

# Spatial sorting, attitudes and the use of green space in Brussels

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## Abstract

Extensive evidence exists on the benefits provided by urban green space (UGS) but evidence is lacking about whether and how socio-economic benefits accrue to all residents or disproportionately depending on their socio-economic status or residential location. We model joint effects of socio-economic and locational attributes on attitudes and use of UGS in Brussels (BE). The analysis is based on a survey conducted along an urban-suburban continuum with respondents sampled across non-park public space. Patterns of use are depicted by the frequency and the distance travelled to the most used UGS. Attitudes are analysed along three dimensions: willingness to (i) pay for UGS, (ii) trade off housing for green space and (iii) substitute private for public green. Our results stress the importance of separating effects of attitudes from socio-economic and locational effects to quantify UGS use, and suggest endogenous effects of green space with residential sorting.

*Keywords:* urban green space, residential sorting, green space amenities, stated attitudes, Brussels

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## **Abstract**

Extensive evidence exists on the benefits provided by urban green space (UGS) but evidence is lacking about whether and how socio-economic benefits accrue to all residents or disproportionately depending on their socio-economic status or residential location. We model joint effects of socio-economic and locational attributes on attitudes and use of UGS in Brussels (BE). The analysis is based on a survey conducted along an urban-suburban continuum with respondents sampled across non-park public space. Patterns of use are depicted by the frequency and the distance travelled to the most used UGS. Attitudes are analysed along three dimensions: willingness to (i) pay for UGS, (ii) trade off housing for green space and (iii) substitute private for public green. Our results stress the importance of separating effects of attitudes from socio-economic and locational effects to quantify UGS use, and suggest endogenous effects of green space with residential sorting.

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## 1. Introduction

1 Urban green space (UGS) is important for making cities more sustainable.  
2 It provides ecosystem services (De Groot et al., 2002) such as direct uptake  
3 and deposition of pollutants (Jayasooriya et al., 2017) or the reduction of  
4 heat islands (Bowler et al., 2010; Kabisch & Haase, 2014), and makes cities  
5 more attractive to people by increasing quality of life by improving physical,  
6 social and psychological wellbeing (Chiesura, 2004; Fuller & Gaston, 2009;  
7 Keniger et al., 2013) or supporting social interaction (Martin et al., 2004).

8 The actual benefits people derive from UGS vary according to availability  
9 and distance from residential locations (Box & Harrison, 1993; Harrison et al.,  
10 1995; Giles-Corti et al., 2005; Chiesura, 2004; Neuvonen et al., 2007; Kabisch  
11 et al., 2015). Local, national and even supranational regulatory bodies have  
12 therefore set up proximity criteria to improve the general provision of UGS  
13 in cities and ensure equal access by all residents (CABE, 2010). However,  
14 even though the provision of public UGS meets the per capita standard  
15 in many cities, their spatial distribution can still facilitate UGS benefits  
16 disproportionately among the population depending on their socio-economic  
17 status and residential location (Barbosa et al., 2007; Comber et al., 2008).

18 From urban economic theory we know that residential markets sort house-  
19 holds by income along an urban-suburban continuum and the housing/transport  
20 cost trade-off (Fujita, 1989). This trade-off can, however, be dominated in  
21 the presence of exogenous central amenities (e.g. parks) or endogenous effects  
22 (e.g. high income attracting high income) and pull better-off households to-  
23 wards the center, adding discrepancies in accessibility to green amenities by  
24 different socio-economic groups (Brueckner et al., 1999). Tiebout's hypoth-

25 esis implies that marginal benefits from localized amenities are the same  
26 for all households in a given location and that the market leads to efficient  
27 sorting of households by income based on the spatial distribution of the local  
28 amenities. However, empirics (Hoyt & Rosenthal, 1997) point to non-efficient  
29 sorting and endogenous socio-economic sorting in the demand for localized  
30 amenities in general, especially due to self-selection and geographically vary-  
31 ing attitudes.

32 Self-selection in the choice of residential location in proximity to UGS has  
33 been suggested by Krizek (2003) or Maat & de Vries (2006). Socio-economic  
34 characteristics of residential communities influence expectations and demand  
35 for UGS (Byrne & Sipe, 2010; Lo & Jim, 2010). Having a private garden  
36 provides recreational alternatives (Heynen et al., 2006; Ibes, 2015) or owning  
37 a car attenuates the cost of distance (Chang & Liao, 2011) and, thus, alter  
38 the effect of distance to UGS. Also UGS benefits depend not only on the  
39 provision of UGS but also on residents' perception and actual valuation of  
40 ecosystem services (Brown, 2008). Therefore, researchers now argue that we  
41 should further investigate how UGS are distributed relative to social needs  
42 (Nicholls, 2001; Talen, 2010; Byrne, 2012; Lin et al., 2014; Ibes, 2015) and  
43 that distance to UGS should not be considered as the unique limiting factor  
44 of use (Van Herzele & Wiedemann, 2003; Schipperijn et al., 2010).

45 The present study aims at assessing attitudes towards UGS and use of  
46 public UGS by residents of the region of Brussels in Belgium, including joint  
47 effects of individual socio-economic and locational characteristics of the res-  
48 idential environment. Our analysis is based on a survey that includes users  
49 and non-users of UGS from a variety of residential locations, namely having

50 good or bad provision of local green, or being close or distant to the city  
51 centre.

52 The remainder of this article is structured as follows. Section 2 takes stock  
53 of literature and outlines our contribution. Section 3 describes our survey  
54 data and complementary GIS and exploratory analysis. Section 4 presents  
55 descriptive statistics and results of the econometric analysis. Section 5 offers  
56 a discussion and Section 6 concludes.

## 57 **2. State-of-the-art and contribution**

### 58 *2.1. Preferences and attitudes towards UGS*

59 People's attitudes towards UGS have been considered in previous studies  
60 (Balram & Dragičević, 2005; López-Mosquera & Sánchez, 2011; Kabisch &  
61 Haase, 2014; Ives et al., 2017), while few have focussed on people actually  
62 declaring UGS benefits Jim & Chen (2006); Bertram & Rehdanz (2015).  
63 Expressing the value of UGS in monetary units also helps to understand user  
64 preferences and the relative value people place on direct (e.g. recreational)  
65 and indirect (e.g. health) benefits of UGS (Jim & Chen, 2006; de Groot  
66 et al., 2012). There are multiple approaches to quantify attitudes towards  
67 UGS in monetary terms, examples being hedonic pricing (e.g. Tyrväinen &  
68 Väänänen, 1998; Bolund & Hunhammar, 1999; Lutzenhiser & Netusil, 2001;  
69 Dehring & Dunse, 2006; Jim & Chen, 2006; Tyrväinen et al., 2007; Sander &  
70 Zhao, 2015), contingent valuation (e.g. Tyrväinen & Väänänen, 1998; Balram  
71 & Dragičević, 2005; Bateman et al., 2009) or choice experiments (e.g. Christie  
72 et al., 2007; De Valck et al., 2014, 2017).

73 *2.2. Locational effects and self-selection*

74 Few studies provide comprehensive statistical analyses of which factors  
75 influence attitudes (e.g. Jim & Chen, 2006; López-Mosquera & Sánchez, 2011;  
76 De Valck et al., 2017), the use of UGS (e.g. Giles-Corti et al., 2005; Maat  
77 & de Vries, 2006; Schipperijn et al., 2010) and the distance travelled to use  
78 UGS (e.g. Schipperijn et al., 2010). Even fewer studies adopt a comprehen-  
79 sive perspective linking those three aspects. Bell et al. (2003) explain the  
80 use of urban fringe woodlands by different age groups with individuals' pref-  
81 erences and accessibility. They highlight the differences in attitudes per age  
82 group and subsequent use patterns. Lin et al. (2014) compare availability  
83 and accessibility effects to attitude effects on frequency and duration of urban  
84 park use. They find that opportunity is the principal driver for use and that  
85 people with a strong attitude towards nature are more likely users. Zhang  
86 et al. (2015) address the level of satisfaction during physical activities in pub-  
87 lic UGS and reasons for not using green space in proximity. They find that  
88 the living context, vegetation quality and accessibility influence satisfaction  
89 levels; further, physical activities and UGS size explain UGS use patterns.  
90 Tyrväinen et al. (2007) study the spatial distribution of UGS values derived  
91 by inhabitants based on attitudes towards and use of green areas, socio-  
92 demographic characteristics and childhood environment. The study stresses  
93 the need to understanding values of UGS for local residents with different  
94 characteristics. Maat & de Vries (2006) test the compensation theory where  
95 an individual who lives in a less green environment would value a park more  
96 and visit it more often. They find no such compensation behaviour, which  
97 they explain by self-selection within the residential choice process.

