

Effects of urban green space on human cognition: A systematic search and scoping review

Shengjie Liu^a, Hung Chak Ho^{b,c}, John P. Wilson^{a,d}

^a*Spatial Sciences Institute, Dornsife College of Letters, Arts and Sciences, University of Southern California, Los Angeles, CA 90089, USA*

^b*Department of Public and International Affairs, City University of Hong Kong, Kowloon, Hong Kong*

^c*Social Determinants of Health Initiative, City University of Hong Kong, Kowloon, Hong Kong*

^d*Department of Population and Public Health Sciences, Keck School of Medicine, Departments of Civil & Environmental Engineering and Computer Science, Viterbi School of Engineering, Department of Sociology, Dornsife College of Letters, Arts and Sciences, and the School of Architecture, University of Southern California, Los Angeles, CA 90089, USA*

Abstract

Urban green space is associated with cognitive functions, but the underlying mechanisms remain unclear due to limited research. Given the diverse forms of green space, which lead to distinct health effects, it is essential to differentiate between types of green space. In this review, we propose a novel conceptual framework categorizing three primary effects of green space on cognitive function: functional, spatial, and perceptual. We then conduct a scoping review using the Web of Science, identifying 37 relevant studies. Among them, 20 studies employ modeling to explore potential mechanisms, while 17 studies infer pathways indirectly. Most studies examine reduced air pollution and increased physical activity as mediating factors, with stronger support for air pollution reduction as a protective mediator. However, evidence on physical activities as a mediator remains mixed. Some studies suggest that merely perceiving green space enhances brain activity, and exposure to nature is linked to improved test performance. Other potential pathways, such as heat reduction and social interaction, remain underexplored. We highlight the limitations of current methods in distinguishing various forms of green space and emphasize the need for advanced methods, such as local climate zones and street view imagery, for more precise assessment.

Keywords: urban greenery, environmental exposure, cognitive development and impairment, local climate zone, street view imagery, built environment

1. Introduction

Studies have shown that urban green space can benefit people's health (Maas et al., 2006; Barboza et al., 2021; Hunter et al., 2023). With more than half of the world's population currently living in urban areas and this number expected to exceed two-thirds by 2050 (Kohlhase, 2013), it is essential to understand the mechanisms of this relationship—specifically, the pathways through which urban green space benefits health.

In the existing literature, Lachowycz and Jones (2013) developed a socio-ecological framework to understand this relationship, highlighting three key factors: the presence of a park and

its environmental benefits, the pleasure and relaxation derived from viewing greenery, and the usage of green space for physical activities and interactions. Hartig et al. (2014) listed four pathways linking the natural environment to health and well-being: reduced air pollution, increased physical activity, enhanced social interactions, and reduced stress. Additional mechanisms may include noise reduction, improved immune system, urban heat island mitigation, and optimized sunlight exposure (Lee and Maheswaran, 2011). Building on existing frameworks, the effects of green space on human health can be summarized as functional, spatial, and perceptual effects. Functional effects refer to the biochemical interactions of vegetation with the atmosphere and environment, such as reducing air pollution via dry deposition and photosynthesis (Diener and Mudu, 2021), cooling urban heat via evapotranspiration and shading (Cheung et al., 2022), and mitigating noise pollution as trees and shrubs act as natural barriers (Van Renterghem et al., 2012; Dzhambov and Dimitrova, 2014). Spatial effects arise from the physical presence of green space, offering urban residents with a retreat from the built environment dominated by concrete and asphalt, which may be perceived as stressful (Grahn and Stigsdotter, 2010), while also fostering opportunities for recreation, social interaction, and engagement with nature (Sugiyama et al., 2008). Perceptual effects stem from the aesthetic and psychological benefits of green space, enhancing residents' perception of their living environment (Kothencz et al., 2017; Chang et al., 2021). Beautiful landscapes and natural settings offer enjoyment, contributing to well-being and overall satisfaction with urban life (Grahn and Stigsdotter, 2010).

