

1 Article

2 Assessing the Impact of Urban Greenery Captured by 3 Street View Imagery on Fear of Crime in Guangzhou, 4 China

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18 **Abstract:** Previous literature has examined the relationship between the amount of green space and
19 perceived safety in urban areas, but little is known about the effect of street-view neighborhood
20 greenery on fear of crime. Using a deep learning approach, we obtained the street-view greenery
21 data from massive street view images in central Guangzhou and tested the relationships and
22 mechanisms between street-view greenery and fear of neighborhood crime. Results demonstrated
23 that street-view greenery was associated with a lower fear of crime, and its relationship was
24 mediated by perceived physical incivilities and moderated by perceived social integration. This
25 study provides a solution for decreasing fear of crime by handling a microenvironmental
26 characteristic. Moreover, we concluded that both physical design and social factors should be
27 considered in attempts to understand the fear of crime and urban greenery planning.

28 **Keywords:** fear of crime; street-view greenery; neighborhood incivilities; social integration; China

31 1. Introduction

32 Many people are afraid of densely vegetated places. For instance, in the studies on urban parks,
33 densely wooded areas have consistently been related to fear [1,2]. However, well-maintained or high-
34 visibility vegetation is associated with a lower fear [3,4]. Recently, there is a rising interest in studying
35 the relationship between the amount of urban green space and perceived neighborhood safety [5-7].

36 The urban green space data based on satellite images are widely used by scholars [8,9]. Although
37 satellite images are continuous, timely, and informative, some small objects are difficult to detect due
38 to the medium spatial resolution. More importantly, the downward-facing satellite represents a bird-
39 eye view, and cannot accurately reflect people's perceived green space from the ground-level
40 perspective. In recent years, the Street View Index (SVI) at eye-level [10] has been widely used. Eye-
41 level street greenery can reflect more actual exposures to green space, including small-sized and
42 vertical natural elements (e.g. trees along a street, small plants, and grasses, etc.) [11]. Besides, most
43 street-view greenery represents highly visible vegetation, which might is a type of greenery that is
44 beneficial to safety perception.

45 As the most populous country, China has experienced unprecedented rapid urbanization in
46 recent decades, which reduces the opportunities for urban residents to be exposed to natural outdoor
47 environment [12]. Thus, clarifying the underlying linkage between residential street-view greenery
48 and fear of crime will not only fill the knowledge gap about the Chinese context but also provide
49 guidance for police departments, urban planners, and policymakers on how to improve
50 neighborhood safety perception via urban environment design programs.

51 More research is desired in this area. Using massive street view data, this study evaluates the
52 street-view greenery in Guangzhou neighborhoods via a deep learning technique and examines the
53 relationships and paths between street-view greenery and fear of crime in the Chinese setting under
54 the guidance of environmental criminology theories.

55 *1.1. Determinants of the fear of crime*

56 Fear of crime is a criminological concept and its definition is not uniform. Abundant
57 criminological studies suggest that fear of crime refers to resident's concerns and worries about
58 encountering criminal incidents [13-15]. The dominant theoretical perspective of fear of crime is the
59 vulnerability model. According to this model, those physically or socially vulnerable people are
60 expected to report higher levels of fear of crime. Women and the elderly are typically regarded as
61 physically vulnerable populations, meanwhile, those with a lower level of personal income and/or
62 education are considered as socially vulnerable populations. Research has demonstrated that the
63 groups of women, the elderly, the lower-income, and the less educated tend to report higher levels
64 of fear [16-18]. The second influential theory is the disorder model which directly focuses on
65 community characteristics. A neighborhood with physical deterioration (e.g. damaged public
66 facilities) reflects residents in this neighborhood are lack social control, thereby causing fear of crime
67 [19,20]. The third model, known as the social integration model, focuses on social integration, which
68 is defined as people's sense of belonging to their local community as well as their attachment to the
69 neighborhood [21]. The social integration model posits that the more socially integrated the person,
70 the lower their vulnerability and fear of crime [22].

71 These findings are critical and can provide policy recommendations and practical program
72 guidance. However, another type of urban characteristics has also attracted the attention of
73 researchers and policymakers, namely, the physical characteristics of urban neighborhoods
74 themselves.

75 *1.2. Urban greenery and fear of crime: evidence*

76 Traditionally, due to the closed space environment, green spaces contribute to higher levels of
77 fear in several studies [1,2,23]. Some research use images to test the relationship between green space
78 and fear in residential settings. A study on the safety ratings for vegetation photographs showed that
79 pictures with closed woodland space and natural shrubby edge space are unsafe for participates [1].
80 Similar results were obtained from an experiment using computer-based photo simulations [3]. Some
81 explored this relationship by questionnaires or surveys in the actual residential settings. Mak & Jim
82 [24] reported that people were afraid of becoming a victim of crime when visiting the parks in Hong
83 Kong. Meanwhile, several studies have found that people's fear of crime depends on the visibility of
84 green space. Woodland vegetation with good visual accessibility is beneficial to people's perception
85 of safety [3,4]. That is, closed green space (such as dense woodland and natural shrubby) are
86 associated with high fear levels, while open green space (such as grassland) is associated with low
87 fear levels. Moreover, literature has demonstrated that fear in green space is associated with people's
88 age, gender, household composition, ethnicity, socioeconomic status, and prior victimization [25-27].
89 Residents in Kuala Lumpur, Malaysia, had some defensive behaviors against crime in parks, but this
90 behavior was observed only among women [28].

91 A basic concept can easily be misunderstood. In many of the above studies, fear is a universal
92 feeling (not crime-specific fear), or a situational fear (it will disappear soon after leaving that
93 environment) [29]. Indeed, some research also involved the association between vegetation and
94 situational fear of crime and showed similar findings [28]. Regarding the stable fear of crime in the

95 neighborhood, Maas and her colleagues [5] first illustrated the relationship between the amount of
96 green space and perception of neighborhood safety, with a case in the Netherlands. Their findings
97 suggested that residents living in a greener surrounding environment reported increased feelings of
98 social safety. In the highly urbanized areas, women and older people had higher levels of social safety,
99 while men were generally not affected by green space. Then, Mouratidis [7] verified the relationship
100 between urban tree cover and increased perceived safety in 45 neighborhoods of the Oslo
101 metropolitan area.

102 Some circumstantial evidence also suggests a possible relationship between green space and fear
103 of neighborhood crime. Although crime and fear of crime are not similar and even not correlated, a
104 series of research has demonstrated that more residential vegetation can reduce crime rates [30-32].
105 Moreover, studies have examined the relationship between vegetation and lower levels of incivilities
106 [33,34]. For example, fewer problems are reported in neighborhoods with higher levels of maintained
107 gardening. Additional evidence from studies that examine the relationship between residential
108 vegetation and residents' levels of aggression, violence, physical and mental health [35]. The majority
109 have illustrated the positive effect of green space on residents' wellbeing and physical health [8,9].
110 The higher the feeling of wellbeing and the healthier the body, the less likely the person will become
111 a vulnerable individual, and thus might a lower fear of crime.