98 *2.3. Capturing joint effects*

99 We aim to contribute to the literature on UGS benefits by quantifying  
100 and explaining (i) people’s stated attitudes towards UGS, (ii) the influence  
101 of these attitudes along with locational and socio-economic aspects on the  
102 frequency of UGS use and (iii) the travel distance to UGS. We therefore  
103 attempt to reveal interdependencies between socio-economic, locational and  
104 attitudinal effects on patterns of UGS use. This paper also constitutes an  
105 innovation in that it considers three dimensions of attitudes: a direct mon-  
106 etary valuation of UGS (willingness to pay), a trade-off with housing space  
107 (willingness to trade housing space for the provision of UGS) and a poten-  
108 tial substitution with private green space (willingness to substitute private  
109 green space for public green space). Moreover, most studies on the impact  
110 of UGS access and attitudes on use originate from the United States, the  
111 United Kingdom, Australia and a growing number from China (Wolch et al.,  
112 2014). Our analysis of Brussels, thus, provides empirical knowledge on a  
113 continental European city where the socio-spatial distributions of UGS may  
114 differ from other parts of the world. The spatial distribution of activities,  
115 including UGS, in European cities derive from a long history of urbanisa-  
116 tion. In Brussels, there is a gradient of increasing vegetation cover from the  
117 medieval centre to the more recent suburban developments (Van de Voorde,  
118 2017). The spatial distribution of socio-economic profiles does not strictly  
119 follow this gradient.

### 120 **3. Data and methods**

#### 121 *3.1. Study area and survey*

122 The study area is the capital region of Brussels, with a population of 1.2  
123 million and an area of 161  $km^2$  (BISA, 2017). With  $>30m^2$  per capita UGS  
124 provision, Brussels offers a high percentage of UGS coverage in comparison to  
125 other European agglomerations with more than 100 000 inhabitants (Fuller  
126 & Gaston, 2009). UGS represents more than 40% of the urban area and the  
127 mean UGS size is 2.3 ha (OSM, 2016). The largest UGS (Bois de la Cambre,  
128 123 ha) is situated in the South East of the study area (Figure 1).

129 In June 2016, a face-to-face survey was conducted among 540 residents of  
130 the study area in 18 non-park public places (for instance commercial centres,  
131 street intersections, public squares). Non-parks spaces were needed in order  
132 to reach both non users and users of parks, and in order for informal UGS to  
133 be declared by respondents. The survey sites were chosen along two urban-  
134 suburban transects and across neighbourhoods differently provisioned with  
135 green space. The aim was to reach respondents with various socio-economic  
136 backgrounds within different combinations of residential environments (e.g.  
137 central but green, central without green, leafy suburbs) in line with litera-  
138 ture on sorting and amenities (Tiebout). The survey was conducted by one  
139 of the authors and a graduate student trained as interviewer. The question-  
140 naire (available online at [greenspace.uni.lu](http://greenspace.uni.lu)) consists of closed questions, but  
141 the face-to-face interaction permitted clarifications in order to increase the  
142 reliability of answers.



143 *3.2. Socio-economic characteristics*

144 Individual and household data were retrieved, including age, occupational  
145 status, household size, highest level of diploma, car ownership and nation-  
146 ality. From the literature, we expect they influence both attitudes towards  
147 UGS and the choice of residential location. Expected causal relationships are  
148 described in Section 3.5.

149 Multiple correspondence analysis (MCA) and clustering (k-means along  
150 the first 2 correspondence axes) were used to reduce dimensionality and ap-  
151 proximating income level, leading to 4 distinct socio-economic profiles (Table  
152 A.1 in the appendix). We depict the groups as : "young/students", which  
153 include predominantly respondents of age less than 25 or with student status;  
154 "low", which represent a low socio-economic situation mainly explained by  
155 being members of a single or couple household, above the age of 55 and being  
156 at home as main occupation; "medium", which represent a mixed-up group of  
157 either educated multi-person households with members working part-time/  
158 being at home or educated single households with full-term occupation and  
159 therefore medium household income; and finally "high", predominantly full-  
160 time employees aged 25 to 55 with university degrees, living in multi-person  
161 and multi-worker households, often owning 2 cars. Residential locations of  
162 respondents are mapped by socio-economic profiles in Figure 1, showing no  
163 particular spatial pattern, by construction, given the selection of survey sites  
164 and respondents explained above.

165 In addition to socio-economic profiles, we asked respondents whether they  
166 own a dog, assuming that walking the dog triggers a daily use of UGS.

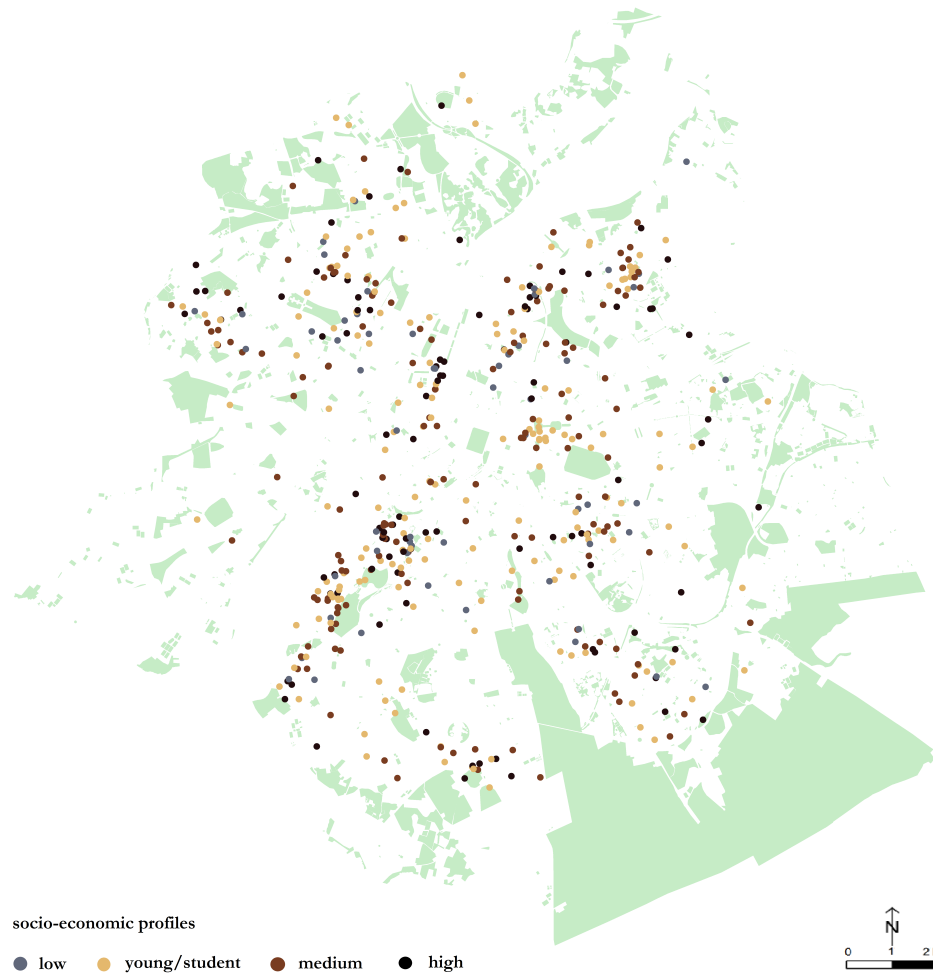


Figure 1: Map of respondents' residential location, classified by socio-economic profiles as result of a MCA and cluster: grey - low income, yellow - young/student, red - medium income, black - high income. All available UGS are shown as light green polygons (OSM data).

167 *3.3. Residential environment*

168 Respondents were asked to point their home on a map, from which we  
169 derived residential characteristics. The Euclidean distance between the res-  
170 idential place and the city hall was calculated in order to capture centrality  
171 effects and the related housing/transport cost trade-off (Fujita, 1989). In or-  
172 der to test for compensatory effects respondents were asked whether they had  
173 access to a private garden (i.e. private green which is not publicly accessible)  
174 or a balcony.

175 We computed two objective measures of UGS provision around residential  
176 locations: the network distance by foot to the nearest UGS and the share of  
177 UGS within a 400m buffer (in line with Van Herzele & Wiedemann (2003) and  
178 which is also the distance which the first quartile of our respondents travels  
179 to their most used UGS). The OSM network data was used to determine  
180 cost paths to all UGS and identify the closest UGS. The UGS considered for  
181 describing the residential environment were taken from OSM, in accordance  
182 with the work of Le Texier et al. (forthcoming) who find that OSM is a good  
183 compromise between a remotely sensed vegetation map (Landsat NDVI) and  
184 official cadastral data in order to retrieve UGS. A subjective measure of  
185 UGS was also retrieved: respondents stated their level of satisfaction with  
186 the quality of UGS within a 5-minutes-walk from their home.

187 *3.4. Stated attitudes and use of UGS*

188 Individual attitudes towards UGS were recorded through three dimen-  
189 sions: willingness to (i) pay for the provision of UGS (WTP), (ii) substitute  
190 housing space for the provision of UGS (WTSH) and (iii) substitute private  
191 green space for UGS (WTSG). For the first attitude (WTP), respondents

192 were asked with an open-ended question to state their willingness to pay  
193 through local annual taxes for free use of UGS in the region, given that the  
194 city spends currently 1.18EUR/year per inhabitant on UGS (authors' own  
195 computation from allocated municipal budget; precise questions available at  
196 greenspace.uni.lu). The responses were transformed ex-post into a scaled in-  
197 dex to ease generalization since the budget for UGS may vary from year to  
198 year and the analysis aims at identifying distinct respondent groups rather  
199 than reporting EUR amounts. The data-driven scaling is also a response  
200 to potential biases of contingent valuation analysis as discussed by Boyle  
201 (2003) and, thus, translates the stated monetary valuation into distinctive  
202 groups of respondents. For the second attitude dimension, WTSH, respon-  
203 dents were asked when deciding where to live whether they would accept  
204 living in a smaller house if there is satisfying green space nearby. Similarly,  
205 the third dimension, WTSG, was captured by asking whether they would in  
206 this situation accept a smaller garden.

