, China was the only region that experienced significant greening of the BUAs in the old urban boundaries in the past 20 years, while the 3-year average E_{max} values within the old urban boundaries of the other Central Asian countries decreased. (Figure 4c). This may be due to two reasons. First, the urban green infrastructure in Xinjiang, China increased. Second, the type of urban expansion was predominantly infilling expansion in the Central Asian countries [30].

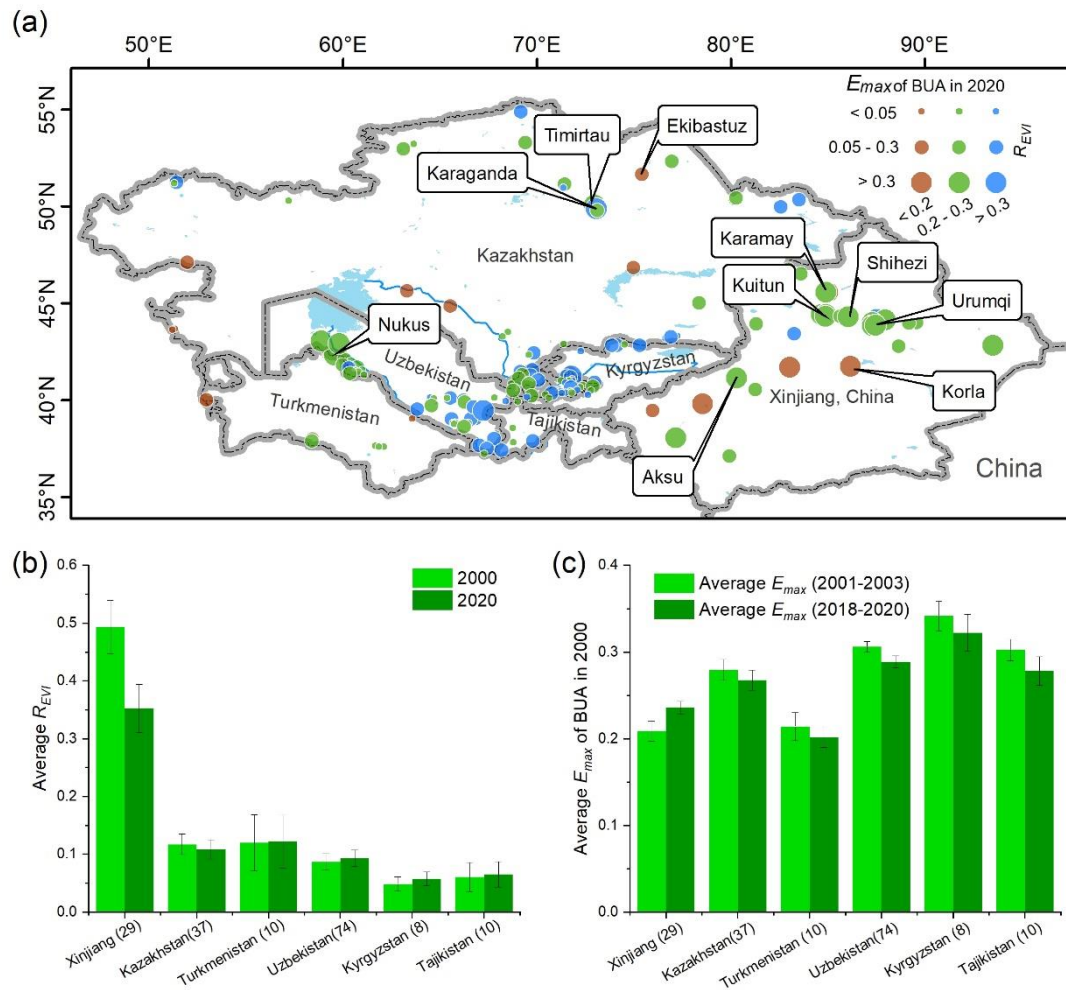


Figure 4. Distribution of cities with greening urban area. **(a)** Spatial distribution of the 168 cities with BUAs in different levels of greening (R_{EVI}). The top 10 cities with greened BUAs are marked with city names. **(b)** The bars are the average R_{EVI} values of the cities in 2000 and 2020. **(c)** The bars are the 3-year average E_{max} values of the BUAs in 2000 in two periods (2001–2003 and 2018–2020). In **(c)**, the error bars denote the standard error of the mean (SEM) for Xinjiang and countries in Central Asia.