112 In sum, residential vegetation has been linked with a lower crime rate, fewer incivilities, and less
113 aggressive and violent behavior. There is a variety of evidence suggesting that vegetation may be
114 linked to a lower fear of crime in residential neighborhoods. But empirical evidence is still lacking in
115 the street-view greenery field, as well as in non-Western contexts.

116 1.3. Urban greenery and fear of Crime: Theory

117 Not merely on direct and indirect evidence, there are some theoretical signs that the greater
118 perception of safety is justified. In criminological research, surveillance is an inhibitor of criminal
119 activities. One of the earliest and most influential statements is Jacobs's *Death and Life of Great*
120 *American Cities* [36]. Jacobs [36] suggested that the presence of more 'eyes on the street' would deter
121 criminal activities. In urban areas, it has shown that outdoor spaces with more trees are consistently
122 greater used by residents than spaces without trees [37,38]. The more trees in an area, the greater the
123 number of ambient population. Offenders avoid areas with a greater likelihood of surveillance and
124 intervention because their activities might easily be observed in these areas. Thus, a greater amount
125 of vegetation might increase surveillance. Furthermore, Newman [39] proposed the 'defensible space'
126 and developed the concept of 'eyes on the street'. Newman [39] suggested that criminals might be
127 deterred by environmental signs even if no one is present. This may be a function of territorial
128 marking, as Kuo & Sullivan [40] proposed that well-maintained vegetation may constitute a
129 particular territorial marker/sign. Therefore, implied surveillance can also prevent crime, and
130 possibly the fear of crime.

131 *Moderation.* Increased street-view greenery might reduce people's fear of walking through their
132 neighborhood because of a higher level of neighborhood social control. If residents are in a socially
133 integrated community, it can provide an informal social fabric that will enforce local social control of
134 urban streets and might make the beneficial role of green space more significant. Thus, the usage of
135 the street is very critical, especially when it involves a certain degree of local social integration [41].
136 Jacobs [36] highlighted the need for design and social factors to ensure vibrant and diverse street
137 usage to enhance urban safety. Once the social and physical structure of the neighborhood's streets
138 is fully integrated, the harmful effects of strangers on the street will be eliminated and possibly begin
139 to provide help. For instance, in a report on the interaction effect of street traffic and social integration
140 on fear of crime [42], street traffic inhibited fear of crime when social integration was high.

141 *Mediation.* The relationship between the micro-environment and fear of crime is in two ways.
142 First, direct signs of crime can indicate the possibility of crime. Second, environmental clues might
143 arouse fear of crime, and the environmental clues symbolize the neighborhood's ability to exercise
144 informal social control. Collective efficacy and neighborhood incivilities may be driven by
145 neighborhood structural characteristics, such as neighborhood poverty level, ethnic diversity, and

146 crime rates. Brunton-Smith, Jackson & Sutherland [43] suggested that collective efficacy and
147 neighborhood incivilities are the bridges by which neighborhood structure characteristics affect fear
148 of violence. On one hand, urban street-view greenery might provide opportunities for neighborhood
149 interaction, and contribute to social integration and attachment [44]. Meanwhile, social integration is
150 a typical predictor of inhibiting fear of crime, so green space in neighborhoods might reduce the level
151 of fear among residents. On the other hand, a higher level of street-view greenery heralds a more
152 orderly and low-incivility neighborhood, thus might lead to a lower level of fear of crime.

153 Therefore, we propose that street-view greenery may deter fear of crime in the following ways:
154 by increasing informal and implied surveillance of the neighborhood, and by increasing the
155 probability of intervention. Moreover, perceived conditions (such as neighborhood incivilities and
156 social integration) might moderate or mediate the relationship between the street-view green space
157 and fear of crime.

158 1.4. The present study

159 The aim of this study to analyze the associations and mechanisms between street-view greenery
160 and fear of crime among residents, by using street-view greenery extracted from massive street-view
161 pictures in central Guangzhou, China. The following are the research questions:

- 162 1. Is street-view greenery associated with a lower fear of neighborhood crime?
- 163 2. What are the potential moderating and mediating roles of perceived conditions (physical and
164 social incivilities, and social integration) on the relationship between street-view greenery and fear
165 of crime?

166 2. Materials and Methods

167 2.1. Description of the study area

168 Guangzhou (112°57'E – 114°30'E; 22°26'N – 23°56'N) is located in southern China, adjacent to the
169 South China Sea. It is the capital and largest city of Guangdong province and the third-largest city in
170 China, with an estimated population of 15 million in 2019. Guangzhou contains 11 administrative
171 districts over an area of 7,434 km². As the area within the *S81 Guangzhou Ring Expressway* is the central
172 and most urbanized area of Guangzhou, and the street view data is fully covered, we choose this area
173 (central Guangzhou) as the sample area. This area covers an area of 380 km² (Figure 1a).