207 Respondents stated their frequency of UGS use in a discrete manner  
208 ranging from "never/seldom" to "daily".

209 Following for instance Tyrväinen et al. (2007); Zhang et al. (2015); Arn-  
210 berger & Eder (2012), respondents located the UGS they use the most on  
211 a map. An UGS was defined during the survey as any open space in the  
212 city, such as parks, forests or green-ways, that is publicly accessible. No fi-  
213 nite set was presented to the respondents. The size of all most used UGSs  
214 was retrieved from Open Street Map (OSM) data. In addition, respondents'  
215 homes and their most used UGS were paired to retrieve network based travel  
216 distances

217 Tables 1 and 2 provide overall descriptive statistics for the attitudes and  
 218 use variables, and means per socio-economic profiles.

Table 1: Share of respondents and averages for variables used in the analysis.

Variable	Share of respondents [%]
<i>Willingness to pay for UGS</i>	
Less	6.54
Equal $\equiv$ 1.18 EUR/year	24.23
More	39.62
Much more	29.62
<i>Willingness to substitute housing space for UGS</i>	
Not at all	37.12
Somewhat	23.08
Strongly	39.81
<i>Willingness to substitute private green space for UGS</i>	
Not at all	35.77
Somewhat	20.77
Strongly	43.46
<i>Frequency of UGS use</i>	
Never / Seldom	12.12
Monthly	14.04
Weekly	30.58
Several times per week	27.69
Daily	15.58
<i>Provision of UGS near residential location</i>	
< 1 % $\equiv$ < 0.6 ha	9.42
1 - 5 % $\equiv$ 0.6 - 3.1 ha	36.15
5 - 15 % $\equiv$ 3.1 - 9.2 ha	34.04
> 15 % $\equiv$ > 9.2 ha	20.38
<i>Satisfaction with quality of UGS near residential location</i>	
Not at all	14.62
Somewhat	28.27
Very	57.12
<i>Socio-economic profiles</i>	
young / student	11.80
low income	20.08
medium income	31.68
high income	36.43
	Mean
Eucl. distance from residential location to city hall [ <i>km</i> ]	3.40
Walking network distance from residential location to most used UGS [ <i>km</i> ]	2.44
Walking network distance from residential location to nearest UGS [ <i>km</i> ]	0.16
Size of most used UGS [ <i>km</i> <sup>2</sup> ]	1.36

219 Figure 2 shows the links between residential locations and the UGS each

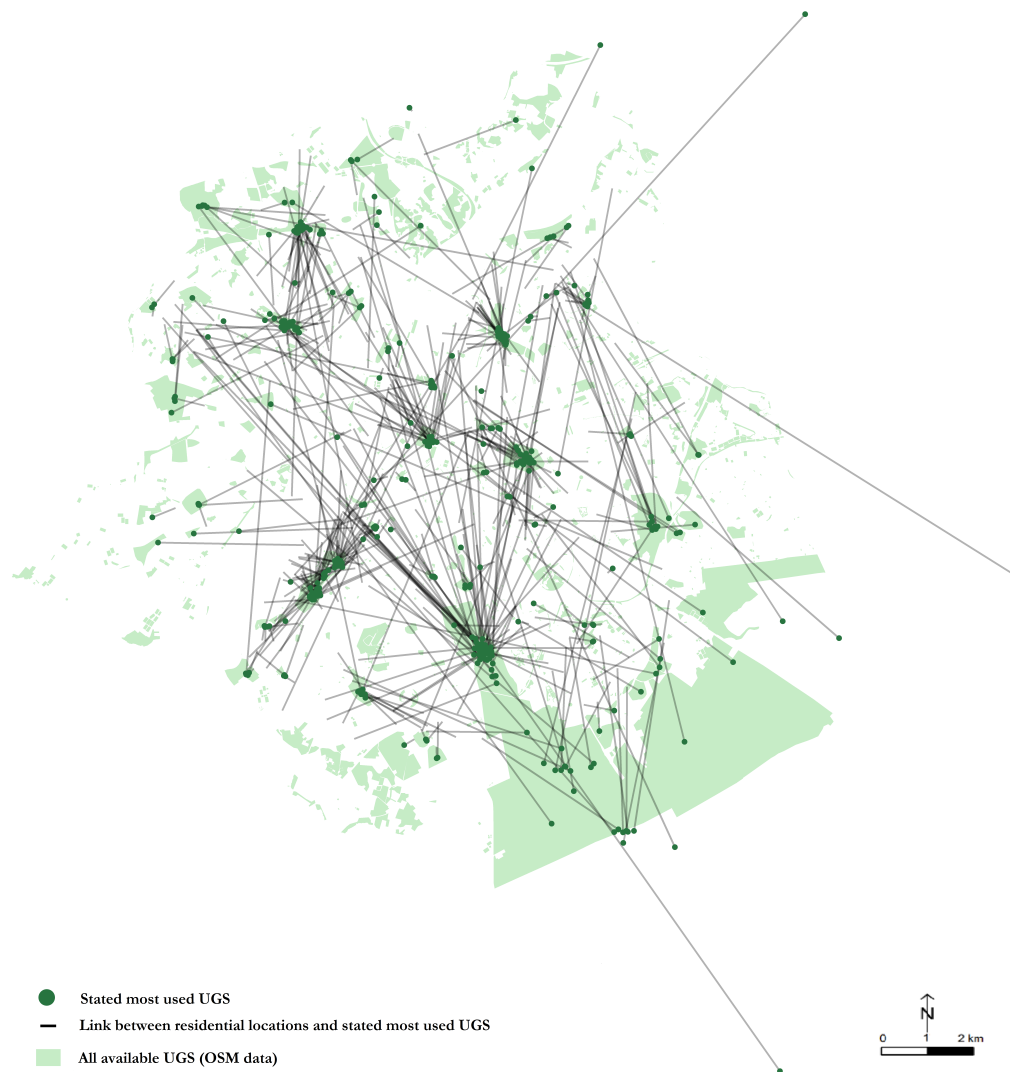


Figure 2: Map of links (black segments) between residential locations of respondents and stated most used UGS (dark green dots). All available UGS in Brussels are shown as light green polygons (OSM data).

Table 2: Means per socio-economic profile.

Variable	low	young/student	medium	high
Distance residential location - city hall	3.54	3.16	3.38	3.44
Distance residential location - most used UGS	2.39	2.26	2.31	2.67
Distance residential location - nearest UGS	0.15	0.15	0.18	0.17
Size of most used UGS	1.05	0.28	1.00	1.40

220 respondent stated to use the most (green dots). All available UGS patches  
 221 are shown as grey polygons (OSM data). Most respondents do not use the  
 222 UGS nearest to their home location. The average network walking distance  
 223 to the most used UGS is 2.54km, while it is 0.16km to the nearest UGS.

### 224 3.5. Hypotheses and econometric design

225 In the absence of a theoretical base that would explicitly link residential  
 226 sorting with UGS attitudes and use in terms of frequency and distance, we  
 227 conduct our exploration along the following three hypotheses. Everything  
 228 else being equal, we expect:

- 229 1. *stated positive attitudes towards UGS*(WTP, WTSH, WTSG) to in-  
 230 crease with higher socio-economic status, having children or a dog,  
 231 higher provision of and satisfaction with the green space around resi-  
 232 dential locations, distance to the centre and lower compensatory capa-  
 233 bilities (e.g. no private garden or balcony).
- 234 2. *frequency of UGS use* to increase with accessibility and availability of  
 235 UGS in proximity of residential locations, positive attitudes towards  
 236 UGS and higher satisfaction with the quantity and quality of UGS in  
 237 residential proximity.
- 238 3. *distance to the most used UGS* to either decrease or increase with pos-  
 239 itive attitudes towards UGS since households choose residential loca-

240 tions where UGS provision is high or choose UGS which satisfy certain  
241 valued criteria.

242 Comparing the different attitudes towards UGS, different factors are ex-  
243 pected to influence each, suggesting to consider them separately in the anal-  
244 ysis. First, monetary valuation of UGS (WTP) is likely determined by socio-  
245 economic characteristics. Second, from urban economic theory we expect the  
246 trade-off effect (WTSH) to be additionally driven by an urban-suburban gra-  
247 dient as housing prices tend to decrease with distance to the centre. Finally,  
248 substitution effects (WSTG) are expected to be influenced by the availability  
249 and quality of compensatory capabilities.

250 Given these assumptions and the nature of the data collected, we con-  
251 struct a three-step approach, modelling successively i) attitudes, ii) frequency  
252 of UGS use and then iii) distance to the most used UGS. Attitudes and fre-  
253 quency of UGS use being categorical, our first two models rely on multinomial  
254 logistic regression (MLR). We did not employ ordinal logistic regression since  
255 the assumption of proportional odds between categories did not hold. Our  
256 third model, i.e. for distance to the most used UGS, is a quantile regres-  
257 sion to retrieve separate coefficients for different ranges of distance. This is  
258 justified by our third hypothesis that distance effects are not linear.