All three functions are critical, but they may arise from different forms of green space. For example, an open grass area may provide space for physical activities and social interactions, but its ability to clean air will be inferior to that of an urban forest with similar coverage. There are various forms of green space. For example, Abhijith et al. (2017) examined the different air pollution abatement performances of trees, hedges, green walls, and green roofs. Beyond conventional green space, blue space is often considered a special type of "green space" due to its similar effects (Lee and Maheswaran, 2011), and both fall under the broader terms "open space" or "public space". Space, as an important component of modern cities, shapes different urban landscapes alongside buildings and infrastructure.

The categorization of urban landscapes can be complicated. There is an existing system called the *local climate zones* (Stewart and Oke, 2012) that describes the components of green space and buildings in urban areas. Green space is, in fact, an umbrella term encompassing various forms of urban landscapes (Zou and Wang, 2021). Within the framework of local climate zones, green space may include dense trees, scattered trees, bush or scrub, low plants, bare rock or paved, bare soil or sand, and water. Their combinations with mixed urban environments (compact or open, low-, mid-, or high-rise) affect the local environment in terms of air, heat, and noise, as well as people's behaviors. For example, compact high-rise urban design, which is typical in Asia, promotes walking and increases daily physical activity (Althoff et al., 2017; Sit et al., 2025). However, compact high-rise development also contributes to urban heat, which may impose health risks (Zheng et al., 2023). It is therefore inaccurate to consider all these urban landscapes as a single type of green space or to assume they provide equally important benefits to human health.

Human health is a broad concept encompassing general well-being, including both physical and mental health (Zautra et al., 2010). Of the top 10 causes of disease worldwide in 2019 (WHO, 2020), green space may affect six of them: (1) ischemic heart disease (Liu et al., 2022); (2)

stroke (Paul et al., 2020); (3) chronic obstructive pulmonary disease (Fan et al., 2020); (4) trachea, bronchus, and lung cancers (Mueller et al., 2022); (5) Alzheimer's disease and other dementia (cognitive functions) (Astell-Burt et al., 2020); and (6) diabetes mellitus (Yang et al., 2023). Focusing on cognitive functions as an example, studies have shown that increased physical activities are associated with lower risks of cognitive impairment in older adults (Dzhambov et al., 2019; Yu et al., 2018; Ruiz-Gonzalez et al., 2021), whereas exposure to higher ambient air pollution, particularly PM_{2.5} and NO₂, is associated to an increased risk of dementia (Chen et al., 2017; Yuchi et al., 2020; Zhu et al., 2023). Additionally, contact with natural environments may benefit cognitive functions and mental health (Bratman et al., 2012). The three effects of urban green space on human health have all been found to provide potential cognitive benefits.

Studies have also illustrated the benefits of urban green space on cognition, either as protection against cognitive decline for adults and older adults (Klompmaker et al., 2021; Paul et al., 2020) or for cognitive development in children (Dadvand et al., 2015). Since different forms of green space vary in their functional, spatial, and perceptual effects, it is crucial to understand how urban green space affects cognitive functions rather than simply stating that "green space is good for cognition". Existing literature provides evidence of the link between green space (exposure) and cognition (outcome), as well as the link between mediators (such as air pollution and physical activity) and cognition. However, it is crucial to recognize the connection between exposure and outcome through mediators. In this scoping review, we summarize current research on green space and cognition (both development and decline) and extract key insights on the functional, spatial, and perceptual effects of green space on cognitive functions. We highlight studies that directly examined pathways and those that used multiple types of green space to infer pathways. This synthesized knowledge will be valuable for designing urban green spaces that enhance living environments.