174 2.2. Data

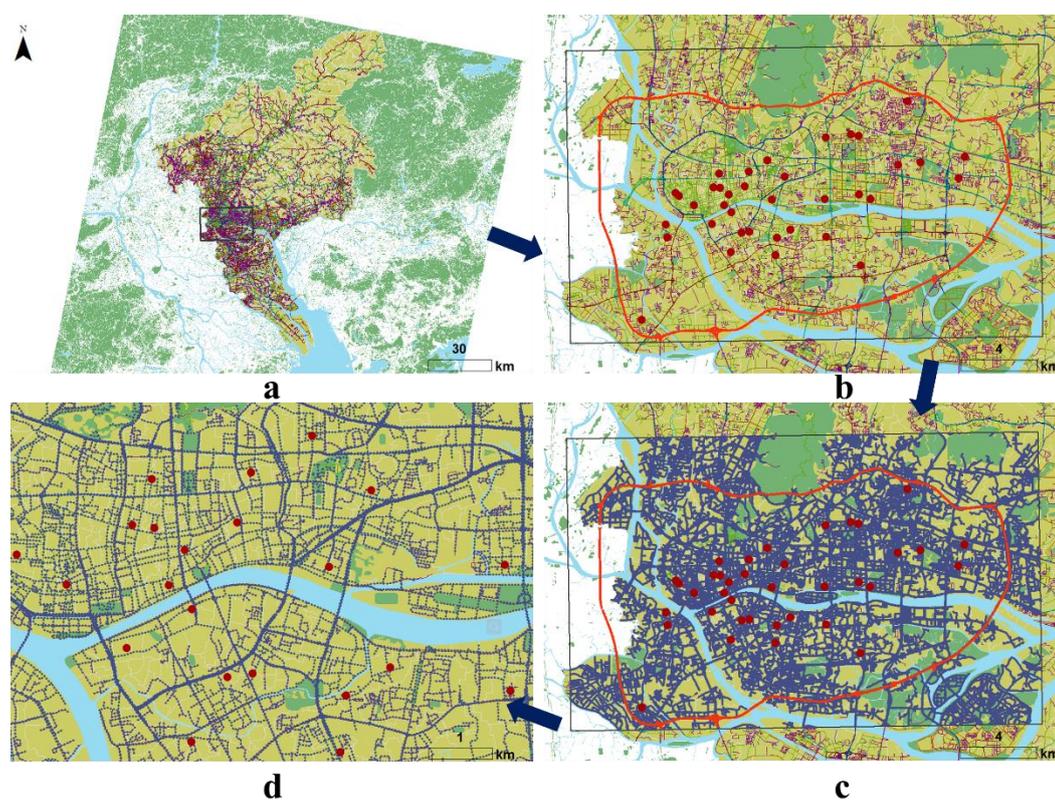
175 2.2.1 Fear of crime data

176 The current study employs data from the Project on Public Safety in Guangzhou Neighborhoods
177 (PPSGN), an interdisciplinary study aimed at understanding how neighborhood contexts affect the
178 safety perception of residents in Guangzhou, China. The survey was designed by the research team and
179 was distributed from January to April 2016 by the HOUSONWELL market research
180 (<http://www.hswell.com/>), a professional and well-known market research company in China. Since
181 college students and people under the age of 18 are greatly affected by the campus environment, they
182 were politely declined to participate in the questionnaires. Using a probability sample procedure, the
183 PPSGN collected information from 1,994 Guangzhou residents in 91 neighborhoods, with a response
184 rate of 87.4%. As the case area in this study was central Guangzhou, finally, 773 participants from 37
185 neighborhoods in the total sample were selected (Figure 1b).

186 2.2.2 Street-view greenery data

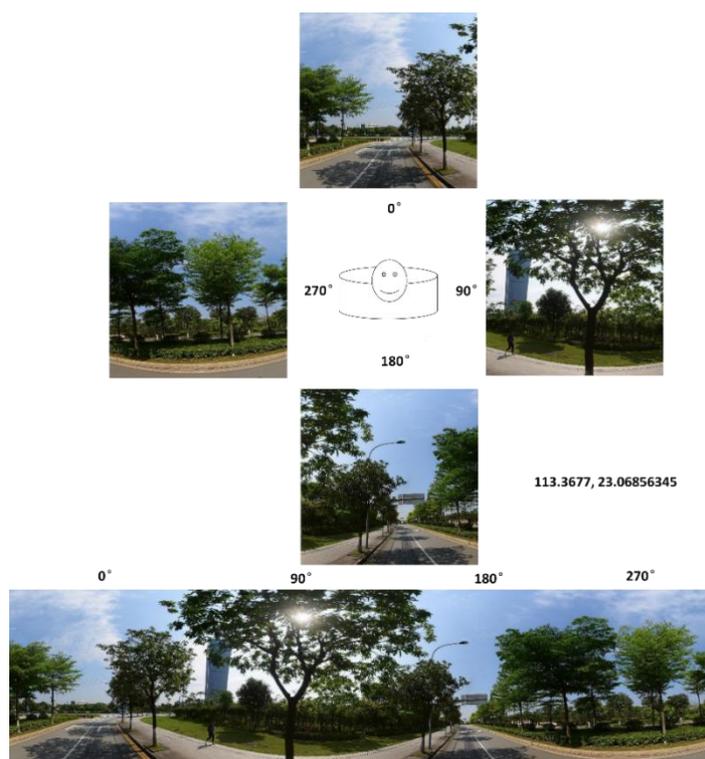
187 We evaluated green space per neighborhood based on a series of street view images collected in
188 December 2018. The images were crawled from Baidu Map, the Chinese equivalent of Google Maps
189 (Google products cannot be accessed in mainland China). It is China's leading map services provider,
190 has the largest image coverage, and providing street view pictures taken from various positions. Several

191 studies involving Chinese cities have used Baidu Map data [45,46]. Based on the Guangzhou road map,
192 we constructed point transects along with the road network (Figure 1 c & d). The sampled points are
193 50m apart, which is a compromise between the detail and the cost. Given these locations, the closest
194 pictures in the horizontal direction were queried through an URL link and crawled through the API
195 (application programming interface). To include the entire streetscape at each sampling point, we
196 acquired images taken in the four directions (i.e., 0, 90, 180, and 270 degrees) (Figure 2). The size of each
197 image is 512×512 pixels with a vertical angle of 0 degrees (pitch=0). In total, through the Python
198 programming language, 73,289 street view images with four directions were obtained.
199



200

201 **Figure 1.** Distribution of sampled points for capturing street greenery images. (a. Guangzhou map; b.
202 Central Guangzhou (red circles refer to sampled neighborhoods; red ring refers to S81 Guangzhou
203 Ring Expressway); c. purple points refer to street-view sample points; d. a magnified area with street-
204 view sample points.)



205

206

Figure 2. Street-view greenery images capturing example

207 2.3. Measures

208 2.3.1. Self-reported fear of crime indicator

209 *Fear of crime.* Consistent with previous work [47,48], fear of crime was assessed using a question
 210 with two items from the PPSGN survey: ‘how fearful are you to walk around the neighborhood after
 211 10 p.m. (a. fear of robbery; and b. fear of harassment)?’ Respondents were asked to indicate a response
 212 to the above question on a 5-point Likert-type scale, where 1 stands ‘not fearful’, to 5 stands ‘very
 213 fearful’. Higher values indicate a greater feeling of fear ($M = 2.34$, $SD = 1.03$, $\alpha = .79$).

214 2.3.2. Individual-level variables

215 *Demographic Information.* In general, female and old people might have high levels of fear, so gender
 216 (female=1) and age (continuous variable) were assessed. This study also measured personal income
 217 (ordinal variable, ranging from 1 to 7, meaning from low to high) and the education level (ordinal
 218 variable, ranging from 1 to 8, meaning from low to high). The measure of prior victimization was to ask
 219 whether the respondent had previously been victimized in the last three years (1 = yes).

220 *Perceived physical and social incivilities.* Respondents were asked about a series of conditions
 221 separately and whether it was present in their surroundings. Perceptions of neighborhood incivilities
 222 are composite variables (summed across several variables). The variable perceived physical incivilities
 223 reflects a scale ranging from 0 to 5, including the following four items: abandoned cars and/or trash,
 224 damaged public facilities and/or poor lighting, graffiti and/or disordered advertisements, and noisy
 225 neighborhood environment ($\alpha = .82$). The variable perceived social incivilities reflects an increasing scale
 226 ranging from 0 to 5, including the following four items: drunken persons on the streets, teenagers
 227 gathering on the streets, suspicious strangers, and residents conflicting on the streets ($\alpha = .84$). In the total
 228 sample, the mean is 2.82 and 1.85 for physical incivilities and social incivilities, respectively.

229 *Social integration.* Respondents were asked four items about their perception of social integration
 230 in their neighborhood. It reflects a scale ranging from 0 to 5, comprised of visiting informally with

231 neighbors, chatting with neighbors, borrowing things like tools from neighbors, and belonging to a
232 network in which neighbors help each other ($M = 3.42$, $SD = .83$, $\alpha = .77$).

233 2.3.3. Neighborhood-level variables

234 *Neighborhood street-view greenery.* Using a buffer around the neighborhood to define the
235 neighborhood boundary can reflect more realistic walkable distances, which has been widely adopted
236 by prior studies [49,50]. Thus, we measured .5 km, 1 km, and 1.5 km buffer area of each neighborhood
237 by ArcGIS 10.3 software. Then, average street-view greenery within a .5 km ($M = .27$, $SD = .07$), a 1.0 km
238 ($M = .25$, $SD = .05$), and a 1.5 km ($M = .25$, $SD = .04$) radius around each neighborhood centroid were
239 calculated. The street-view greenery within a 1.0 km radius is for main effect analysis, while the street-
240 view greenery within the radius of .5 km and 1.5 km is for sensitivity analyses to ensure the reliability
241 of the results.