## 259 4. Results

260 The sequence of 3 models is presented for each attitude dimension in turn,  
261 i.e. WTP (4.1), WTSH (4.2) and WTSG (4.3). The models presented were  
262 selected based on goodness of fit. The goodness of fit of the models was tested  
263 with the Hosmer-Lemeshow test, the accuracy rate of correct prediction and



264 the standard errors as reported in appendix B. For clarity of the expos,  
265 coefficients and relative probabilities for the MLR models of attitudes and  
266 frequency are deferred to appendix B and quantile regression estimates for  
267 the distance travelled are deferred to appendix C. In the text, we keep a  
268 graphical scheme summarizing results for each attitude dimension in turn.

#### 269 *4.1. Relation between WTP, frequency of use and distance to most used UGS*

##### 270 *4.1.1. Willingness to pay for UGS (WTP)*

271 Table B.1 shows the coefficients and probabilities resulting from the MLR  
272 model with WTP as dependent variable. We find that WTP is significantly  
273 determined by socio-economic profiles, with higher profiles (i.e. indirectly  
274 higher income) willing to pay more. Interestingly, socio-economics is the  
275 only variable influencing WTP for UGS. Satisfaction with the quality of  
276 UGS within a 5-minutes walk, compensatory opportunities (balcony, private  
277 garden) or urban-suburban differences play no role.

##### 278 *4.1.2. Frequency of UGS use*

279 As expected, households' attitudes significantly influence how often re-  
280 spondents use UGS. In general, the higher the willingness to pay for UGS is,  
281 the more likely are respondents to use UGS more often (Table B.2). Con-  
282 versely, a respondent who stated a low valuation of UGS is most likely to  
283 use UGS seldom or never. Hence, since the "low" group is associated with  
284 the highest probability among all groups to express a WTP of less than 1.18  
285 EUR/year, the "low" group is also more likely to use UGS only seldom or  
286 never. The association between WTP and daily visits is, however, not sig-  
287 nificant. Daily visits are instead well explained by the use of UGS to walk

288 the dog and a strong satisfaction with the quality of UGS within a 5-minutes  
289 walk from residential locations.

290 In sum, WTP for UGS (controlled for respondents' socio-economic profile)  
291 significantly explains frequent visits of UGS, while the daily use of UGS is  
292 rather explained by the purpose of UGS use (e.g. walking the dog) and the  
293 quality of UGS provision. Using WTP as reference for residents' valuation of  
294 UGS suggests that non-use of UGS is, thus, a choice rather than the result  
295 of constraints. A lack of quantity of UGS provision or the urban-suburban  
296 gradient show no explicit influence and were, thus, removed from the model.

#### 297 *4.1.3. Distance to most used UGS*

298 We expect the frequency of visits to influence the distance travelled. Yet,  
299 since we observed that attitudes influence the frequency of use, we choose only  
300 the attitudes as potential explanatory factor. Results are reported in Figure  
301 C.1. Interestingly, WTP or urban-suburban differences show no significant  
302 effect, although larger distances are associated with suburban locations and  
303 low WTP. The distance respondents travel to visit their most used UGS is  
304 significantly associated with the quantity of UGS around their residential  
305 location and characteristics of the most used UGS, i.e. its size. A low  
306 provision of UGS ( $< 5\%$ ) and the size of UGS explain, therefore, extra travel  
307 to UGS. While the relationship between the size of the most used UGS and  
308 distance travelled is linear, less than 1% of UGS around a residential location  
309 shows peak effects on the 0.7 distance quantile. Note that we do not observe  
310 a simple significant urban-suburban effect and that there is no strict linear  
311 increase of the quantity of UGS around residential locations with distance to  
312 the centre.

313 Figure 3 summarises the findings from the three-step modelling approach  
 314 using WTP. Socio-economic characteristics influence WTP, which in turn  
 315 influences the frequency of use (but not distance) along with characteristics  
 316 of the residential environment.

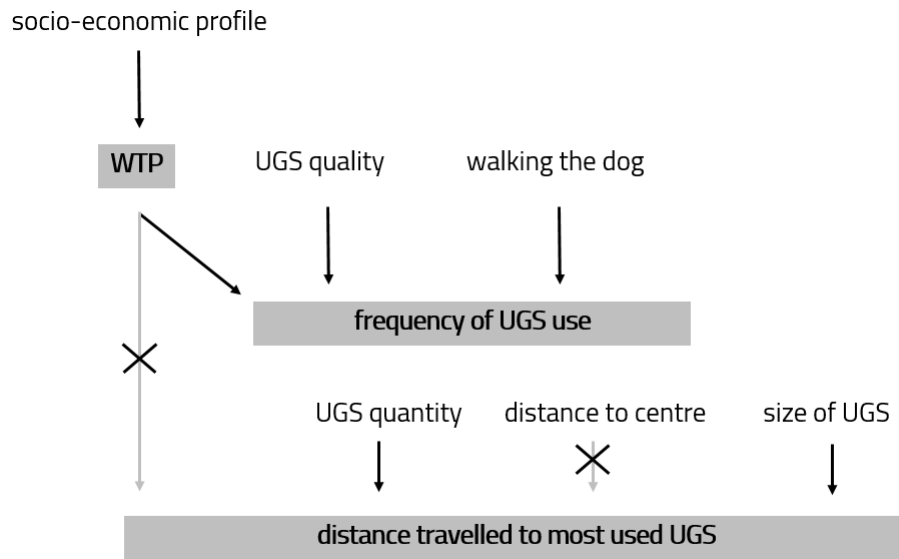


Figure 3: The Figure depicts the three-step modelling approach with the three dependent variables (grey boxes) i) respondents' willingness to pay for UGS (WTP), ii) the frequency of UGS use and iii) the distance which respondents travel to their most used UGS. Black arrows indicate the factors that significantly influence each of the dependent variables.

317 *4.2. Relation between trade-offs of housing space and UGS, frequency of use*  
 318 *and distance to most used UGS*

319 *4.2.1. Willingness to trade-off housing space for UGS (WTSH)*

320 From urban economic theory we expect respondents who value housing  
 321 space to live further away from the centre. At the same time, suburban res-  
 322 idents also value proximity to green spaces (private or public). As expected,

323 we find a significant urban-suburban gradient effect: compared to residents  
324 living less than 1km away from the city hall, suburban residents are signifi-  
325 cantly less likely to trade-off housing space for UGS (Table B.3). We conclude  
326 on a pure urban-suburban effect and no significant effect of perceived quality  
327 of UGS near locations.

328 Socio-economic characteristics influence the trade-off as expected from  
329 urban theory (Brueckner et al., 1999). Respondents of the group "medium"  
330 have the highest probability to be strongly in favour of the trade-off, while the  
331 groups "low" and "young/student" are less likely. This could be explained  
332 by the "medium" groups likely having children and, thus, searching public  
333 places to spend time outside (e.g. Lo & Jim, 2010); while "low" groups, that  
334 is mainly retired residents, value housing space since they spend more time  
335 at home. "High" groups, however, are associated with a surprisingly high  
336 probability of not being willing to trade off housing for UGS space.

#### 337 *4.2.2. Frequency of UGS use*

338 WTSH explains significantly why respondents are more likely to visit UGS  
339 at least on a monthly basis (Table B.4). To explain daily visits, however, the  
340 provision of local UGS (quantity and perceived quality) and the main reason  
341 to use UGS (in this case walking the dog) are significant. Respondents who  
342 stated no willingness to trade off housing for UGS space (most likely low  
343 income groups and students) are the ones most likely to never/seldom visit  
344 UGS. Locations with less than 5% UGS are less likely to encourage UGS  
345 usage more than once per week. Good perceived quality of UGS raises the  
346 relative probability of daily UGS use. In sum, these results suggest that  
347 respondents use UGS seldom because their valuation of UGS is low rather

348 than due to a lack of provision of UGS.

#### 349 *4.2.3. Distance to most used UGS*

350 We find that the trade-off between housing and UGS space has no signif-  
351 icant influence on the distance travelled to the most used UGS. The quality  
352 and quantity of UGS provision and the size of the most used UGS significantly  
353 explain the distance. A provision of less than 1% explains in particular why  
354 respondents travel further to use UGS (around the 0.7 quantile); effects of  
355 low provision on small (<0.4 quantile) and large (>0.8 quantile) distances are  
356 lower (Figure C.2). This non-linearity suggests an underlying self-selection  
357 by respondents on their residential location. Conversely, as the satisfaction  
358 with local UGS quality decreases, respondents travel further to use UGS. No  
359 significant urban-suburban effect is observed.

360 Figure 4 summarises the findings from the three-step modelling approach  
361 using WTSH. The willingness to trade off has a significant urban-suburban  
362 structure after controlling for socio-economic variation.

