



Article

Urban Green Spaces and Housing Prices: An Alternative Perspective

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Abstract: Urban green spaces (UGS) are essential components of sustainable cities that provide many benefits to urban residents, such as recreation or aesthetics. Urban residents may be willing to pay for some of these ecosystem services. Indeed, studies investigating the formation of housing prices through hedonic pricing analysis have shown that UGS can influence housing prices. Hedonic pricing analysis puts housing units at the center of analysis. In this study, we investigate whether an alternative perspective provides additional insights into the effects of UGS on pricing. The proposed approach puts UGS into the center of analysis by applying an analysis of buffer zones to housing prices, thus, linking approaches from urban economics and landscape ecology. Such an analysis could deepen our understanding of the effects of UGS on housing prices by analyzing price—distance slopes around UGS-thus supporting more informed decisions on efficient UGS management and urban planning. Our results using a case study on Leipzig, Germany, demonstrate that the size of UGS affects price—distance slopes around them. We conclude that further investigations should be fathomed to unleash the potential of applying the analysis of buffer zones around UGS as a method to inform sustainable UGS design in cities.

Keywords: urban green spaces; valuation; buffer zones; hedonic pricing; real estate

1. Introduction

Urbanization is a global trend resulting in the rapidly growing urban population [1]. There is a significant number of factors that determine the values of housing properties in cities such as characteristics of housing and neighborhood (e.g., the presence of child care centers, schools, hospitals, level of crime, etc.), distance to the central business district and places of work. Hectic lifestyles, limited space for rest and mental and physical recovery, turn modern cities into stressful places of living [2]. Urban green spaces (UGS) provide valuable ecosystem services to residents, i.e., deliver "benefits people obtain from ecosystems" [3] (see also the broader perspectives of Daily (1997) or Constanza et al. (1997) [4,5]). The concept of ecosystem services broadly entails the services which are actually perceived, appreciated and valued by humans, such as recreational opportunities, aesthetic enjoyment, water retention and purification, and microclimate regulation (i.e., cooling effect) for cities [1].

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1.1. Hedonic Pricing Analysis of UGS

Hedonic pricing analysis is one method to valuate those ecosystem services that residents perceive and appreciate. Hedonic pricing analysis works under the assumption that urban residents may be willing to pay for some of the ecosystem services provided by UGS since housing prices implicitly reflect the extent to which a variety of amenities, including ecosystem services, are appreciated in certain locations [6]. Thus, many studies have investigated their relationship with real estate prices [4,5,7–21]. Hedonic pricing analysis is the most commonly applied approach for such investigations. Technically, this means explaining the price of a housing unit by considering UGS-related variables, for example, distance to the next UGS or share of UGS in the vicinity together with other relevant variables. Whereas, this approach is straight forward to analyze, it also faces a disadvantage: the spatial distribution of UGS in a city can hamper the analysis of UGS as one of the several elements influencing housing prices [22–24].

For instance, in a city with numerous small and dispersed and a few large UGS, the closest UGS to a given housing unit is more likely to be small. Thus, the effects of the larger UGS may be underestimated in a classical hedonic approach if merely considering the nearest UGS, because price effects will only be quantified for the proximal housing units. This might lead to methodological artefacts and blur the effects of specific characteristics of UGS on prices.

We illustrate this potential effect in a thought experiment (Figure 1) for a specific case: Assuming, we value the impact of UGS on the price of the housing unit located at 9th Avenue and 57th Street in central Manhattan, New York City. When using distance as an indicator of the relevant UGS, this property is linked to Baisley Park—although the dwellers are likely to emphasize that their location is proximal to New York's famous Central Park. Our example indicates the following: The closest UGS is not necessarily the most relevant one for the inhabitants. There are situations in which residents do not prefer the nearest UGS compared with one farther away, for example, due to easier accessibility, size or other features.

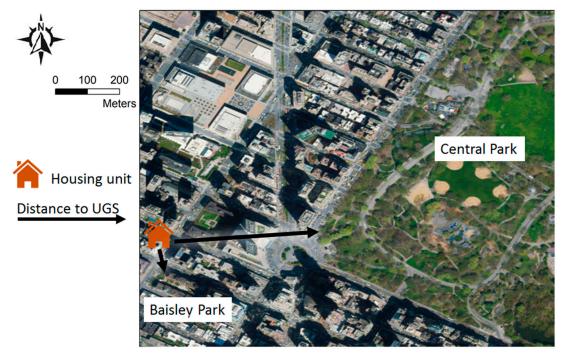


Figure 1. A thought experiment: the closest UGS may not be the most relevant. (ESRI aerial imagery basemap, 2019).

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1.2. Gradient Analysis in Landscape Ecology

Putting the UGS into the center of analysis could help to disentangle the effects of different UGS. One approach to do so is the analysis of buffer zones around UGS as commonly performed in landscape ecology. Urban-to-rural gradients can be analyzed for the assessment of ecosystem service provision [1], changes of landscape patterns [25], urban expansion [26–28], or changes in UGS structure [29,30]. In contrast to gradients from the central business district (CBD), gradients from UGS have been employed, for instance, to investigate their cooling effect [31–33]. Thus, the analysis of buffer zones around UGS tests whether gradients going away from this UGS exist with respect to a certain variable. In doing so, the impact of this UGS onto the surroundings is investigated.