242 When investigating the relationship between fear of crime and the visibility of green space in a
243 person's living environment, other neighborhood-level factors that could influence fear of crime ought
244 to be controlled. People tend to feel less safe in neighborhoods with higher concentrated disadvantages
245 [51]. Concentrated disadvantages are related to the percentage of immigrants, neighborhood income
246 level, and crime rates [52], which leads us to include the percentage of population with a bachelor's
247 degree or higher¹, the percentage of migrant population, and the robbery rate in the neighborhood.

248 *Proportion of population with a bachelor's degree or higher.* In a neighborhood, percentage of population
249 with a bachelor's degree or higher was calculated as the ratio between the number of people with a
250 bachelor's degree or higher and the number of neighborhood inhabitants ($M = .19$, $SD = .22$). Relevant
251 population data were obtained from the 2010 Census².

252 *Proportion of migrant population.* In a neighborhood, migrant population percentage was calculated
253 as the ratio between the number of non-local hukou population (migrants from outside Guangzhou)
254 and the number of neighborhood inhabitants (from 2010 Census) ($M = .17$, $SD = .13$).

255 *Robbery rate.* Robbery rate (per 100,000 of the population) was calculated for each of the 37
256 neighborhoods in the sample by averaging official police counts of robbery for 2014 and 2015, dividing
257 by the total neighborhood population (from 2010 Census), and multiplying by 100,000. Robbery crime
258 data was obtained from Guangzhou Public Security Bureau ($M = 53.68$, $SD = 81.30$).

259 2.4. Deep learning for image segmentation

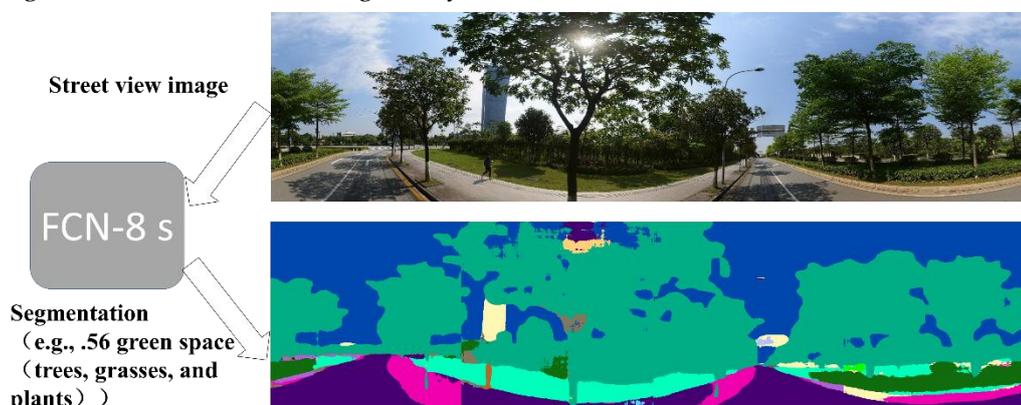
260 Yang et al. [10] proposed a Green View Index (GVI) to evaluate the visibility of urban greenery,
261 which was defined as the ratio of the total green area of four pictures taken at a street intersection to the
262 total area of the four pictures. Then, Li et al. [53] assessed the street-view greenery by pixel-wise
263 classification based on massive google street view imagery. However, there are some limitations in
264 pixel-wise classification using additive colors of the picture, such as the indistinguishable natural and
265 artificial green objects. To overcome the limitation, the present study implemented a deep learning
266 approach to extract street-view greenery from massive street images.

267 We applied a semantic segmentation technique called FCN-8s (full convolutional network for
268 semantic segmentation) to identify green space from street view image data [11,54]. The FCN-8s was
269 trained based on the ADE_20K dataset. This approach can identify 150 elements, including common
270 ground objects such as trees, sky, persons, cars, roads, and buildings. The trained FCN-8s displayed
271 good performance for street image scene segmentation, and a series of research has used the trained
272 FCN-8s to assess streetscapes in Chinese cities with desirable results [11,50]. For the technical details of
273 the trained FCN-8s model, see the work of Yao and his colleagues [55]. After the image segmentation,
274 the proportion of green space (including trees, grass, and plants) was calculated (Figure 3).

¹ There is no neighborhood-level income data in China, so we choose neighborhood education level as a substitute variable, because the personal income and the education level are usually considered to be highly related.

² The 2010 Census refers to the 2010 National Census, which is the latest census data in China so far, with various attribute data of the smallest administrative unit.

275 In each sampling point, streetscape green space represents the ratio of the number of green space
 276 pixels per image summed over the four directions to the total number of pixels per image summed over
 277 the four directions. Then, the data of points were aggregated to the neighborhood by average method,
 278 and the neighborhood-level street-view greenery was obtained.



279

280

Figure 3. Street view image segmentation by a fully convolutional network (FCN-8s).

281 2.5. Analytical Strategy

282 As the dataset shows a nested structure, this study evaluated the associations between street-
 283 view greenness and fear of crime through multilevel regression modeling, with controlling several
 284 individual and neighborhood variables. Numerous studies on fear of crime have illustrated the
 285 satisfactory performance for multilevel modeling [5,20]. The multilevel analysis was performed with
 286 HLM 6.08 [56] in this present study. Here, the models were expressed as follows:

$$287 \text{ Fear of crime} = \beta_0 + \sum_q \beta_q X_{qi} + r_i \quad (1)$$

$$288 \beta_0 = \gamma_{00} + \sum_s \gamma_s W_s + \mu_0$$

$$289 \beta_q(\text{Mean } X_{qi}) = \gamma_{q0} + \mu_q$$

290 In the equation, β_0 was the intercept; β_q was the slope of the factor q on the fear of crime; X_{qi}
 291 was the factor q for individual i ; and r_i was the error term that was represented to the random effects.
 292 Where β_0 was modeled as a function of neighborhood contextual characteristics W_s (residential
 293 surrounding street-view greenery, etc.); W_s was the factor s for neighborhood j ; the models were
 294 specified with a random error coefficient μ_q . This allowed the intercept to vary as a function of
 295 neighborhood features and any unique neighborhood effects.

296 First of all, the present study estimated a one-way analysis of variance with random effects, to
 297 explore how much influence was produced by the neighborhood-level factors. Consistent with
 298 Wyant [57], the intraclass correlation coefficient indicated that 15.19% variation of the fear of crime
 299 can be explained by neighborhood-level variables, and the rest was attributed to individual-level
 300 change and random error. Then, separate analyses were conducted for surrounding green space
 301 within a .5 km, a 1.0 km, and a 1.5 km radius. In a multilevel regression model, a variable with no
 302 meaningful zero value needs to be centered. Therefore, all individual-level variables were mean
 303 grouped, and all neighborhood-level variables were grand mean centered in this current study.