#### 363 *4.3. Relation between substitution of private for public green space, frequency 364 of use and distance to most used UGS*

##### 365 *4.3.1. Willingness to substitute private for public green space (WTSG)*

366 Interestingly, respondents who are satisfied with the quality of UGS nearby  
367 their residential location are more likely to trade off private with public green  
368 space (Table B.5). Although we find that satisfaction with local UGS quality  
369 is generally higher in suburban locations, the urban-suburban gradient alone  
370 does not explain the preference significantly. "Low" groups have the lowest  
371 probability to be in favour of such a trade-off, while all other groups are less

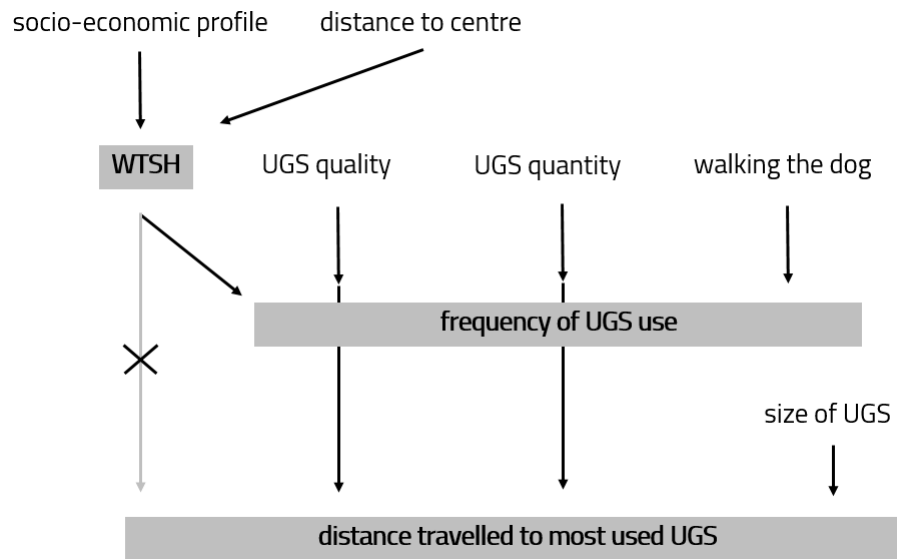


Figure 4: The Figure depicts the three-step modelling approach with the three dependent variables (grey boxes) i) respondents' willingness to trade off housing for public urban green space (WTSH), ii) the frequency of UGS use and iii) the distance which respondents travel to their most used UGS.

372 likely to reject. This is contradictory to Talen (2010) who finds that UGS can  
 373 substitute for private green space for apartment dwellers but not for owners  
 374 of houses with private gardens.

#### 375 4.3.2. Frequency of UGS use

376 We do not find a significant effect of substitutability of private for public  
 377 green space on the frequency of use (Table B.6). Instead, a low quantity of  
 378 UGS around residential locations is associated with a higher probability of  
 379 fewer visits to UGS. This suggests that substitutability of private for public  
 380 green space is a weak indicator of frequency of use since green spaces fulfil  
 381 different functions at different levels (private vs. public).

382 *4.3.3. Distance to most used UGS*

383 Although the trade-off between private and public green space does not in-  
384 fluence the frequency, it does influence the distance travelled. Strong WTSG  
385 significantly reduces the distance travelled, in particular at large quantiles  
386 (Figure C.3). This suggests that people who are the most willing to trade  
387 private UGS for public ones are the ones who use the closest ones as if the  
388 public space was a continuity of the private place. Since the substitutability  
389 was shown to partly depend on the perceived quality of nearby UGS, qual-  
390 itative provision of UGS can lead to a reduction in travel distance to use  
391 UGS. This indirect relation also suggests that the economic value of UGS is  
392 already considered in the choice of residential location.

393 The quantity of UGS provided around residential locations affects the  
394 distance travelled (especially around the 0.7 quantile). Thus, an indifference  
395 towards the spatial distribution of UGS is associated with respondents who  
396 travel further to use UGS. The constraint of UGS quantity and proximity  
397 explains why respondents travel further, despite again the size of the UGS  
398 which acts as a wide-range pull-factor. However, the non-linear effect of UGS  
399 quantity indicates underlying endogenous effects, that is residential sorting.

400 Figure 5 summarises the findings from the three-step modelling approach  
401 using WTSG. Substitution effects vary across socio-economic profiles and are  
402 dependent on locational aspects (perceived quality of local UGS). Attitudes  
403 as well as characteristics of the residential environment significantly influence  
404 patterns of UGS use.

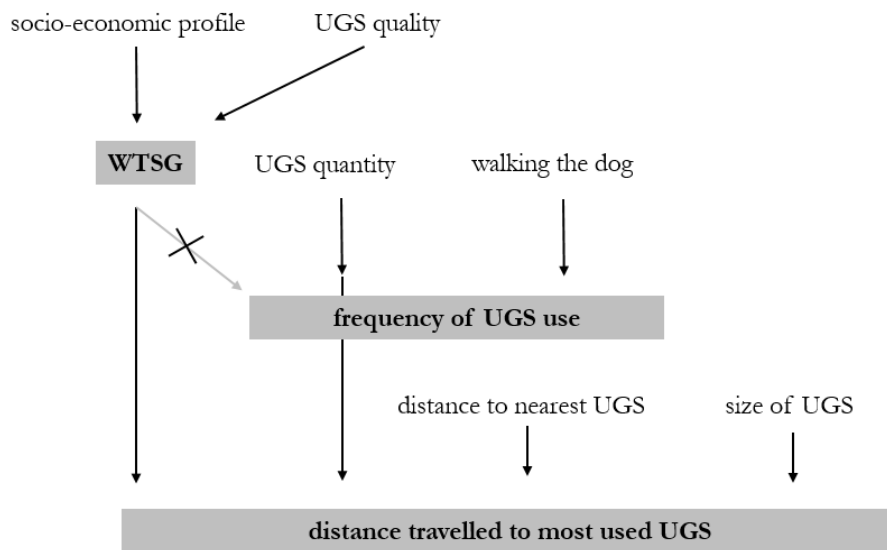


Figure 5: The Figure depicts the three-step modelling approach with the three dependent variables (grey boxes) i) respondents' willingness to private for public green space (WTSG), ii) the frequency of UGS use and iii) the distance which respondents travel to their most used UGS. Black arrows indicate the factors that significantly influence each of the dependent variables.



## 405 5. Discussion

406 Our analysis shows that there exist complex interdependencies and non-  
407 linear effects with regards to the valuation and the use of UGS. Those inter-  
408 dependencies require a careful consideration and generalisation from other  
409 cases before designing urban planning strategies. In this section, we high-  
410 light our main findings, pertaining to (i) attitudes and (ii) use of UGS, and  
411 confront them against empirical evidence obtained from other cases in the  
412 literature.

### 413 5.1. *Explaining attitudes towards UGS*

414 Our study highlights the importance of differentiating between the role  
415 of socio-economic background and spatial aspects to understand residents'  
416 attitudes.

417 *Socio-economic profiles influence attitudes.* Compared to other studies that  
418 look separately at single socio-economic criteria, we have retrieved socio-  
419 economic profiles. Our results indicate that these profiles influence attitudes  
420 towards UGS. More precisely, highly educated, full-time employed people  
421 from all age groups (i.e. indirectly higher income groups) are more likely to  
422 express strong positive attitudes (WTP, WTSH, WTSG) than older, retired  
423 and less educated respondents (i.e. indirectly lower income groups). This is a  
424 rejoinder to Sander & Zhao (2015) and Jim & Chen (2006) who respectively  
425 find an increase in WTP with income, and with income and age. We depart  
426 partially from Lorenzo et al. (2000) who find an influence of income on WTP,  
427 but not of age or education, and depart more strongly from Tyrväinen &

428 Väänänen (1998) who find no association between WTP and income. This  
429 suggests different influences of socio-economic profiles across attitudes.

430 *Local UGS provision explains attitudes towards UGS.* We have found that  
431 willingness to substitute private for public green space increases with the per-  
432 ceived quality of local UGS. This follows, for instance, Balram & Dragičević  
433 (2005); Ambrey & Fleming (2014); Brander & Koetse (2011) who find that  
434 attitude towards green may also be influenced by the spatial environment  
435 surrounding an individual. Since low income groups are in our study associ-  
436 ated with the lowest probability to substitute private green for public green  
437 space, our results suggest that low income people living in a poor quality UGS  
438 neighbourhood would be least likely among all groups to substitute private  
439 for public green space. From discussion with the respondents we can derive  
440 the assumption that the people with the most vulnerable socio-economic sta-  
441 tus are the ones who are the most afraid to lose private property in favour  
442 of public space. Thus, precarity of the economic situation is a potential  
443 explanation.

444 Respondents of our survey in Brussels, who already have satisfactory  
445 quantity and quality of UGS around their home location, tend to value UGS  
446 more. On these grounds, we can conclude that respondents (who were able  
447 to) already took the UGS distribution into consideration when choosing their  
448 residential location. This goes in the direction of Brueckner's theory (1997)  
449 which states that the relative location of different income groups depends on  
450 the spatial patterns of amenities within cities.

451 *Public and private green play different roles.* We do not find explicit com-  
452 pensatory effects as suggested, for instance, by Maat & de Vries (2006) who

453 test whether someone living in a less green environment would value UGS  
454 more. Dehring & Dunse (2006) show that the value given to public UGS  
455 varies according to the property type households live in. Tu et al. (2016)  
456 find that preferences are heterogeneous due to private garden ownership.  
457 Talen (2010) finds that apartment dwellers would more likely substitute pri-  
458 vate green space for UGS than owners of homes who likely have a private  
459 garden. Our results rather suggest that public and private green space differ  
460 in the ecosystem services they provide and are only partly substitutable. Like  
461 Barbosa et al. (2007) we find that wealthier households are more likely to  
462 make such a trade-off, but show that socio-economic background alone does  
463 not explain such preferences.