1.3. An Alternative Perspective of Analysing UGS Effects on Prices

We propose to investigate if both methodologies can be linked in order to exploit their advantages: hedonic pricing offering a hands-on, easy to communicate monetary assessment of the value contributed by various characteristics (e.g., features of UGS), and analysis of buffer zones offering spatial insight into the impact analysis centered around the UGS. We suggest a combination of both in a way that housing prices are analyzed while putting UGS at the center of this investigation as it is demonstrated in Figure 2. This figure visualizes how a hedonic pricing approach (a) differs from our proposed approach (b).

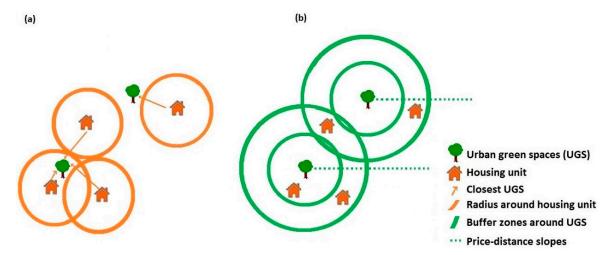


Figure 2. Hedonic pricing analysis (a) and analysis of buffer zones (b).

The alternative approach places UGS in the middle of the analysis. Here, each of the UGS is considered individually; a single housing unit can belong to different buffers around different UGS. Housing units are used to compute mean housing prices at different distances to the UGS. In contrast, housing units are the starting point for hedonic studies (Figure 2a), and the same UGS can be associated with several housing units at once.

To summarize, the proposed approach shifts the starting point from the social–economic or property market sphere of the hedonic pricing analysis into the ecological sphere. We believe that this approach could be an inspiring next step to stimulate the dialogue between scholars from different disciplines about the impacts of UGS onto housing prices.

1.4. Aims and Organization of the Study

To test the usefulness of such an analysis of buffer zones around UGS, we take a first step in applying this method to the city of Leipzig, Germany. The objective is to start a scientific discussion about the proposed approach that, in turn, will lead to its further development. To do so, we search for empirical evidence of price—distance slopes around UGS and test whether size as an exemplary UGS characteristic, as well as neighboring UGS, influence these price—distance slopes.

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Therefore, we investigate the following hypotheses:

- H1: there are price-distance slopes around UGS.
 - O We expect to find price-distance slopes around UGS.
- H2: differences in price-distance slopes are related to differences in UGS size.
 - We expect that price-distance slopes around large UGS are more pronounced since prices around small UGS are potentially influenced also by larger UGS close by (Figure 1).
- H3: neighboring UGS influence price-distance slopes.
 - We test if the effects of UGS on prices superimpose each other, implying that other UGS in the vicinity influence their effect on prices in their own buffers.

This study builds on the lines of Liebelt et al. [11,23], which analyzed the influence of UGS on housing prices in Leipzig on the city and district scales by using a classical hedonic pricing setting. These studies have found that UGS characteristics (e.g., UGS size) affect housing prices. Notably, these effects differed when analyzed on a city or district level, indicating that spatial heterogeneity requires additional sensible investigation. The partly counter-intuitive results of the previous analysis [15] motivated us to explore the analysis of buffer zones around UGS as a tool to deepen our understanding of the impact of UGS characteristics on housing prices.

In Section 2, we provide information on data and method related to our case study. Section 3 presents the results. Section 4 discusses the hypotheses and the general concept, study limitations, and suggestions for further research. Section 5 concludes.

2. Materials and Methods

2.1. Case Study

The study area is the city of Leipzig, Germany, with 29,760 ha and approx. 575,000 inhabitants [34]. Leipzig has many UGS of approximately 4900 ha, compared with the 6300 ha of residential areas (Figures 3 and 4a, Section 2.2 for more details). The city is notable for its UGS that are easily accessible and vary in size, with some large parks located not far from the city center and forests located within the city boundary.

Leipzig is a city where renting is more common than purchasing housing units due to several reasons. First, social housing introduced after World War II which made renting of housing units less expensive than purchasing [35]. Second, homeowners in Germany, in contrast to other European countries like the Netherlands or Spain, benefit from subsidies to a lesser extent [35]. Finally, rental prices in Germany have remained relatively stable over a long period of time [35]. During the last few years, however, rental and sales price started to increase, reflecting urbanization and segregation in Germany, in general, and in Leipzig, in particular [36].

In total, residential housing consists of multi-family houses (88%) and detached and semi-detached houses (12%). One third of the housing stock was constructed during the Wilhelminian period (i.e., Gründerzeit until 1918), around 19% of housing units were built between 1919 and 1948, 29% of houses were built in GDR period between 1960 and 1990 (these are mostly high-rise buildings with 4–12 stories [37]); and 19% of modern housing units built in post GDR period (i.e., after 1990) [38].

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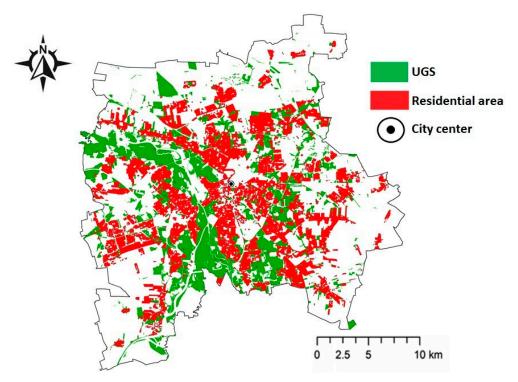


Figure 3. City of Leipzig—composition of urban green spaces (UGS) vs. residential area.