304 2.5.1. Moderator analyses

305 Moderation analyses, which were performed to examine the moderation effects of perceived
 306 conditions on fear of crime, were conducted using HLM models. Before the analyses, all variables
 307 were centered to avoid potential multicollinearity problems [58]. At the first step, after the operation

308 of the level-1 model, the error terms of perceived perceptions (perceived physical and social disorder,
 309 and social integration) were checked. If the error term of the perceived condition variable does not
 310 vary significantly in the equation ($p > .05$), it indicates that the effect of the variable on fear of crime
 311 is mainly explained by individual-level variables, and the role of the neighborhood environment is
 312 not important. If the variance component of the variable significantly varies across neighborhood
 313 context, a series of neighborhood variables would be entered to the intercept to detect the variance
 314 of the effect of the perceived condition, exploring the function of the neighborhood-level street-view
 315 greenery after controlling for the effects of other neighborhood variables [52,59,60]. The equations are
 316 below:

$$317 \quad \text{Fear of crime} = \beta_0 + \sum_q \beta_q X_{qi} + r_i \quad (2)$$

$$318 \quad \beta_0 = \gamma_{00} + \sum_s \gamma_s W_s + \mu_0$$

$$319 \quad \beta_q (\text{Mean } X_{qi}) = \gamma_{q0} + \sum_s \gamma_s W_s + \mu_q$$

320 2.5.2. Mediated analyses

321 Mediation analysis was undertaken in four steps following Baron & Kenny [61] and previous
 322 research [62]. Conditions for mediation are that the predictor variable (street-view greenery) must
 323 influence the mediator; and that the mediator must influence the outcome variable (fear of crime);
 324 and that the association between the predictor and outcome is eliminated or weakened when the
 325 mediator is included in the model. The mediating role of the perceived condition on the relationship
 326 between street-view greenery and fear of crime was assessed by conducting three separate statistical
 327 models. First, the correlations were calculated between the possible mediators (physical and social
 328 incivilities, and social integration) and fear of crime. Only for combinations with statistically
 329 significant correlations are possible mediation considered. In these cases, the second step was to
 330 compute a linear regression where the possible mediator was entered as the dependent variable and
 331 street-view greenery was taken as the independent variable. The last step was to perform a multiple
 332 regression using a saturation model, in which both the mediator and street-view greenery were
 333 entered as independent variables, and the fear of crime was input as the dependent variable. 5,000
 334 bootstrap samples were used to examine the statistical significance of the mediating variable. This
 335 method achieves 95% confidence intervals (CI) of the indirect effect. A bootstrapped 95% CI not
 336 straddling 0 is considered statistically significant [63]. Mediation analysis was conducted using the
 337 PROCESS v3.5 for SPSS Statistics 22 (Model 4, Hayes [63]).
 338

339 3. Results

340 3.1. Descriptive statistics

341 Among the finally chosen 73,289 image sites in central Guangzhou, street-view greenery scores
 342 vary from 0 to .86, with a mean value of .234. The histogram distribution of street-view greenness for
 343 all image sites in Guangzhou is shown in Figure 4c. Compared with the vegetation greenness by
 344 satellite imagery (NDVI, the normalized difference vegetation index), the visualization of greenness
 345 in the street shows similar but different spatial characteristics (Figure 4 a & b). Converged to a planar
 346 spatial scale, street-view greenery of neighborhood under different radii show similar values,
 347 with .27, .25, and .25 respectively.

348 The descriptive statistics of participants' individual and neighborhood characteristics were
 349 presented in Table 1. The sample was composed of 49.1% women, with an average age of 40.42 years.
 350 The average fear of participants was not high (Mean = 2.34). We also took the correlation analysis for
 351 the individual and neighborhood characteristics, separately. No problems with collinearity were
 352 detected.

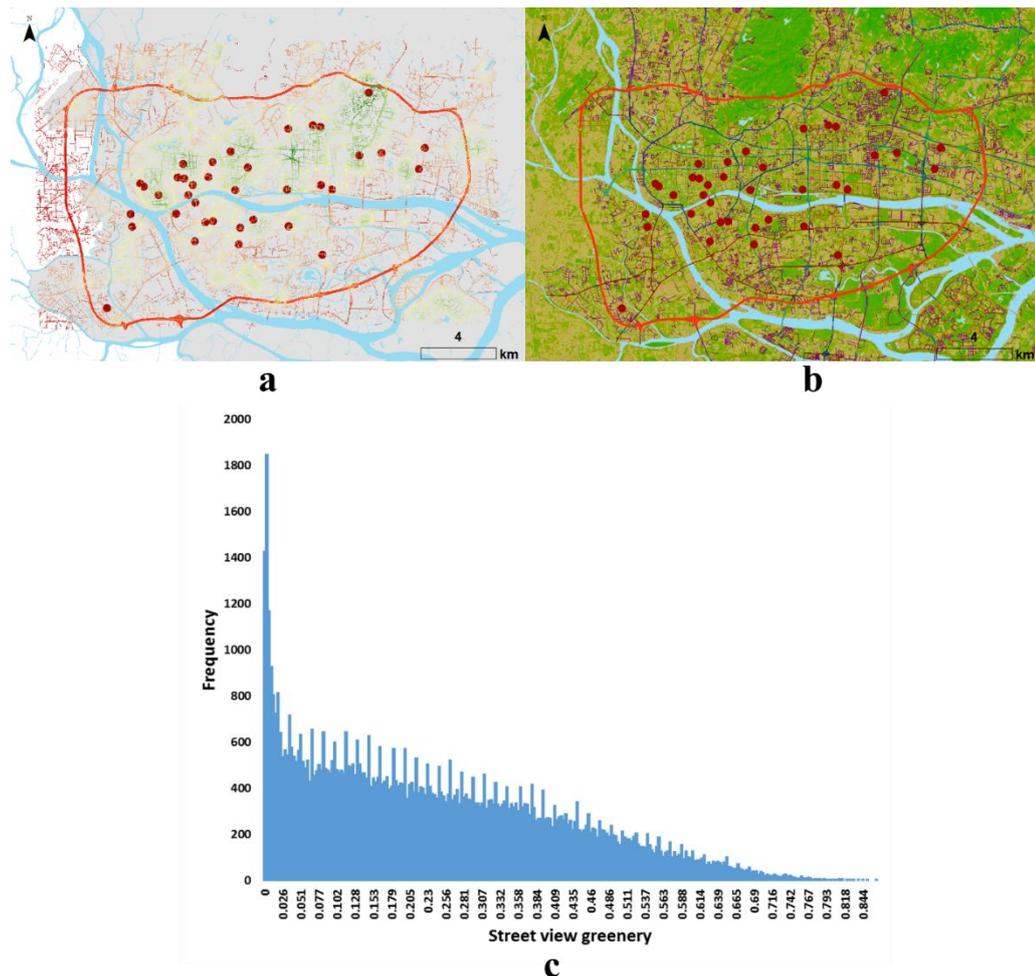
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Table 1. The characteristics of the study variables