464 *Housing space trade-off depends on centrality.* A willingness to trade off pri-  
465 vate housing space for public space exists and is directly controlled by the  
466 urban-suburban structure. In contrast, a willingness to pay or substitute  
467 private green for more UGS is not related to centrality. To the best of our  
468 knowledge this centrality effect has not been tested earlier as such. It suggests  
469 that when dealing with green space substitution issues, bundling housing and  
470 garden consumptions within a single spatial good as is usually the case in  
471 urban economics is not relevant.

## 472 *5.2. Explaining patterns of UGS use*

473 Our analysis suggests that neither UGS provision nor attitudes alone can  
474 explain the frequency of UGS use or the distance travelled to use UGS; in  
475 fact, both need to be considered in conjunction.

476 *Weekly use relates to attitudes while local provision matters for daily use.*  
477 We find that attitudes (WTP and WTSH) explain whether someone uses  
478 UGS frequently, while the provision of UGS (quantity, quality and proximity)  
479 explains in particular the daily use of UGS. This second finding is in line  
480 with Tyrväinen & Miettinen (2000); Maat & de Vries (2006) who find that  
481 availability of UGS tends to increase the use of UGS.

482 *Non-linear proximity effects on travel distance to UGS.* In addition, we find  
483 that local provision of UGS has no linear effect on the distance to the most  
484 used UGS. Lack of provision, therefore, does not univocally lead to larger  
485 travelled distances to UGS, but might indicate indifference to the spatial  
486 distribution of UGS relative to someone's home location.

487 Lanzendorf (2000) shows that residents in low-density neighbourhoods  
488 comprising detached houses with gardens travel shorter distances than res-  
489 idents in a more urbanized neighbourhood in Cologne. According to Van  
490 Herzele & Wiedemann (2003) respondents use UGS at different distances  
491 from their home based on the functional levels of UGS. In Brussels, respon-  
492 dents travel further to use larger UGS as also found by Giles-Corti et al.  
493 (2005) in Perth. Moreover, we find that they travel further in case they ex-  
494 press preferences towards private rather than public green space or because  
495 of a lack of provision near their home location. In addition, less frequent  
496 UGS use is associated with visits of UGS further from residential locations.

497 *Willingness to substitute private green influences travel distance to UGS.*  
498 The willingness to substitute private green is the only attitude dimension  
499 that significantly explains the distance to the most used UGS. In fact, the  
500 stronger the willingness is, the shorter is the maximum distance travelled,

501 that is the less people travel across the city to use UGS. Since the substitution  
502 effect also reflects the satisfaction with the quality of local UGS we conclude  
503 that the residential location is by choice in proximity to high quality UGS.

504 *No spatial heterogeneity of use beyond residential sorting.* Although the quan-  
505 tity of UGS is obviously higher in the North-West and South-East of Brussels,  
506 we do not find spatial patterns in stated attitudes, frequency of use or dis-  
507 tance travelled to use UGS. This is likely due to the careful consideration of  
508 survey locations with well assorted respondent groups.

509 Shafer et al. (2013) show that access and opportunity to use UGS depend  
510 on households' socio-economic status and location. Ernstson (2013) find that  
511 UGS is disproportionally available to a subset of the urban population. In  
512 Brussels, either or both socio-economic status and location only indirectly  
513 influence the use of UGS through residential sorting.

514 *No explicit urban-suburban differences in use and travel distance.* We do not  
515 observe an explicit urban-suburban effect in our econometric analysis. Dis-  
516 tance to the centre itself does not explain the use and distance to UGS  
517 significantly. However, there is residential sorting along the urban-suburban  
518 gradient. The quantity and perceived quality of UGS as well as the hous-  
519 ing/UGS trade-off are also subject to an urban-suburban gradient. Respon-  
520 dents who live further away from the centre are generally more satisfied with  
521 the quality of UGS and have a larger share of UGS nearby than those closer  
522 to the centre. Results suggest that attitudes towards UGS partly underlie  
523 the location choice of our respondents.

## 524 **6. Conclusion**

525 Urban green spaces provide benefits to the urban population. These ben-  
526 efits do not accrue equally to all residents because the use of green space  
527 depends on their socio-economic characteristics, residential location and at-  
528 titude towards green space and their complex interdependencies.

529 We have quantified residents' attitudes towards UGS in Brussels and their  
530 influence on patterns of UGS use while controlling for socio-economic aspects  
531 and the urban-suburban gradient.

532 We have provided evidence that both socio-economic and locational as-  
533 pects influence residents' attitude towards UGS and patterns of UGS use.  
534 Attitudes towards UGS expressed in monetary terms, as a trade-off with  
535 housing space or as a substitution with private green, are all explained by  
536 socio-economic characteristics. The effect of residential location varies across  
537 these dimensions. In addition, we have seen that attitudes add to locational  
538 and socio-economic effects to explain frequency of use and travel distance to  
539 UGS.

540 Our results contribute still rare empirical findings for a European city.  
541 We corroborate previous studies in the role of local UGS provision on UGS  
542 use but show that attitudes are key factors to be taken into account for  
543 understanding use. Moreover, attitudes themselves are impacted by socio-  
544 economic and locational variables.

545 Although this research is limited in its scope, our results suggest that a  
546 spatially equal distribution of UGS would not necessarily provide benefits  
547 to all residents. Our findings rather call for the provision of strategically  
548 distributed high quality UGS which meets user preferences and can, thus,

549 reduce the need for private green space and steer both the use of UGS and  
550 residential choices.

551 Future research should aim at an inter-city comparison for further gen-  
552 eralization by replicating the analysis based on the same survey conducted  
553 also in other European cities. Our results also stress the need for theorizing  
554 and comprehensively addressing endogenous and non-linear effects between  
555 residential choice and UGS practice. In particular, we revealed interdepen-  
556 dencies between the various explanatory factors, which highlights the need to  
557 better understand how residential preferences and constraints impact residen-  
558 tial sorting, hence observed differences in UGS use. Future research should  
559 be carefully designed to account for the revealed interdependencies between  
560 attitudes, socio-economic and locational variables and the use of UGS. With-  
561 out measured evidence and trade-offs, planning recommendations may well  
562 lead to inappropriate UGS locations and imbalances in use and benefits in  
563 the longer term after relocations and differences in monetary power across  
564 households.

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756 **Appendix**757 **A. Socio-economic profiles**Table A.1: Share [%] of observations per socio-economic profile and total numbers  $N$ .

Response	<i>Socio-economic profile</i>				<i>Total</i>
	low	student	medium	high	
$N$	63	191	160	106	520
<i>Age</i>					
<25	0.00	84.21	3.27	0.57	61
25-35	1.03	14.04	40.52	40.34	146
35-45	2.06	1.75	21.57	30.68	96
45-55	10.31	0.00	18.30	26.70	94
55-65	36.08	0.00	15.69	1.14	64
>65	50.52	0.00	0.65	0.57	59
<i>Diploma</i>					
Primary	23.71	10.53	10.46	2.27	50
Secondary	47.42	57.89	43.79	11.93	181
University	28.87	31.58	45.75	85.80	289
<i>Household size</i>					
Single	51.55	19.30	28.10	3.98	115
2	41.24	14.04	30.07	24.43	147
3 or 4	4.12	33.33	28.10	55.11	177
5 or 6	1.03	29.82	9.15	14.77	65
>6	2.06	3.51	4.58	1.70	16
<i>Occupation</i>					
At home	90.72	8.77	40.52	1.70	171
Parttime	4.12	15.79	13.73	11.36	56
Student	0.00	68.42	0.65	0.57	47
Working	5.15	7.02	45.10	86.36	246
<i>Nationality</i>					
European	84.54	64.91	75.16	86.93	418
Non-European	15.46	35.09	24.84	13.07	102

758 **B. MLR models**

Table B.1: Predicted probabilities of willingness to pay for UGS (WTP) and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: WTP</i>			
	Less (1)	Equal (2)	More (3)	Much More (4)
<i>Socio-economic profile</i>				
low	0.12	0.36	0.36	0.16
young/student	0.09	0.20 -0.115 (0.624)	0.26 0.028 (0.615)	0.42 1.345** (0.626)
medium	0.05	0.22 0.376 (0.516)	0.43 1.040** (0.502)	0.29 1.504*** (0.545)
high	0.04	0.20 0.539 (0.532)	0.46 1.378*** (0.517)	0.30 1.801*** (0.558)
Constant		1.070*** (0.335)	1.070*** (0.335)	0.223 (0.387)
Akaike Inf. Crit.				1,198.961
Prediction accuracy				0.43
<i>Note:</i>		*p<0.1; **p<0.05; ***p<0.01		