2.2. Real Estate and Spatial Data

This study focusses on flats available to rent since, first of all, renting is the most common type of housing; and, second, it was the largest group of housing for which data was available. The real estate data comprises 261,827 flats available for rent in Leipzig for the period of 2007–2013, obtained from the German real estate web portal Immobilienscout24. We carefully examined the data to avoid inconsistencies and exclude unrealistic outliers by applying cut-off criteria: the minimum size for flats was set to 15 m² and the maximum to 300 m². Additional corrections related to handling of missing data include logical recoding and imputation (described in detail in Liebelt et al. [11,23]). We obtained spatial data on UGS from Germany's nationally standardized official topographic-cartographic information system for the year 2010 with a spatial resolution of 1:25,000. We combined various UGS types, namely, allotments, forests, cemeteries, and parks, into a single category. We assume that these provide ecosystem services to local residents in a similar extent and are the most suitable to test our hypotheses. However, in order to prove that our previous statement is meaningful, we also analyzed UGS types separately. In total, 744 UGS were included in the analysis, with an average size of 6.9 ha (ranging from 4 m² to 495.5 ha). We did not restrict the size of UGS to maintain all raw data and explore the influence of all sizes of UGS.

2.3. Analysis of Buffer Zones around UGS

2.3.1. Data Preparation

In total, 744 UGS were characterized regarding their *distance to the city center (CBD)*, *UGS size* (in ha), and *UGS share* (%) of other UGS around each UGS in a 2000 m radius. Spatial data analysis and preparation of the study variables were performed in ArcGIS v.10.5. All statistics were calculated using R v. 3.4.0 software [39].

For the price–distance slopes (Section 2.3.2), we calculated the mean rental price per m² in 20 buffers around each UGS. We created buffers of 100 m each, because this buffer size has been commonly used in landscape ecology [40–42]; additionally, we found a sufficient number of flats per buffer with this buffer size. Housing units within a specific buffer were used to calculate average rental

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prices per m². Thus, for each of the UGS, all flats within 20 buffers, that is, up to a distance of 2000 m, were considered (Figure 4). This distance was chosen to capture the price effects of large UGS also. UGS without a large-scale effect on prices result in nonsignificant or nonlinear price—distance slopes (Section 3.1). In the analysis of buffer zones, flats can belong to several buffers of various UGS simultaneously, whereas no double-counting of UGS occurs.

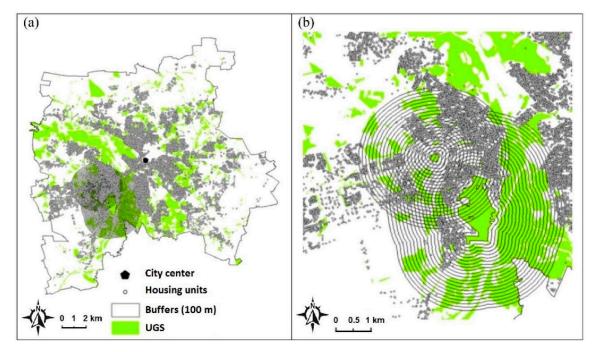


Figure 4. Analysis of buffer zones around UGS with two exemplary UGS *. * same method performed for all 744 UGS; (a) UGS and housing units of Leipzig, and 20 buffers around two randomly selected UGS; (b) a cropped version of (a), presenting buffers around the two UGS in greater detail.

Table 1 lists all the variables considered in the analysis. Mean price is a dependent variable representing the mean rental price for flats located in a particular buffer (from 1 to 20). As a basis for calculation, we used monthly listing flat prices (i.e., requested renting prices stated on the web by the offering party). *UGS size* was calculated for each of the 744 UGS in hectares. CBD distance was calculated in meters for each of the UGS, with the central train station used as the city center. Additional details on *UGS types* are presented in Section 2.2.

Table 1. Variables used in the analysis of buffer zones for 744 UGS.

Variable and Units	Description	
Mean price (Euro/m ²) in buffers from 100 m to 2000 m	Mean rental price for all flats located in a buffer, calculated for 20 buffers of 100 m radius each.	
UGS Size (ha)	Size of UGS (ha).	
CBD Distance (m)	Distance from central business district (CBD) to UGS.	
UGS Share (%)	Share (in %) of UGS in the buffer area of 2000 m around the UGS.	
UGS Types	Types of UGS: allotment, forest, cemetery or park.	

2.3.2. Determining Price-Distance Slopes

Price—distance slopes indicate whether an increase or decrease in rental prices occur with increasing distance to UGS (i.e., while 'moving' from buffer 1 to 20). The visual inspection of changes in price over distance to UGS also indicated nonlinear relationships. Therefore, we applied a Box–Cox

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transformation [43] to identify other appropriate functional forms [44]. The Box–Cox transformation procedures [45] for all UGS indicated that the following values of the coefficient λ occur most often: $\lambda = 0$, $\lambda = 1$, $\lambda = -1$, $\lambda = -2$, $\lambda = -3$, leading to the transformations presented in Table 2 [46]. Thus, for each UGS, five linear regressions were fitted with transformed mean flat price per m² per buffer as a dependent variable and buffers (buffer 1 to buffer 20) as independent variable. As the main focus of this study is the change in prices rather than absolute values, we focused on the resulting slopes only and considered slopes significant at the 95% level.