| | Mean | SD | Min | Max | | % (N) |
|---|-------|-------|-----|--------|-------------------|-------------|
| Individual level | | | | | | |
| Age | 40.21 | 14.68 | 19 | 84 | | |
| Gender (female=1) | .49 | .50 | 0 | 1 | Female | 49.3% (381) |
| | | | | | Male | 50.7% (392) |
| Level of education | 4.46 | 1.64 | 1 | 8 | | |
| Personal income | 3.67 | 1.24 | 1 | 7 | | |
| Prior victimization experience (yes=1) | .35 | .48 | 0 | 1 | Victimization | 35.2% (272) |
| | | | | | Non-victimization | 64.8% (501) |
| Perceived physical incivilities | 2.82 | .83 | 1 | 5 | | |
| Perceived social incivilities | 1.85 | .76 | 1 | 5 | | |
| Perceived social integration | 3.42 | .83 | 1 | 5 | | |
| Neighborhood level | | | | | | |
| Street-view greenery (0.5 km buffer) | .27 | .07 | .07 | .47 | | |
| Street-view greenery (1.0 km buffer) | .25 | .05 | .10 | .39 | | |
| Street-view greenery (1.5 km buffer) | .25 | .04 | .14 | .36 | | |
| Percentage of population with a bachelor's degree or higher | .19 | .22 | .01 | .86 | | |
| Percentage of migrant population | .17 | .13 | .02 | .65 | | |
| Robbery rate (per 100,000 people) | 53.68 | 81.30 | .00 | 304.16 | | |
| Dependent variable | | | | | | |
| Fear of crime | 2.34 | 1.03 | 1 | 5 | | |

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357

358 **Figure 4.** Descriptive statistics (a. visualization of street-view greenery in the street; b. vegetation
 359 greenness by satellite imagery on February 7, 2016; c. histogram distribution of street-view greenery.)

360 3.2. Main effects of street-view greenery

361 In line with expectations, after controlling relevant variables, street-view greenery within 1.0 km
 362 was significantly related to fear of crime (coeff. = -3.25, $p < .05$, Model 1, Table 2), thus, respondents
 363 exposed to more street-view greenery in the living environment reported lower levels of fear than
 364 respondents exposed to less street-view greenery in the living environment.

365 To evaluate the robustness of our findings to our selection of 1.0 km buffer size, we repeated the
 366 aforementioned main analyses using street-view greenery in buffers of .5 km and 1.5 km. Sensitivity
 367 analyses showed consistent results for different buffer sizes. Surrounding street-view greenery
 368 within .5 km was associated with lower levels of fear of crime, while street-view greenery within 1.5
 369 km was not statistically significantly associated with fear of crime. That is, higher street-view
 370 greenery in the distant environment did not impact people's fear of crime in their neighborhood. This
 371 is in line with the assumption that cumulative opportunity metrics.

372 Some of the controls were statistically significant. Women were more fearful than men, while
 373 the older reported a lower fear level than young people. Perceived physical and social incivilities
 374 were positively correlated with fear of crime. Respondents who perceived higher physical or social
 375 incivilities in the neighborhood had a higher fear level. Perceived social integration was associated
 376 with a lower fear of crime. At the individual level, we found no evidence that levels of education,
 377 personal income, and prior victimization were significantly related to fear of crime. At the
 378 neighborhood level, neighborhood migrant population is statistically associated with a lower fear of
 379 crime.

380
381**Table 2.** Multilevel results of the effect of street-view greenery on fear of crime (with robust standard errors)

| | Model 1 | | | Model 2 | | |
|---|---------|----------|----------|---------|----------|----------|
| | Coeff. | SE | T | Coeff. | SE | T |
| Intercept | 2.22** | .08 | 29.47 | 2.22** | .08 | 29.50 |
| <i>Level 1 – Between individuals</i> | | | | | | |
| Age | -.01* | .00 | -2.30 | -.01* | .00 | -2.27 |
| Gender (female=1) | .38** | .08 | 4.87 | .37** | .08 | 4.59 |
| Level of education | .01 | .03 | .23 | .01 | .03 | .36 |
| Personal income | .03 | .03 | 1.00 | .03 | .03 | .92 |
| Prior victimization experience (yes=1) | .10 | .11 | .93 | .09 | .11 | .89 |
| Perceived physical incivilities | .14** | .05 | 2.82 | .13* | .05 | 2.64 |
| Perceived social incivilities | .26** | .08 | 3.37 | .26** | .08 | 3.38 |
| Perceived social integration | -.10* | .04 | -2.48 | -.09* | .04 | -2.27 |
| <i>Level 2 – Between communities</i> | | | | | | |
| Street-view greenery (SVG) (1.0 km buffer) | -3.25* | 1.59 | -2.05 | -3.51* | 1.63 | -2.16 |
| Percentage of population with a bachelor's degree or higher | -.31 | .34 | -.92 | -.37 | .33 | -1.10 |
| Percentage of migrant population | -1.33* | .66 | -2.05 | -1.31* | .69 | -1.89 |
| Robbery rate (per 100,000 people) | .00 | .00 | 1.07 | .00 | .00 | .88 |
| <i>Cross-level Interaction</i> | | | | | | |
| Social integration × SVG | | | | -1.42* | .78 | -1.82 |
| Percentage of population with a bachelor's degree or higher × SVG | | | | .31 | .26 | 1.18 |
| Percentage of migrant population × SVG | | | | -.02 | .34 | -.07 |
| Robbery rate × SVG | | | | -.00 | .00 | -.63 |
| Variance components | Var. | χ^2 | <i>P</i> | Var. | χ^2 | <i>P</i> |
| Between individuals | .92 | | | .93 | | |
| Between neighborhoods β_0 | .20 | 176.85 | .00 | .20 | 176.12 | .00 |
| Slope of the perceived social integration | .01 | 28.75 | .02 | .01 | 27.64 | >.5 |

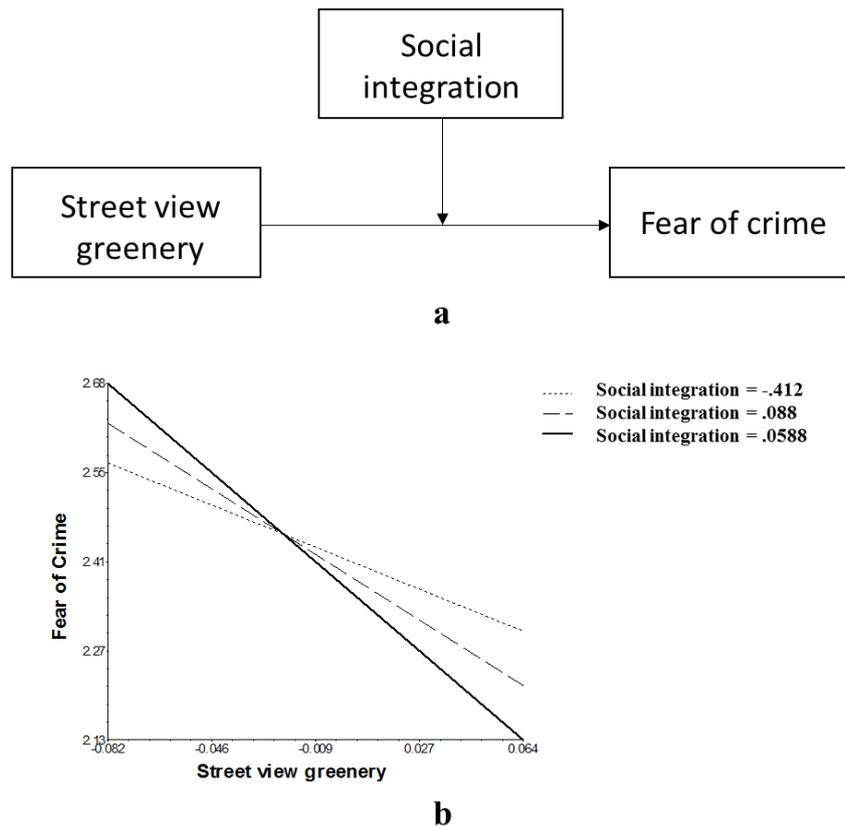
382 Note: SVG = street-view greenery (1.0 km buffer). * $p < .05$, ** $p < .01$.