2. Characterizing Types of Green Space

Green space is an umbrella terminology encompassing various types of spaces (Lee and Maheswaran, 2011; Panduro and Veie, 2013; Abhijith et al., 2017; Zou and Wang, 2021). These spaces can be categorized based on the volume of vegetation. Figure 1 illustrates three examples. Figure 1a shows an open grass area in Los Angeles. Such small open grass areas are common worldwide, particularly in public spaces like universities and parks, offering spaces for entertainment and relaxation (Thompson, 2002). Figure 1b depicts a large urban forest (Central Park) in New York City, and similar settings can be found in Paris (Wood of Vincennes) and other cities around the world (Hopkins, 2015). Compared to open grass, urban forests contain more biomass and are more effective in reducing air pollution (Godina et al., 2023), blocking traffic noise, and mitigating urban heat islands (Gaffin et al., 2008). However, urban forests are generally less accessible than open grass areas, as they have limited entry points, reducing opportunities for casual daily use (Ha et al., 2022). Figure 1c shows Sai Wan Pier in Hong Kong, where people gather to enjoy the sunset. Similar settings include the Santa Monica Pier and other coastal city piers (Fullerton, 2011; Pryor, 2022). Although these various forms of open space have little to no vegetation, they serve similar social and recreational functions as open grass areas or urban forests by providing spaces for gathering and scenic enjoyment (Mueller et al., 2020; Pryor, 2022). Since open

93 space without greenery is often more accessible than urban forests and offers larger areas than
 94 small open grass patches—important for physical activities—the number of people visiting (and
 95 the associated health benefits) may be greater (Mueller et al., 2020).



Figure 1: Examples of urban green space (from left to right: open grass, urban forest, and open space without green). Photos of open grass and open space without green are from the authors. Credit (photo of the urban forest): Central Park Conservancy (<https://www.centralparknyc.org>).

96 2.1. Functional, Spatial, and Perceptual Effects of Green Space

97 A more detailed framework categorizes urban green space into six forms based on their ability
 98 to clean air (Abhijith et al., 2017). We adopt this framework and extend their ratings to include
 99 potential spatial and perceptual effects (Table 1). The ratings for functional effects, primarily air
 100 pollution reduction, are adapted from Abhijith et al. (2017), while the ratings for spatial effects are
 101 based on typical size, with larger areas receiving better ratings for spatial benefits. Trees gener-
 102 ally provide good functional effects, including roadside trees, dense forests with access, and dense
 103 forests without access, but their spatial effects range from moderate to poor due to limited accessi-
 104 bility. Open grass has moderate functional effects for cleaning air but not for blocking noise, while
 105 offering good spatial effects for physical activities and social interaction. Green wall/roof (Abhi-
 106 jith et al., 2017), a specialized form of urban greening, particularly in mid- and high-rise built
 107 environments, has lower functional effects for cleaning air or absorbing heat and does not pro-
 108 vide space. Maximizing the value of urban green space is critical, as land in cities is expensive.
 109 Therefore, understanding the mechanisms linking urban green space to human health is essential
 110 for allocating more beneficial forms of green space within the local built environment.

Table 1: Forms of urban green space and grades of their functional, spatial, and perceptual effects on health

Type	Functional	Spatial	Perceptual
Roadside trees/hedges	good	poor	
Green walls/roofs	moderate	poor	
Open grass	moderate	good	
Open space without green	poor	good	depends on the landscape
Dense forests with access	good	moderate	
Dense forests without access	good	poor	

111 **3. Search Criteria**

112 We searched the keywords (“green space” or “greenness”) and (“Alzheimer” or “dementia”)
113 or (“cognitive” or “cognition”) in all full texts on the Web of Science up to December 31, 2024,
114 yielding a collection of 1105 documents. After an initial screening, 888 articles were included,
115 removing review articles, editorials, abstracts, and preprints. The diagram illustrating the selection
116 process is provided in the supplementary materials. Among the 888 articles, 20 were excluded for
117 not being in English, 468 for not being epidemiology studies or not examining associations, 232
118 for not addressing cognitive functions, 81 for not considering green space exposure, and 46 for not
119 examining pathways. Some articles met multiple exclusion criteria, but we listed the first identified
120 reason for exclusion. Although mental health (e.g., depression) is often related to cognitive studies,
121 we excluded studies focused on mental health as it differs from cognitive outcomes. After these
122 exclusions, 41 articles underwent further examination by reviewing their methodology, data, and
123 results. Four additional articles were excluded because they did not examine pathways. Ultimately,
124 37 articles were included in this review: 20 directly examined pathways between green space and
125 cognitive functions, while 17 examined them indirectly. A full list of included and excluded studies
126 is provided in the supplementary materials.