λ	Required Transformation	Explanation	
1	None	Price	
0	Natural log	log(price)	
-1	Reciprocal, i.e., power of −1	1/price	
-2	Power of −2	1/(price^2)	
-3	Power of -3	1/(price^3)	

Table 2. Transformations of mean price per buffer before quantifying price-distance slopes.

2.3.3. Hypotheses Testing

Appendix A summarizes our analysis from data preparation to hypothesis testing. First, we test whether there are price—distance slopes around UGS (H1). Second, we ran linear regressions to analyze the relationships of *UGS size* and *UGS share* with price—distance slopes. To account for data skewness (e.g., Appendix E, Appendix F, Appendix G) and outliers [47], we also applied log and square root transformations of *UGS size* and *UGS share* [48].

We expect to observe relatively low values of adjusted R² as an outcome of the analysis. Our reasoning was that we explained mean rental prices at different distances to UGS only with *UGS size* and *UGS share*. Housing variables (e.g., size and presence of a balcony), which play an important role in explaining prices [7,15,49,50] were not considered because we focused explicitly on UGS in order to test whether we find any empirical evidence for price–distance slopes around UGS.

3. Results

This section reports the results for the empirical case study, first explaining the general results on common patterns for UGS that were included and not included in the analysis (Hypothesis 1). Subsequently, the results are provided for analysis of UGS included in the models to test Hypotheses 2 and 3.

3.1. Hypothesis 1

Figure 5; Figure 6 present the amalgam of price—distance slopes for UGS for the city of Leipzig based on average mean price—distance slopes across all UGS. The average price tends to decrease as the distance from the UGS increases. All three potential patterns for the gradients of mean prices (Euro/m²) along the buffers of UGS were present. We observed increasing (Figure 5a) and decreasing prices (Figure 5b) as well as no clear pattern (Figure 5c) when moving from one of the UGS' nearest buffers to those farther away.

The data also indicated that for some UGS, flats are present either in very few buffers or not at all. Hence, we excluded 37 UGS from further analysis (i.e., cases when flats were present in less than four buffers) as the analysis of buffer zones cannot be meaningful in such cases.

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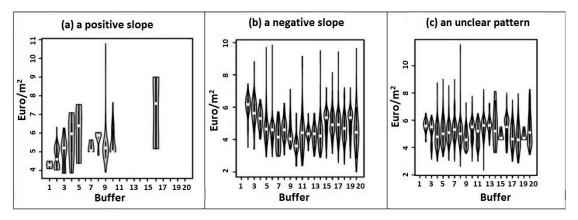


Figure 5. Price-distance slopes examples for 20 buffers around UGS with positive slope (**a**), negative slope (**b**), and unclear pattern (**c**).

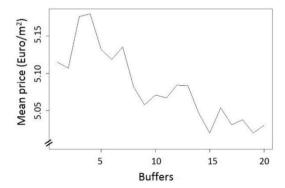


Figure 6. Price gradients across all UGS in Leipzig*. * Distribution of mean (untransformed) prices per buffer.

When fitting the price—distance slopes by applying the various price transformations (as indicated in Table 2), 389 UGS were not included in any of the five regression models that relate distance to UGS with a price (Section 2.3.2) at the 95% significance level. As a result, 355 UGS were included in one or more regression models. Appendix B provides details on how many price—distance slopes were statistically significant for each of the price transformations. Overall, we found some empirical evidence on Hypothesis 1 that price—distance slopes around UGS exist; and, therefore, it was accepted.

We also applied a t-test to determine which variables might be related to the fact whether price—distance slopes are significant or not. Figure 7a,b illustrates that CBD distance (p < 0.001) and UGS share in 2000 m buffer (p < 0.0001) are clear indicators, whereas UGS size (Figure 7c) is not. In other words, we observe a greater number of significant price—distance slopes for UGS located closer to the CBD and in areas with a higher UGS density in the 2000 m buffer.

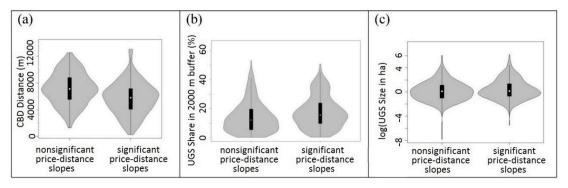


Figure 7. Comparison of significant vs. nonsignificant price—distance slopes for CBD distance (**a**), *UGS* share in 2000 m buffer (**b**), and *UGS* size (**c**).

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We also tested whether price—distance slopes are different for different *UGS types*. Appendix C summarizes the results of the linear regressions performed to test the relationship between UGS type and the price—distance slopes. None were statistically significant. Additionally, graphical inspection does not hint at any relationship (Appendix D).

3.2. Hypothesis 2

To test Hypothesis 2, we ran a *t*-test for positive and negative price–distance slopes and UGS size. The *t*-tests were significant (95% confidence interval) for logarithmically transformed *UGS sizes* and logarithmically transformed price–distance slopes (Figure 8, examples of nonsignificant relationships are in Appendix E, scatterplots for *UGS size* and price–distance slopes are in Appendix F). Results indicated that negative price–distance slopes are more likely to be observed for UGS of smaller size than positive price–distance slopes.

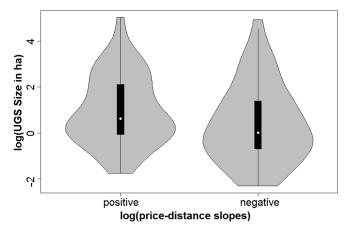


Figure 8. Comparison of positive vs. negative price–distance slopes for different *UGS size*. Variables underwent logarithmic transformations to account for data skewness and outliers.