383

384 *3.3. Moderator analyses*

385 After the operation of the level-1 model, only the error terms of social integration varied
386 significantly in the equations ($p < .05$), which indicated that the effect of social integration on fear of
387 crime was partly explained by neighborhood-level environment. Then, a series of neighborhood
388 variables were entered to the intercept to detect whether the effect of social integration on fear of
389 crime was influenced by neighborhood street-view greenery within 1 km. Analyses (Model 2, Table
390 2) revealed that fear of crime was significantly predicted by the interaction of street-view greenery ×
391 informal social integration (coeff. = -1.42, $p < .05$). The results demonstrated that informal social
392 integration magnified the beneficial effect of street-view greenery on reduced fear of crime (see Figure
393 5a for the theoretical framework). As shown in Figure 5b, when social integration is higher, the
394 streetscape greenery has a stronger inhibitory effect on fear of crime, and vice versa.

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Figure 5. The results of moderator analyses (a. the theoretical framework of the moderating effect. b. moderating effect of social integration and street-view greenery on fear of crime (negative values appeared because of variables centered. Graphed by HLM software).)

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3.4. Mediated analyses

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We conducted a series of mediation effect analysis to test whether perceived conditions (physical and social incivilities, social integration) mediate the relationship between street-view greenery within 1 km and fear of crime. A mediation analysis PROCESS v3.5 for SPSS was performed.

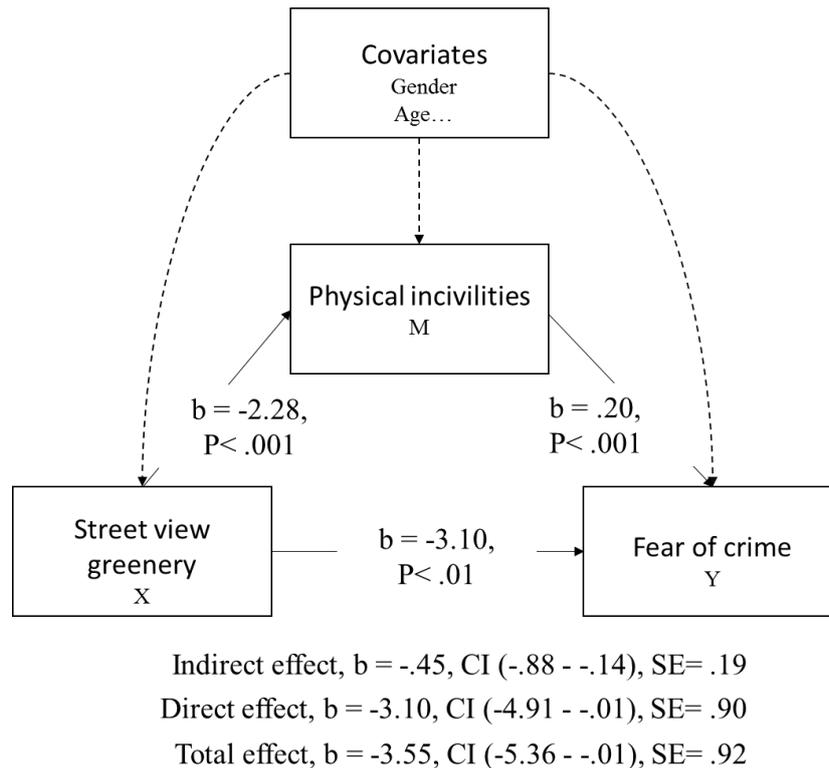
404

Demographic and socioeconomic factors were included as covariates for models presented in this study, as they showed associations with certain variables of the present study. The mediation results of physical incivilities showed (see Figure 6) that: (1) street-view greenery predicted perceived physical incivilities ($b = -2.28$, $p < .001$); (2) physical incivilities predicted fear of crime ($b = .20$, $p < .001$); and (3) street-view greenery predicted fear of crime (the direct effect, $b = -3.10$, $p < .01$). It was also reported that street-view greenery as a predictor of fear of crime was mediated by physical incivilities (the indirect effect, $b = -.45$, CI (confidence interval) $(-.88$ to $-.14)$, $SE = .19$). The mediation test was performed by calculating CI (confidence interval) for the 'indirect effect' using bootstrap methods. For the size of the effect, all CI (confidence interval) did not contain zero point estimates, thus, 0 can be excluded as a probable value for the direct and indirect effects.

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When other perceived conditions (perceived social incivilities and social integration) were included as the mediator for models separately, the mediation effect was not significant. Thus, no empirical evidence demonstrates the effect of street-view greenery on fear of crime via social incivilities or social integration.

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Figure 6. Mediation model of street-view greenery as a predictor of fear of crime, mediated by physical incivilities (b is an unstandardized regression coefficient, and CI is the confidence interval for the bootstrap approaches between BootLLCI and BootULCI).

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4. Discussion and Conclusions

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This study is the first to examine the linkage between fear of crime and exposure to green space at the street level among residents in China. Although remote sensing-based metrics of green space are widespread in criminological studies, we take an alternative avenue to assess green space, relying on deep learning technique and street view data. The results lead us to conclude that street-view greenery in people's living environment is generally associated with a lower fear of neighborhood crime. This relationship is concurrent with the positive relationship between green space and people's health and previous research. Enhancing urban green space not only makes people healthier but also helps to make them feel safer. The moderation and mediation analyses show that social integration magnifies the beneficial effect of street-view greenery on fear of crime, and perceived physical incivilities partially mediates the relationship between street-view greenery and fear of crime. It makes a further understanding of the relationship between green space and fear of crime, and guides that the safety improvement programs need to consider both the physical design and social factors.

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4.1. Contributions to the understanding of green space and fear of crime

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One contribution of this work is to further illustrate an environmental criminological perspective, that is, surrounding urban green space is associated with lower fear levels. Greener streets receive greater use, thereby increasing informal surveillance. Also, many forms of street vegetation preserve visibility and therefore it does not promote fear of crime in the neighborhood. This finding suggests that fear of crime prevention concerns do not justify removing street vegetation in highly urbanized areas.