3.4. People living in greening BUAs

The increase in urban parks and green spaces can regulate rainwater [31] and alleviate the urban heat island effect [32, 33], and several studies have reported that urban greening is beneficial to human health, including improving mental health [34] and cardiovascular conditions [35], and reduces violence and crime [36]. Therefore, in this study, it was assumed that the residents living in significant greening BUA were likely to benefit from the improvement in the surrounding urban environment. We overlapped the greening BUAs of the 168 cities with the grided population data for 2020 and calculated the population that directly benefited from the greening of the BUAs in each city. As is shown in Figure 5a, among the 168 cities, in 39 cities (green and blue points), more than 10,000 people directly benefited from the greening of the BUAs. These cities were mainly (24 cities) distributed in Xinjiang, China and were sparsely (15 cities) distributed in the five Central Asian countries. Of the eight cities in which a population of more than 100,000 directly benefited (blue points), seven were in Xinjiang, China. Among the Central Asian countries, only Almaty had a population of more than 100,000 that benefited from the greening of the BUAs.

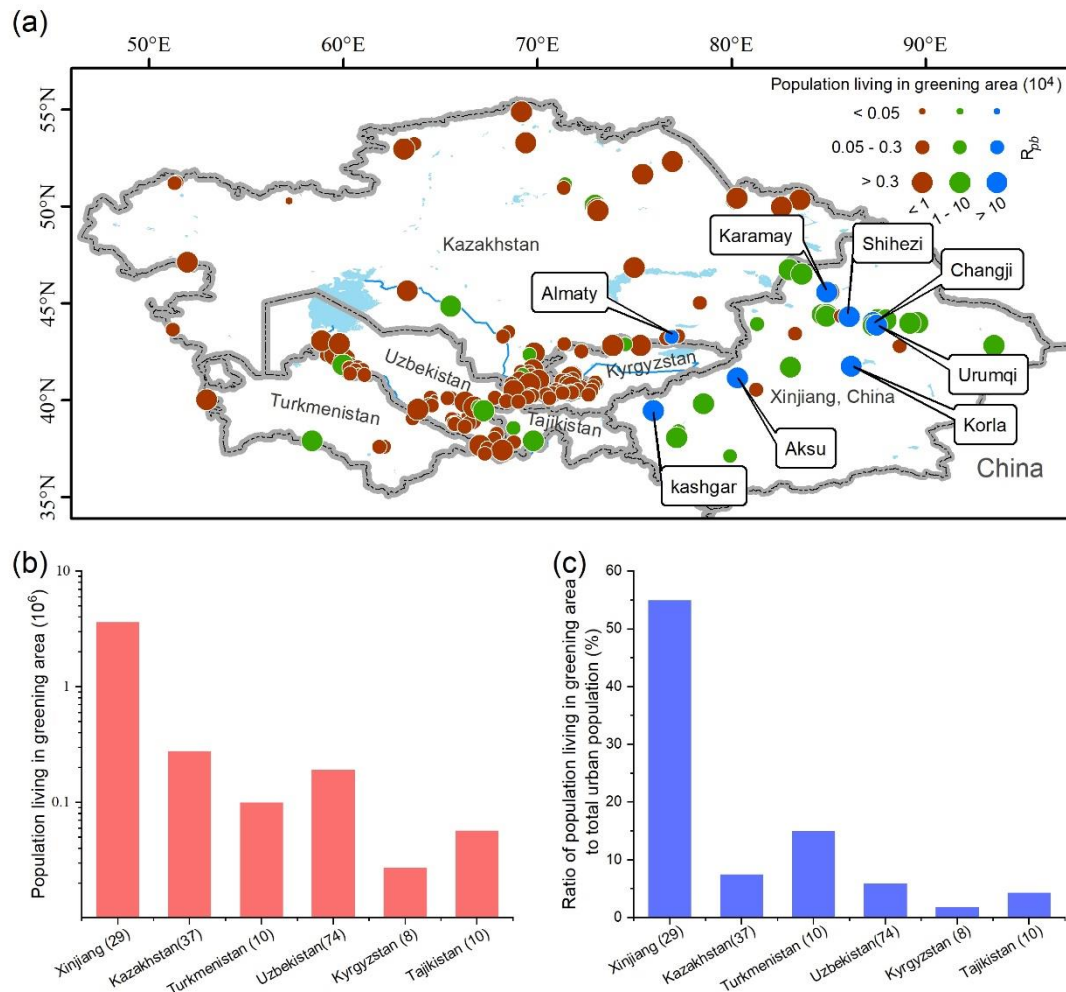


Figure 5. Distribution of the cities with urban populations living in greened built-up areas (BUAs). **(a)** Distribution of the 168 cities with different size populations living in significant greening BUAs (noted as PLGB) from 2000 to 2020. The top 8 cities with PLGBs of greater than 100,000 are labeled with their names (blue points). **(b)** The bars are the total PLGB in Xinjiang, China and other countries. **(c)** The bars are the ratio of the total urban population living in greened BUAs to the total urban population in study area.

There were 3.63 million urban residents living in greening BUAs in Xinjiang, China. In Kazakhstan, Uzbekistan, and Turkmenistan, the populations living in greening BUAs reached 280,000, 190,000, and 100,000, respectively. In Tajikistan and Kyrgyzstan, the populations living in greening BUAs were less than 100,000. The beneficiaries in Xinjiang accounted for 85% of the total beneficiaries in the 168 cities. In particular, more than half (55%) of the urban population in Xinjiang was directly benefited from the significant increase in urban greenness (Figure 5c). In contrast, among the five Central Asian countries, nearly 15% of the urban population in Turkmenistan benefited from the significant increase in urban greenness, while less than 10% of the urban population in the four Central Asian countries benefited from a significant increase in the level of urban greening.