The statistical tests described above demonstrate some empirical evidence for the acceptance of H2: differences in price–distance slopes are related to differences in *UGS size*.

3.3. Hypothesis 3

To test Hypothesis 3, we plotted *UGS share* in a buffer of 2000 m around each UGS against price–distance slopes (Appendix G) and performed statistical tests (Table 3).

Transformation of	Transformation of		Estimate for		
Price-Distance Slope ^{1,2}	UGS Share in 2000 m Buffer ³	Intercept	UGS Share in 2000 m Buffer	Adj. R ²	<i>p</i> -Value
	none	-0.0006 *	0.0002	0.001	0.270
None	log(UGS share)	-0.001 *	0.0002	0.006	0.113
	sqrt(UGS Share)	-0.0008 *	0.0001	0.003	0.180
Natural log	None	-0.0003	0.00001	0.004	0.187
	log(UGS share)	-0.0005	0.0002	0.008	0.121
	sqrt(UGS share)	-0.0005	0.0001	0.008	0.112
Reciprocal	None	1.9×10^{-6}	-6.5×10^{-8}	0.001	0.274
	log(UGS share)	1.7×10^{-6}	-4.0×10^{-7}	-0.004	0.638
	sqrt(UGS share)	3.0×10^{-6}	-5.6×10^{-7}	0.002	0.254
Power of −2	None	3.8×10^{-9}	-2.0×10^{-10}	-0.005	0.888
	log(UGS share)	-8.4×10^{-8}	5.9×10^{-8}	0.0066	0.140
	sgrt(UGS share)	1.9×10^{-9}	1.2×10^{-8}	-0.005	0.872

Table 3. Regression results to test if *UGS share* influences price–distance slopes.

¹ See Table 2. ² price–distance slopes that sometimes underwent various transformations (see Section 2.3.2)-dependent variable; ³ *UGS share* in 2000 buffer that sometimes underwent various transformations to account for data skewness and outliers (see Section 2.3.3)—as independent variable.

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We observed no statistically significant relationship (p > 0.001) between neighboring UGS and the price–distance slopes (Table 3). Thus, H3: neighboring UGS influence price–distance slopes, was rejected.

4. Discussion

4.1. Hypothesis 1

The analysis of buffer zones around UGS applied in this study provides additional details on the price -distance slopes with regard to the variable *UGS distance* that has been very often analyzed in hedonic studies [7,13,49–52]. In the majority of hedonic studies, *UGS distance* demonstrated negative effects on real estate prices [49,52,53], that is, the shorter the distance to UGS, the higher the housing price. Notably, an opposite effect of *UGS distance* on prices was found for the Leipzig case study [15].

The analysis undertaken in Section 3 leads to the acceptance of Hypothesis 1. There are two meaningful outcomes in respect to Hypothesis 1 that are worth mentioning. Firstly, about half of the UGS have significant price—distance slopes, regardless of the housing characteristics influencing prices. Secondly, there is a huge diversity (positive and negative slopes, different transformations possible) that needs more work in the future.

Moreover, our analysis did not find different price—distance slopes for different *UGS types* such as allotments, forests, cemeteries or parks. For hedonic studies, some authors have argued that analyzing various types of UGS separately, instead of a single aggregated UGS variable (that combines various *UGS types*), might provide more detailed and precise results [54]. Other authors have argued it is not always feasible due to a large number of variables [15]. The fact that we did not find any evidence of different price—distance slopes depending on *UGS types* indicates that our decision to aggregate the types of UGS did not lead to a substantial loss of information in this case study.

4.2. Hypothesis 2

Before discussing the results for Hypothesis 2, assessing the magnitude of the effects is worthwhile. The analysis demonstrated very low values of explanation of variable variances (adjusted R²). This result should not lead to the misjudgment that these outcomes are not worthy of further discussion. Moreover, our intention (Section 2.3.3) was not to explain the variance of prices as best as possible but to provide additional details on UGS and test whether we would observe significant linear models and the direction of effects. Overall, the results support Hypothesis 2: differences in price—distance slopes are related to differences in UGS. This hypothesis covers the interaction of two attributes commonly analyzed in hedonic studies: *UGS size* and *distance to UGS*.

For *UGS size*, hedonic studies have found a positive effect on housing prices [15,55], arguing that larger UGS offer additional recreational opportunities and greater access to flora and fauna in contrast to smaller UGS [56]. A negative impact of *UGS distance* on prices is commonly assumed and has been found in hedonic studies [34,37], implying that prices decrease with increasing distance to UGS. Notably, for the case study Leipzig, we have found that *UGS distance* had a positive influence on the flats' sale prices (i.e., that housing prices increase with increasing distance to UGS) [15].

The results of these buffer zones focused study allowed investigating the interaction between the two variables-*UGS* size and distance to *UGS*. Environmental interest groups and urban planners can use the results of such regression analysis to show that *UGS* size has a positive impact (i.e., positive regression coefficient) on price–distance slopes, that is, rental price increases with increasing distance to *UGS* (i.e., positive price–distance slopes) for *UGS* of bigger size. The opposite phenomenon applies for smaller *UGS*: mean prices decrease with increasing distance to *UGS*.