Although this is the first study to demonstrate such a link, the findings are consistent with previous work linking vegetation with lower levels of incivilities [33], perceived social safety [44], as well as previous work linking vegetation with lower levels of crime rates [40]. Consistent with the findings of Maas et al. [5] and Mouratidis [7], we found that the positive relationship with safety

447 perception is linked to green space. The results obtained here are based on street-view greenery and
448 neighborhood perception of safety, whereas previous literature is based on satellite imagery and
449 social safety perception. The convergence of findings from such different measures lends confidence
450 that in urban residential settings, the relationship between street-view vegetation and fear of crime
451 in the neighborhood is negative. The higher the percentage of street-view greenery index, the less
452 fear of crime.

453 Based on the deep learning approach, some scholars [53,64] collected perceived safety data and
454 did some research on the relationship between perceived safety and environmental characteristics.
455 The principle of street-view safety perception scores is to show participants two geotagged images
456 and ask them to choose the image that looks safer. Then, any picture can get a safety perception score
457 by the trained model based on the crowdsourced data. Li, Zhang, and Li [53] found that the visibility
458 of green vegetation plays an important role in enhancing perceived streetscape safety. Different from
459 the perceived safety by deep learning, in our study, fear of crime refers to the respondents' perception
460 of safety in the neighborhood where they live. Even so, the work of Li and his colleagues indicated
461 that green space contributes to enhanced safety perception of images. It provides support for
462 explaining the impact mechanism of street green space on residential fear of crime.

463 Another contribution of the work here was to help resolve a previous puzzle on residential
464 vegetation and feelings of safety. It is generally believed that objective or perceived green space is
465 related to fear of crime because densely-vegetated urban neighborhoods represent lower levels of
466 poverty and incivilities/disorder. This makes people mistakenly believe that green space itself is not
467 associated with safety perception. The finding here shows that street-view greenery is systematically
468 and independently associated with a lower fear of neighborhood crime, after controlling other
469 individual and neighborhood characteristics. Moreover, this study demonstrates that neighborhood
470 physical incivilities/disorder partially but not completely mediates the relationship between street-
471 view greenery and fear of crime. Thus, in essence, street-view greenery has direct and indirect effects
472 on fear of crime.

473 When considering the interaction between social integration and street-view greenery, it shows
474 that in the high street-view environment, people with high social integration are likely to have a
475 higher safety perception compared with people with low social integration. Thus, the effect of street-
476 view greenery on fear of crime is related to social factors of the neighborhood. However, mediation
477 analyses did not reveal that social integration mediates the relationship between street-view greenery
478 and fear of crime. Based on our results, which display a stronger case for social integration act as a
479 moderator than a mediator. In fact, social integration is a frequently mentioned moderator, which
480 moderated the relationship between street traffic and fear of crime [42] and the relationship between
481 perceived neighborhood disorder and fear of crime [65]. This finding demonstrates that the effect of
482 street-view greenery on fear of neighborhood crime might be linked to both individual and
483 neighborhood characteristics.

484 4.2. *Practice implications and further research directions*

485 As street vegetation does not necessarily promote fear of crime and may even inhibit fear in
486 urban neighborhoods, it seems appropriate to increase street greenery in crowded inner-city areas to
487 create safer neighborhoods for residents. Further, a series of street greenery strategies are
488 recommended. High-visibility trees and grassland should be planted more, because of providing
489 good visibility. Then, green space strategies should continuously maintain the street green space to
490 avoid the negative impact of green space on fear of crime. Finally, since green space can be used as a
491 sign of territory, so planting green space in main road nodes can promote a greater beneficial effect.

492 The findings also have implications in other aspects. Research results show that fear of crime is
493 simultaneously influenced by physical environment and social factors (moderation and mediation
494 effects). Ultimately, the largest reductions in fear of crime will come from strategies that address the
495 social and cultural factors, such as reducing intense poverty, improving the level of collective efficacy,
496 and reducing the level of incivilities in the neighborhood. These strategies can help play a positive
497 role in physical design such as planting street greenery on the safety perception.

498 Our findings set the stage for further exploration of the relationship between urban street
499 vegetation and fear of crime. With the development of deep learning technique, the results of image
500 segmentation are constantly enriched. Future studies might use a deep learning approach to detect
501 the maintenance level and specific characteristics of vegetation (e.g. height, living vegetation volume,
502 and vegetation health) in the street scenes in large quantities, and examine the fear of neighborhood
503 crime associated with these factors. More detailed street-view greenery can examine more
504 sophisticated relationships.

505 Street view images and machine learning are powerful tools for environmental exposure
506 assessments in urban landscapes. As companies such as Baidu Map and Google Map provide geo-
507 referenced and publicly available street view image databases with broad spatial coverage, the
508 development of novel location-based environmental exposure measures (e.g. sky view index, street-
509 view water space, etc.) to explore environment–crime relations is also meaningful.

510 Our moderator and mediator models partially explained the mechanism of green space – fear of
511 crime linkage. Future research is needed to examine more potential moderators and mediators, such
512 as perceived neighborhood attachment, informal social control, and physical activity. Also, the effect
513 of green space on fear of crime might vary across types of neighborhoods. The effect of green space
514 in high-income areas and low-income areas might be inconsistent or even opposite. The next step
515 should compare the effect of street green space on fear of crime in different types of neighborhoods.
516 Finally, even if street green space reduces the fear of crime, is this relationship linear or curvilinear?
517 Living in an area with very high green space exposure, the benefits of greenery may be reduced
518 because of the desensitization effect. Future research might also focus on the possible nonlinear
519 relationship between green space and fear of crime.

520 4.3. Limitation

521 The present research has a limited ability to make causal inferences regarding our evaluated
522 associations. The first limitation is the temporal and spatial resolution of street view data. We did not
523 obtain the street view during the same period of the questionnaire survey because of the technological
524 restriction of Baidu street view rules. Street view is not updated every year, which does not accurately
525 reflect the changes in the street environment. Besides, there are some small roads or alleys that people
526 often walk, but street view cars cannot enter to capture images. The spatial distribution of street view
527 is not even in some areas. Secondly, our data do not provide any information on the actual use of
528 green space by the respondents, such as the frequency of visiting green space and time during the
529 green space. It limits to reveal the relationships between objective street green space, perceived street
530 green space, and fear of crime. Although several limitations need yet to be overcome, the present
531 study demonstrates the relationships and mechanisms of a novel street-view greenery and fear of
532 crime.

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534 Fengrui Jing, Suhong Zhou and Jiangyu Song; Formal analysis, Fengrui Jing; Funding acquisition, Lin Liu;
535 Investigation, Lin Liu; Methodology, Fengrui Jing, Lin Liu, Jiangyu Song, Linsen Wang, Hanlin Zhou, Yiwen
536 Wang and Ruofei Ma; Project administration, Lin Liu and Suhong Zhou; Resources, Suhong Zhou, Jiangyu Song
537 and Yiwen Wang; Software, Fengrui Jing, Linsen Wang, Hanlin Zhou, Yiwen Wang and Ruofei Ma; Supervision,
538 Lin Liu and Suhong Zhou; Visualization, Fengrui Jing and Linsen Wang; Writing – original draft, Fengrui Jing;
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