Among the 168 cities, we screened the top 20 cities with the largest beneficiaries (Table 3). It was found that 17 of these cities were in the Xinjiang, China. Urumqi, Korla, Changji, and Aksu were the top four cities with the largest populations living in greening BUAs, accounting for 2.08 million, 204,000, 195,000, and 167,000 people, respectively. In these four cities, the urban greenness of more than 50% of the urban BUAs significantly increased. Compared with Xinjiang, China, the cities in the five Central Asian countries had smaller BUA greening rates (Figure 4b) and lower urban population densities (Figure 2c). According to Table 3, there were only three cities in the Central Asian countries:

Almaty, Ashgabat, and Dushanbe. However, the average 3-year E_{max} in 2018–2020 in Almaty was as high as 0.34, which is the highest among the 20 cities. Similarly, Ashgabat and Dushanbe had average E_{max} of 0.21 and 0.28 in 2018–2020, respectively. In general, the overall ecological environments of these cities were relatively good.

Table 3. Ranking of the 20 large cities by the population living in greened BUAs.

Rank	City	Region	BUA (km ²)	Greened BUA (km ²)	$REVI$ of BUAs in 2020	Urban Population (x10 ⁴)	E_{max} (2018–2020)	population living in greened BUAs (x10 ⁴)	Ratio of PLGB (%)
1	Urumqi	Xinjiang	362	179	0.50	317	0.21	208.5	66
2	Korla	Xinjiang	49	27	0.55	33	0.19	20.4	62
3	Changji	Xinjiang	29	15	0.53	31	0.27	19.5	62
4	Aksu	Xinjiang	41	21	0.51	27	0.21	16.7	61
5	Kashgar	Xinjiang	50	8	0.15	41	0.20	14.4	35
6	Karamay	Xinjiang	49	37	0.75	17	0.24	13.8	83
7	Almaty	Kazakhstan	286	21	0.07	135	0.34	12.6	9
8	Shihezi	Xinjiang	46	14	0.31	31	0.29	11.1	36
9	Hami	Xinjiang	30	10	0.34	19	0.21	9.1	47
10	Ashkhabad	Turkmenistan	69	12	0.17	45	0.21	7.8	17
11	Yining	Xinjiang	76	6	0.07	39	0.25	7.3	18
12	Kuitun	Xinjiang	52	23	0.44	9	0.24	5.6	63
13	Fukang	Xinjiang	14	5	0.37	5	0.28	4.0	76
14	Wusu	Xinjiang	19	10	0.49	6	0.22	3.8	67
15	Dushanbe	Tajikistan	89	4	0.05	98	0.28	3.8	4
16	Dushanzi	Xinjiang	20	14	0.68	5	0.25	3.7	70
17	Hetian	Xinjiang	38	2	0.06	18	0.23	3.6	19
18	Kuche	Xinjiang	23	16	0.71	6	0.19	3.6	64
19	Qitai	Xinjiang	16	5	0.29	5	0.27	2.7	54
20	Bachu	Xinjiang	22	8	0.37	6	0.19	2.3	42