The results should be interpreted with caution: The common assumption—that decreasing distance to UGS causes an increase in real estate prices—holds for UGS of smaller sizes. Thus, we are tempted to assert: If UGS are within a short walk, they are likely to cause an increase of real estate values—for all the often-stated reasons (e.g., scenic view, microclimate effects and areas for

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recreation). For larger UGS, the situation is different and considerably more difficult to interpret. Assuming our result is not unique for the specific case study Leipzig, we conclude that large UGS (greater than or equal to 100 ha) might be regarded by inhabitants as a characteristic element of their neighborhood and a unit with specific land uses and attributes comparable to a sports stadium or amusement park. These facilities are connoted with offering services but can also carry disservices to nearby residential inhabitants, for example, due to incoming and outgoing traffic from the respective locations. Additionally, we found out that a number of other factors have an impact on housing prices in relation with *UGS size* at the district level. For instance, an increase in population density causes an increase of the *UGS size* coefficient [57]. In other words, urban residents are willing to pay higher prices for housing units located next to UGS of small sizes in a case when population density is low and vice versa. One of the potential explanations here is that the competition for a public good is rather low (i.e., small UGS) when there is a low population density. What is more, Leipzig has an outstanding abundance of UGS and it is well known that it is a scarcity that makes things more valuable. These aspects might partly explain the negative influence of decreasing distance on prices for large UGS.

The findings presented in Section 3.2 indicate that the relationship between the UGS-related variables and prices is complex. This result is in line with the literature that indicated that the impact of UGS-related variables in general, as well as *UGS size* and *UGS distance* in particular, are scale-dependent and have different directions regarding the effects on the district level, in contrast to the city level [57].

To summarize, we found that with increasing *UGS size*, the impact of *UGS distance* on rental prices changes from negative to positive. Therefore, we recommend considering the interaction of *UGS distance* and *UGS size* within the hedonic pricing models, for example, by creating a variable ln(size/distance) as suggested by Kong et.al [7]. Additionally, this type of analysis improves the understanding of the roles of large and small UGS.

4.3. Hypothesis 3

Our analysis calls for the rejection of Hypothesis 3: neighboring UGS influence price–distance slopes.

It is critical to not confuse the variable mentioned in this hypothesis (i.e., neighboring UGS) with the variable used in hedonic analysis (i.e., share of UGS in a particular radius around the housing unit) as *UGS share* represents the 'greenness' of the area and can demonstrate a statistically significant impact on prices [10,15]. In case of analysis of buffer zones, with UGS being in the center of the analysis, the presence of other UGS in the surroundings of the analyzed UGS could have blurred the effect that the single UGS has on price–distance slopes.

4.4. Limitations of the Study

This study has several limitations concerning data availability. First, due to the lack of data on private gardens, we could not assess their influence on the price–distance slopes.

Second, as the real estate data in this study depended exclusively on user entries on the real estate website, it did not have high overall reliability, for example, due to missing values. Nevertheless, there is no reason to assume that any systematic error occurred and biased the results. For a discussion on the specific limitation regarding the real estate data, please refer to Liebelt et al. [11].

Third, this study used flats available for rent to apply the analysis of buffer zones around UGS. This factor limited our ability to relate our outcomes to hedonic studies in the literature, in which the majority have analyzed housing units for sale. However, the purchasing of a housing unit is a form of real estate transaction that is far less common in Germany; thus, the insufficient data would have prevented a proper analysis.

4.5. Suggestions for Future Research

With our study, we took the first step towards an analysis of buffer zones around UGS in respect with housing prices. We explored available data for one empirical case study and found that—not

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accounting for housing characteristics—we were able to identify price—distance slopes around UGS which are dependent on *UGS size*. We suggest two ways for the potential further development of the proposed approach. Both would account for the variety of factors that are usually considered in hedonic pricing studies (e.g., proximity to bus stops, schools, city center) and omitted our data exploration. The first option would be to run a classical hedonic analysis with housing units at the center of analysis and afterwards analyze the residuals of the regression function in the analysis of buffer zones. The second (alternative way) would be to include a similar set of independent variables that are commonly used in hedonic pricing analysis into the analysis of buffer zones (i.e., as regression function) by computing, for instance, mean flat size for all buffers. In addition, for obtaining further details or more clarity on the impact of the *UGS size* on the housing prices we suggest performing an analysis with UGS that were grouped based on the size. Moreover, acknowledging potential spatial autocorrelation, it might be meaningful to apply appropriate regression models e.g., [13,42] and to address endogeneity and discuss an identification strategy.

In our study, we considered distance isotopically, that is, without considering the direction (e.g., south, north, east, or west) when determining the price—distance slopes and calculated the other variables. In this respect, we argue that by considering the mean prices within the buffers, we avoided a potential bias of the results because all potential influences of directions (if any) were levelled out. Nevertheless, investigating potential influences that occur in the different directions of the gradients might provide notable outcomes in similar studies.

Our study made use of the hedonic pricing approach, which is only valid for those ecosystem services immediately perceived and appreciated by residents. A very different perspective from this is the common good perspective asking for the total economic value of the UGS in a city (including future generations). This approach could also be tested and assessed by appropriate economic monetization methods e.g., [20,21].

5. Conclusions

This study attempted to analyze the effect of UGS characteristics on housing prices from a different perspective than typically adopted in the hedonic pricing literature. In our explorative analysis of buffer zones, we placed UGS in the center of analysis, in contrast with the classical concept of hedonic studies where housing units are the starting points of all calculations. The objective of the study was to initiate a scientific discussion about the advisability of changing the perspective in the UGS valuation by linking analysis of buffer zones to the hedonic studies. Our study is only the explorative first step and, therefore, further steps should be undertaken in order to enable its efficient and meaningful application.