Table B.2: Predicted probabilities of frequency of UGS use and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: frequency of use</i>				
	Never/Seldom (1)	Monthly (2)	Weekly (3)	Several (4)	Daily (5)
<i>Willingness to pay</i>					
Less	0.30	0.10	0.18	0.25	0.17
Equal	0.15	0.13 0.977 (0.757)	0.32 1.242** (0.609)	0.28 0.772 (0.576)	0.12 0.361 (0.656)
More	0.08	0.17 1.941*** (0.739)	0.08 1.974*** (0.607)	0.31 1.578*** (0.571)	0.11 0.942 (0.645)
Much More	0.11	0.13 1.264* (0.755)	0.11 1.538** (0.612)	0.29 1.045* (0.579)	0.17 1.017 (0.646)
<i>Walking the dog</i>					
No	0.11	0.15	0.11	0.29	0.11
Yes	0.09	0.07 -0.515 (0.704)	0.09 -0.480 (0.591)	0.19 -0.246 (0.581)	0.48 1.672*** (0.540)
<i>Strong satisfaction quality</i>					
No	0.15	0.17	0.17	0.29	0.10
Yes	0.09	0.13 0.249 (0.366)	0.13 0.719** (0.321)	0.34 0.571* (0.324)	0.16 0.993*** (0.377)
Constant		-1.224* (0.705)	-0.858 (0.572)	-0.474 (0.532)	-1.338** (0.612)
Akaike Inf. Crit.					1,469.119
Hosmer-Lemeshow test			X-squared = 18.207, df = 28, p-value = 0.9208		
Prediction accuracy					0.34

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table B.3: Predicted probabilities of willingness to substitute housing space for UGS (WTSH) and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: WTSH</i>		
	Not at all (1)	Somewhat (2)	Strongly (3)
<i>Distance to centre</i>			
< 2 km	0.26	0.30	0.44
2-4km	0.34	0.25 -0.478 (0.338)	0.42 -0.323 (0.309)
4-6km	0.44	0.18 -1.076*** (0.374)	0.38 -0.667*** (0.326)
>6km	0.33	0.23 -0.520 (0.658)	0.43 -0.281 (0.575)
<i>Socio-economic profile</i>			
low	0.48	0.14	0.38
young/student	0.34	0.34 1.207*** (0.447)	0.32 0.178 (0.399)
medium	0.27	0.22 0.970** (0.387)	0.51 0.885*** (0.295)
high	0.34	0.27 0.967*** (0.364)	0.38 0.357 (0.285)
Constant		-0.635 (0.394)	0.118 (0.325)
Akaike Inf. Crit.			1,103.116
Hosmer-Lemeshow test			X-squared = 7.8527, df = 14, p-value = 0.8969
Prediction accuracy			0.47

*Note:* \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table B.4: Predicted probabilities of frequency of UGS use and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: frequency of use</i>				
	Never/Seldom (1)	Monthly (2)	Weekly (3)	Several (4)	Daily (5)
<i>Substitution housing space</i>					
Not at all	0.20	0.11	0.28	0.25	0.15
Somewhat	0.05	0.19 1.950*** (0.550)	0.39 1.717*** (0.495)	0.30 1.572*** (0.505)	0.06 0.488 (0.611)
Strongly	0.09	0.15 1.116*** (0.423)	0.30 0.870** (0.358)	0.31 1.028*** (0.362)	0.16 0.834** (0.402)
<i>Share of UGS within 441m</i>					
>15%	0.09	0.14	0.18	0.40	0.19
5-15%	0.08	0.16 0.182 (0.576)	0.35 0.739 (0.535)	0.27 -0.306 (0.505)	0.13 -0.294 (0.547)
1-5%	0.14	0.13 -0.603 (0.555)	0.35 0.173 (0.506)	0.26 -0.930* (0.475)	0.11 -1.047** (0.527)
<1%	0.12	0.15 -0.277 (0.823)	0.31 0.237 (0.756)	0.34 -0.475 (0.725)	0.09 -1.093 (0.905)
<i>Walking the dog</i>					
No	0.11	0.15	0.33	0.30	0.11
Yes	0.09	0.08 -0.438 (0.708)	0.18 -0.432 (0.593)	0.20 -0.190 (0.584)	0.45 1.628*** (0.542)
<i>Strong satisfaction w. quality</i>					
No	0.14	0.18	0.28	0.30	0.10
Yes	0.09	0.13 0.088 (0.376)	0.35 0.622* (0.328)	0.29 0.398 (0.332)	0.15 0.787** (0.383)
Constant		-0.416 (0.567)	-0.328 (0.518)	0.527 (0.479)	-0.301 (0.536)
Akaike Inf. Crit.					1,455.472
Hosmer-Lemeshow test	44 X-squared = 28.628, df = 32, p-value = 0.638				
Prediction accuracy					0.38

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table B.5: Predicted probabilities of willingness to substitute private for public green space (WTSG) and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: WTSG</i>		
	Not at all	Somewhat	Strongly
	(1)	(2)	(3)
<i>Satisfaction w. UGS quality</i>			
Not at all	0.40	0.18	0.42
Somewhat	0.32	0.29 0.696* (0.402)	0.39 0.111 (0.332)
Very	0.33	0.19 0.232 (0.382)	0.49 0.304 (0.296)
<i>Socio-economic profile</i>			
low	0.51	0.15	0.35
young/student	0.36	0.24 1.174** (0.479)	0.49 0.991** (0.395)
medium	0.30	0.23 0.981*** (0.379)	0.48 0.846*** (0.294)
high	0.33	0.22 0.847** (0.369)	0.45 0.690** (0.283)
Constant		-1.568*** (0.430)	-0.572* (0.323)
Akaike Inf. Crit.			1,024.443
Hosmer-Lemeshow test	X-squared = 4.048, df = 10, p-value = 0.9452		
Prediction accuracy	0.47		
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01		

Table B.6: Predicted probabilities of frequency of UGS use and model coefficients. Predicted probabilities for all categories are reported in each first row; below coefficients based on the listed reference levels; standard errors in brackets.

	<i>Dependent variable: frequency of use</i>				
	Never/Seldom (1)	Monthly (2)	Weekly (3)	Several (4)	Daily (5)
<i>Substitution private green</i>					
Not at all	0.14	0.14	0.35	0.24	0.14
Somewhat	0.10	0.17 0.568 (0.501)	0.34 0.346 (0.438)	0.32 0.646 (0.451)	0.09 -0.092 (0.546)
Strongly	0.12	0.15 0.218 (0.406)	0.27 -0.077 (0.347)	0.31 0.417 (0.357)	0.16 0.297 (0.401)
<i>Share of UGS</i>					
>15%	0.01	0.14	0.18	0.39	0.20
5-15%	0.09	0.16 0.251 (0.567)	0.35 0.763 (0.526)	0.26 -0.289 (0.498)	0.14 -0.281 (0.542)
1-5%	0.17	0.14 -0.547 (0.538)	0.34 0.086 (0.491)	0.25 -0.987** (0.462)	0.11 -1.183** (0.517)
<1%	0.12	0.16 0.027 (0.798)	0.30 0.356 (0.732)	0.33 -0.324 (0.704)	0.09 -1.032 (0.888)
<i>Walking the dog</i>					
No	0.12	0.16	0.33	0.29	0.11
Yes	0.09	0.08 -0.411 (0.700)	0.17 -0.391 (0.584)	0.20 -0.137 (0.578)	0.46 1.713*** (0.537)
Constant		0.152 (0.516)	0.616 (0.474)	1.072** (0.443)	0.428 (0.489)
Akaike Inf. Crit.					1,475.706
Hosmer-Lemeshow test					X-squared = 10.559, df = 24, p-value = 0.9918
Prediction accuracy					0.38

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

759 C. Quantile models

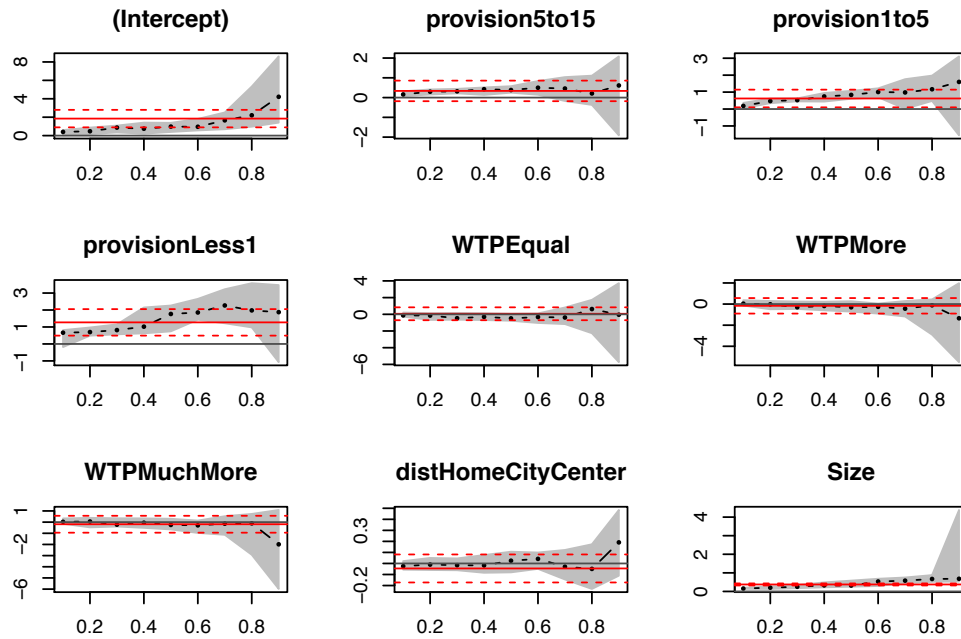


Figure C.1: Quantile regression analysis explaining the distance respondents travel to visit their most used UGS (network walking distance in km) with WTP. A UGS provision of less than 1% ( $p\text{-value} < 0.01$ ), between 1-5% and the size of the UGS are significant (both  $p\text{-value} < 0.001$ ).