4. Discussion and conclusions

In the BRI, Xinjiang and the five Central Asian countries play an important role as a link and the main carrying areas, respectively. However, the current quantitative research on the urban development in this region is insufficient. It is necessary to quantify and compare the urbanization processes of cities in our study area. In this study, high spatial resolution land cover type data, remote sensing vegetation index data, and grided population data were used to analyze the characteristics of the urban expansion, the population changes, and the urban environmental changes in Xinjiang and the five Central Asian countries from 2000 to 2020. In this section, we discuss the four urbanization features analyzed in this study.

(1) From 2000 to 2020, the urbanization in study area was extremely uneven. Xinjiang and Kazakhstan experienced the most rapid urbanization expansion in the study area, and Urumqi and Nur-Sultan were the two cities with the fastest BUA expansion rates (Figure 1). Although Xinjiang,

China has experienced rapid urban expansion in the past 20 years, its average urban BUA was still at a moderate level in 2020.

(2) From 2000 to 2020, the urban population in Xinjiang increased by 3.4 million (56%), making it the most rapid growing population in the study area. Among the Central Asian countries, Kazakhstan and Uzbekistan had the largest urban population increases, i.e., over 1 million. The average urban population densities of all of the Central Asian countries increased. In addition, in Xinjiang, the urban population density decreased from 2000 to 2020 while maintaining rapid urban population growth (Figure 2). Thus, the pressure of the urban population growth in Xinjiang, China was still smaller than other countries (Figure 3).

(3) In our study area, the proportion of greening BUAs was the highest in Xinjiang, China. On average, the level of greenness significantly increased in more than 35% of the urban BUAs in Xinjiang, which was much higher than in the other countries (Figure 4). Although the urban ecological environment in Xinjiang has improved significantly in the past 20 years, the overall average maximum greenness (E_{max}) of the city is still lower than those of most Central Asian countries (Figure 4c). In Xinjiang, China, 3.63 million (55%) urban residents directly benefited from the increase in the urban greenness, which was much higher than in the other countries (Figure 5).

The above conclusions are based on a quantitative and comparative study of the urbanization in 168 cities in Xinjiang, China and the five Central Asian countries, which were data limited region. Our study may be a comprehensive study of Central Asian cities to date. However, in this study, there were still some limitations regarding the existing data and methods.

First, in this study, the urbanization was quantified using WorldPop 100 m high-resolution population data, which in most cases reflects the true population distribution. However, for some specific cities, the dataset needs to be used with extra caution. For example, in Tashkent, the capital of Uzbekistan, according to the WorldPop data, the urban population was less than 450,000 in 2020. However, many data sources indicate that the urban population of this city exceeded 2 million^[2].

In addition, we used the R_{pb} value, that is, the ratio of the population growth rate to the urban expansion rate, to measure the population pressure during urbanization^[37]. However, there is an obvious disadvantage to using the R_{pb} value. If the urban expansion is small and the urban population increase is not obvious, the R_{pb} value will increase abnormally. In future, a more reasonable index to represent the population pressure during urbanization need to be developed.

The BRI is hailed as the largest infrastructure investment in history, and it will result in huge historical opportunities for economic development and infrastructure construction in Central Asia. Future research should use satellite data to analyze and study the cities in Central Asia and other countries from multiple perspectives, which would provide reliable and scientific data for the BRI, thereby promoting the sustainable development and common prosperity of Central Asian cities.

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