Our data exploration was guided by three hypotheses on whether price—distance slopes can be found (Hypothesis 1) and whether *UGS size* (Hypothesis 2) and presence of other UGS (Hypothesis 3) influence these slopes.

In respect with Hypothesis 1 we found empirical evidence on the presence of price–distance slopes around UGS.

Regarding Hypothesis 2, we concluded that *UGS size* influences price—distance slopes. Thus, this hypothesis was accepted. We found that with increasing *UGS size*, the impact of *UGS distance* on rental prices changes from negative to positive, that is, rental prices increase with increasing distance to UGS for the larger UGS and vice versa for the smaller UGS.

For Hypothesis 3, we found no significant influence of neighboring UGS in the radius of 2000 m on the price–distance slopes.

By flipping the perspective from housing units as the starting point (hedonic pricing analysis) to UGS (analysis of buffer zones), we hope to open the floor for further investigations of the effects of UGS on housing prices with UGS being the center of analysis. Thus, we conclude that analysis of buffer zones can be a meaningful approach for UGS valuation once linked to hedonic pricing. There are several potential advantages associated with application of the proposed approach. Firstly, additional detail about the distance to UGS that cannot be captured with a hedonic pricing analysis can

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be obtained. For instance, in this case study, an increase and decrease of rental prices with increasing distance to UGS are observed. As for the case of Leipzig, we argue that analysis around buffer zones can provide additional information regarding the effect of UGS on prices, in contrast to hedonic studies, where univocal results can be caused by an averaging effect [57]. Secondly, investigating such relations with a clear focus on UGS is possible. In this study, we relate the prices-distance slopes to *UGS size* and *share of UGS* in a 2000 m buffer. Changing this perspective means that many more UGS characteristics could be easily investigated, for instance, the shape of UGS, the specific type of UGS such as forest versus park, or the availability of UGS-related infrastructures such as benches, lights or bins. In the long run, such investigations could support more informed decisions on efficient urban UGS management.

On the whole, with this contribution, we encouraged to change perspective from housing units to urban green spaces becoming the center of analysis. At the moment analysis of buffer zones around UGS, as presented in this study, is only a starting point that obviously requires further investigation and development. Further discussion of the methodological approach will show if and which additional insights it can yield.

Author Contributions: Conceptualization V.L., N.S. and S.B.; methodology, V.L., N.S. and S.B.; investigation, V.L.; data curation V.L. and N.S.; formal analysis V.L.; visualization V.L.; writing—original draft, V.L.; writing—review and editing, V.L, N.S. and S.B.

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Conflicts of Interest: The authors declare no conflict of interest.

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Appendix A

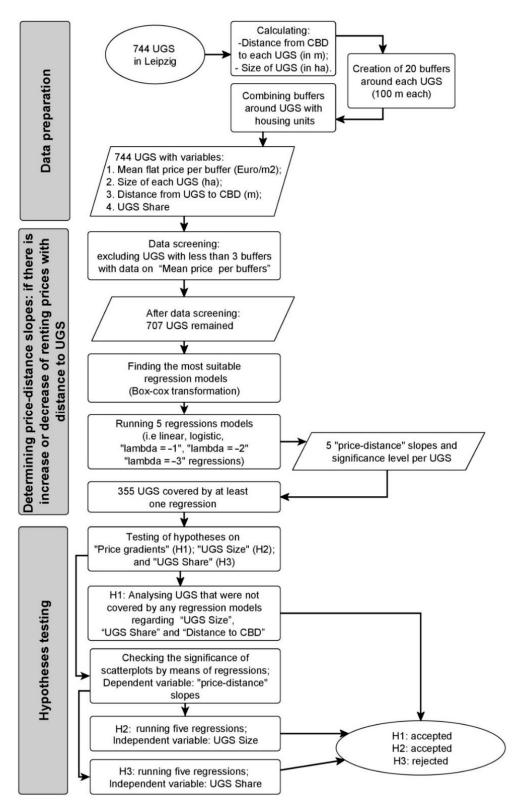


Figure A1. Analysis overview.

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Appendix B

Table A1. Number of cases for different types of regression.* Cut-off means that those UGS containing flats only in the buffers 0–3 were excluded (see Section 2.3.2 for more details).

	Total Number of UGS		
Total for 20 Buffers	744		
I. Regression with untransformed prices (lambda = 1)			
Significance level of 95%, all 20 buffers	245 (32% of total)		
Significance level of 95%, cut-off * applied	233 (31% of total)		
II. Regression with log-transforme	d prices (lambda = 0)		
Significance level of 95%, all 20 buffers	192 (26% of total)		
Significance level of 95%, cut-off applied	180 (24% of total)		
III. Regression with reciprocal transforma	ation of prices (lambda = -1)		
Significance level of 95%, all 20 buffers	188 (25% of total)		
Significance level of 95%, cut-off applied	176 (24% of total)		
IV. Regression with 'power of −2 transform	nation' of prices (lambda= -2)		
Significance level of 95%, all 20 buffers	184 (25% of total)		
Significance level of 95%, cut-off applied	172 (23% of total)		
V. Regression with 'power of −3 transform	ation' of prices (lambda = -3)		
Significance level of 95%, all 20 buffers	190 (26% of total)		
Significance level of 95%, cut-off applied	178 (24% of total)		

Appendix C

Table A2. Regression results to test Hypothesis 1.° See Table 2.