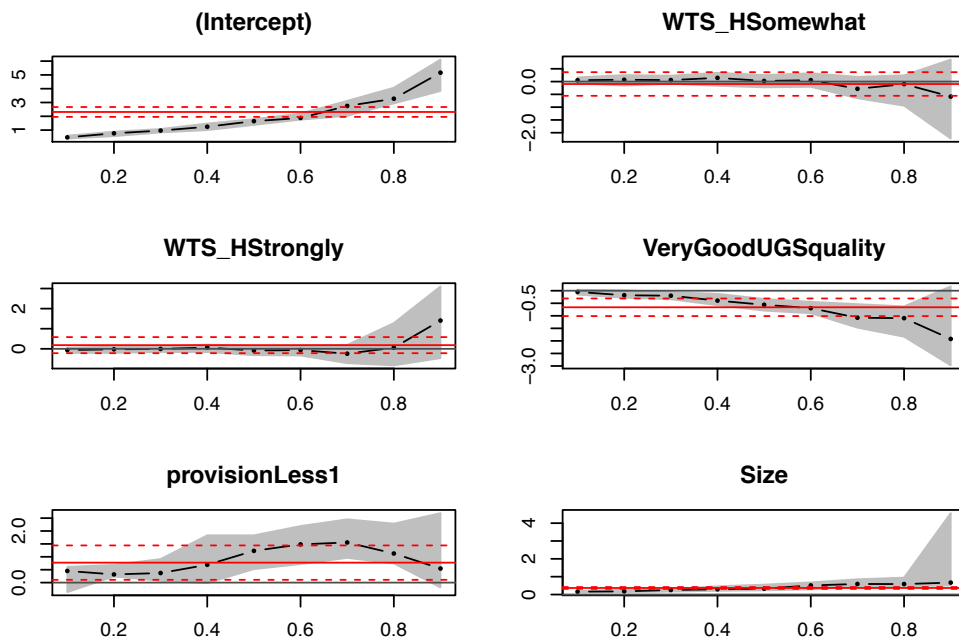


Figure C.2: Quantile regression analysis explaining the distance respondents travel to visit their most used UGS (network walking distance in km) with WTSH. A UGS provision of less than 1%, strong satisfaction with the quality of nearby UGS and the size of the UGS are significant with  $p\text{-value} < 0.001$ .

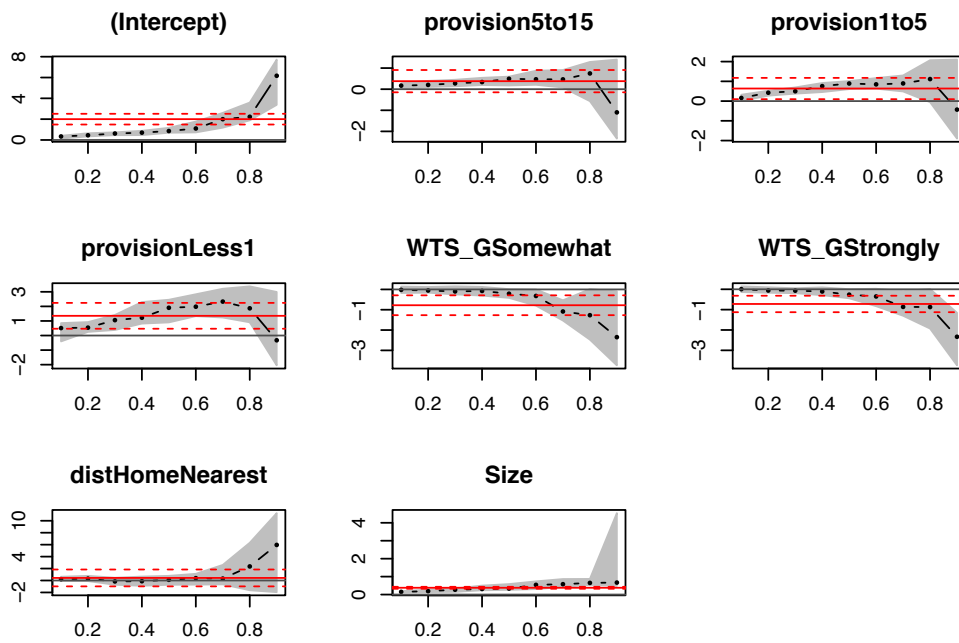
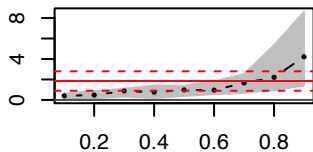
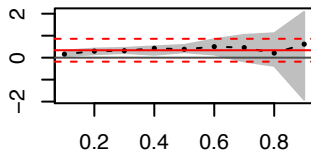
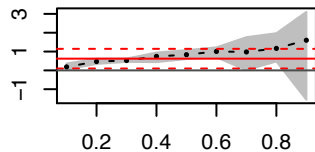
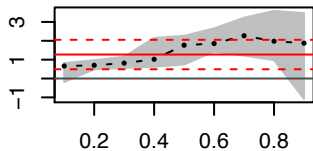
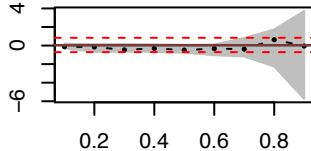
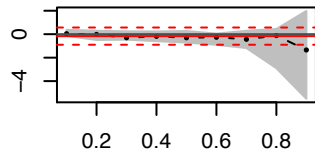
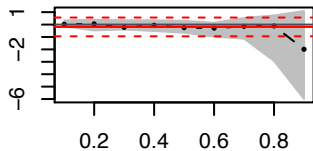
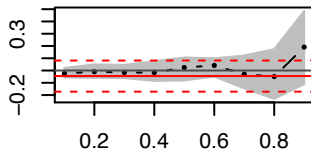
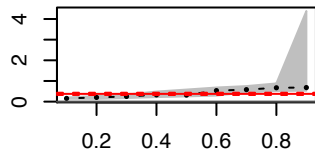
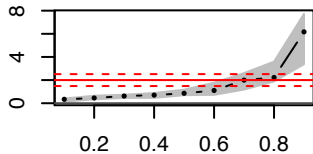
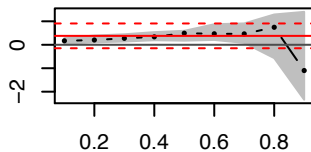
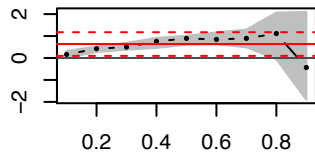
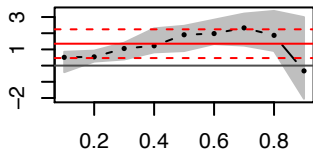
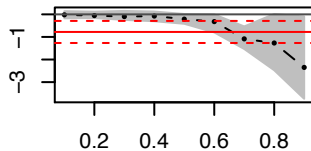
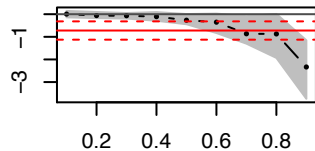
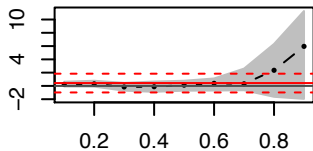
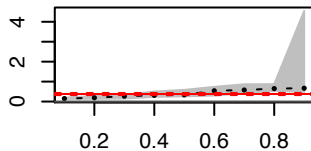
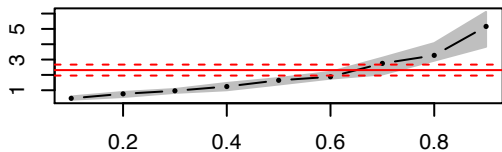


Figure C.3: Quantile regression analysis explaining the distance respondents travel to visit their most used UGS (network walking distance in km) with WTSG. UGS provision, willingness to substitute private for public UGS space, the distance to the nearest UGS and the size of the UGS are significant.

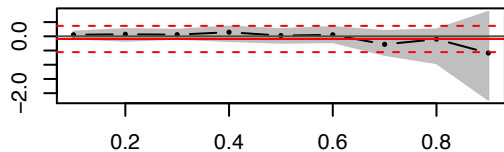
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**(Intercept)****provision5to15****provision1to5****provisionLess1****WTS\_GSomewhat****WTS\_GStrongly****distHomeNearest****Size**

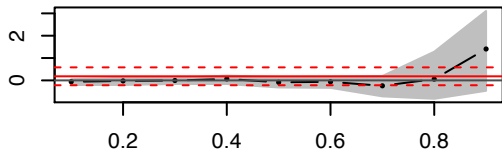
**(Intercept)**



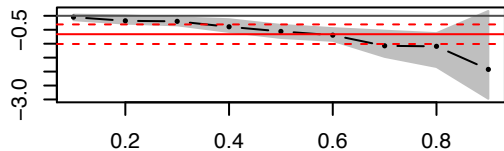
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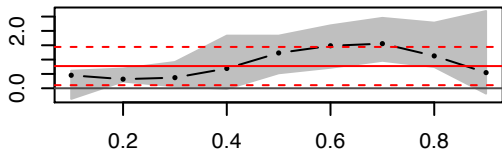
**WTS\_HStrongly**



**VeryGoodUGSquality**



**provisionLess1**



**Size**