Transformation of Price-Distance Slope °	UGS Type	Intercept	Estimate for Selected UGS Type (Binary)	Adj. R2	p Value
None	Allotment	-0.0005	0.0004	0.004	0.15
	Forest	-0.0002	-0.0001	-0.0003	0.59
	Cemetery	-3.5×10^{-4}	-2.2×10^{-6}	-0.004	0.996
	Park	-0.0004	0.0003	-0.002	0.48
Natural log	Allotment	-0.0002	0.0002	-0.001	0.34
	Forest	1.7×10^{-5}	-1.7×10^{-4}	-0.001	0.37
	Cemetery	-1.1×10^{-4}	1.6×10^{-4}	-0.004	0.59
	Park	-1.1×10^{-4}	1.2×10^{-4}	-0.004	0.65
Reciprocal	Allotment	1.2×10^{-6}	-1.5×10^{-6}	0.001	0.26
	Forest	-1.3×10^{-6}	2.7×10^{-6}	0.014	0.06
	Cemetery	9.8×10^{-7}	-4.3×10^{-6}	0.013	0.07
	Park	6.9×10^{-7}	-8.2×10^{-7}	-0.005	0.69

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Appendix D

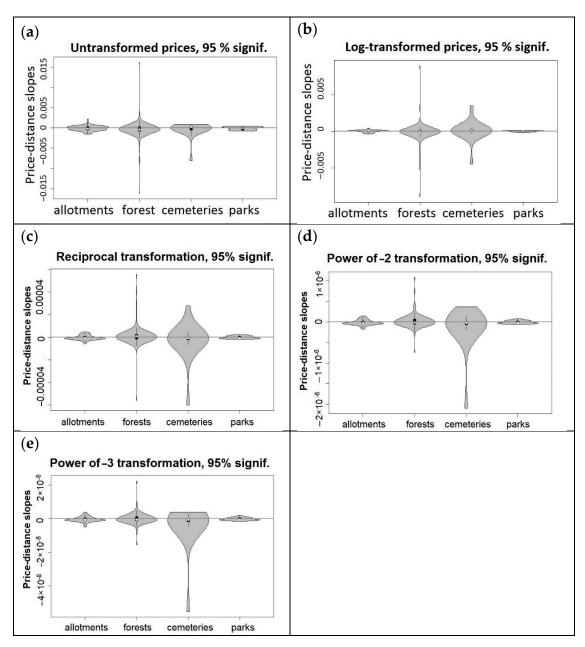


Figure A2. Variation of price–distance slopes for untransformed (a) prices for *UGS types* and prices with logarithmic transformation (b) and transformations of lambda = -1 (c), lambda = -2 (d), and lambda = -3 (e).

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Appendix E

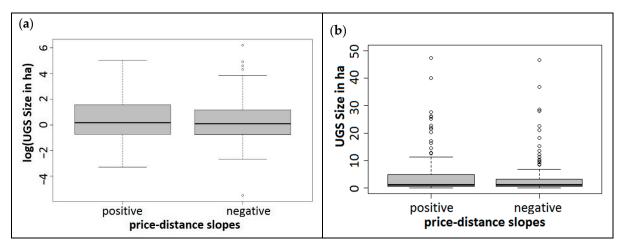


Figure A3. Comparison of positive vs. negative price–distance slopes for different UGS. Some variables underwent logarithmic transformations to account for data skewness and outliers.

Appendix F

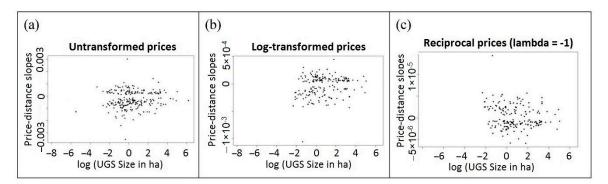


Figure A4. Price—distance slopes in relation to *UGS size* (ha) presenting statistically significant linear relationships for untransformed prices (a), log-transformation (b), and reciprocal transformation (c) of prices with log (UGS size), 95% significance level.

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Appendix G

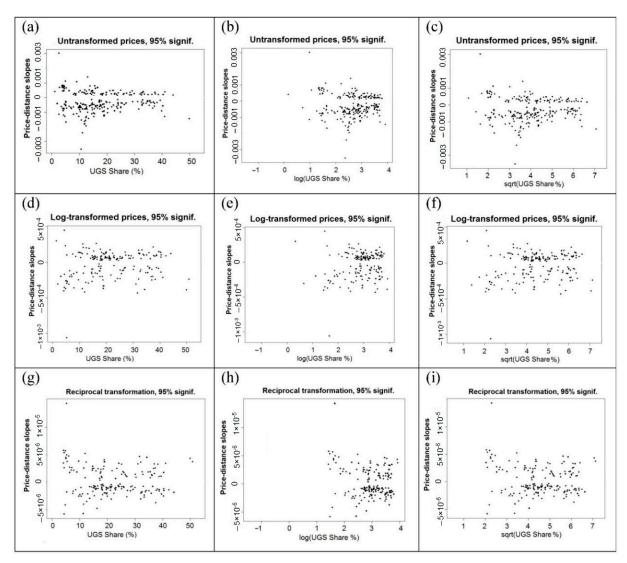


Figure A5. Price-distance slopes in relation to UGS share in 2000 m buffer (%).

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