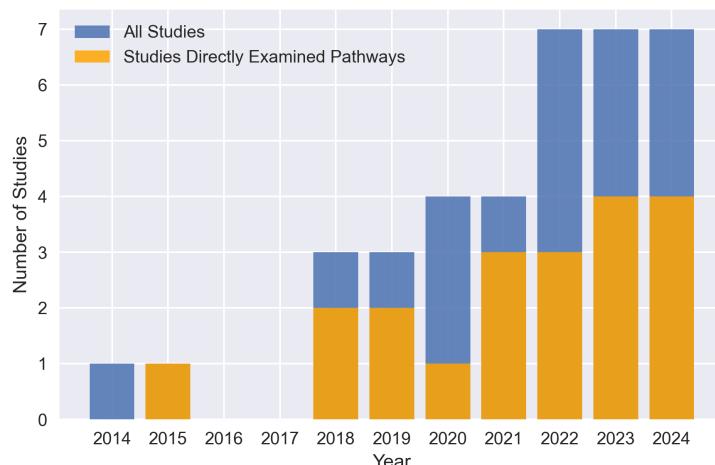


Figure 2: Number of studies each year, covering the period from January 1, 2014 to December 31, 2024. The number of studies that directly examined pathways is highlighted in orange.

127 The number of research articles devoted to this topic has grown over time, as shown in Figure
128 2. The earliest study dates back to 2014, and from 2014 to 2024 (as of September 30, 2024), the
129 number of articles has steadily increased each year. The majority of these works directly examined
130 the mediators or the associations between green space and cognition.

131 **4. Results**

132 We summarized the 20 studies that directly examined pathways based on cognitive develop-
133 ment (school-aged children) in Table 2 and cognitive impairments (mid to older adults aged 45+)

134 in Table 3. One study by Wang et al. (2023) included two age groups: school-aged children and
 135 mid-life adults. Since the significant result was found in adolescents rather than mid-life adults,
 136 and the sample size of adolescents was three times larger, we categorized this article under the
 137 cognitive development group.

Table 2: Summary of 12 out of 20 selected articles directly examining the pathways between green space and cognition (cognitive development group).

Study	Sample	Age	Region	Outcome	Exposure	Model	Pathways
Almeida (2022)	3827	10	Portugal	IQ test	30-m NDVI within 800 m	Generalized mixed models, adjusted for NO ₂ and physical activity (no association with NO ₂)	Spatial (activity), no functional (air)
Asta (2021)	465	7	Italy	IQ test	30-m NDVI within 500 m	Generalized additive models, adjusted for NO ₂ (explained 35% of the association)	Functional (air)
Binter (2022)	5403	4-5	UK, France, Spain, & Greece	Verbal and non-verbal abilities, motor function	30-m NDVI within 500 m	Multi-exposure analysis with Deletion-Substitution-Addition algorithm, adjusted for NO ₂ and PM _{2.5} (explained 75% of the association)	Functional (air)
Chang (2021)	44	mean 23.7	HK (lab)	Biomarker via fMRI (posterior cingulate cortex activity), behavioural tasks	Images of green urban landscape	General linear model (directly related to perceptual effects)	Perceptual
Claesen (2021)	851 schools	Students in Years 3 and 5	Australia	School scores (reading, writing, spelling, grammar & punctuation, & numeracy)	30-m NDVI within 2000 m	Generalized linear models, adjusted for traffic-related air pollution (explained 50-66% of the association)	Functional (air)
Dadvand (2015)	2593	7-10	Spain	12-month change of working memory and attention	5-m NDVI within 250 m (residential), 50 m (commuting route), and 50 m (school)	Multilevel modeling, adjusted for traffic-related air pollution (explained 20-65%)	Functional (air)
Liao (2019)	1312	2	China	Neurodevelopment of children	500-m NDVI within 300 m	Multiple linear regression models, adjusted for traffic-related air pollution PM _{2.5} (explained 14-28% of the association)	Functional (air)
Mason (2022)	65	8.25±0.43	Italy	Math and attention test	Exposure to natural green space (vs. controlled group in normal classroom)	Linear mixed models, adjusted for perceived restorativeness score (explained 71% of the association)	Spatial (interaction)
Wallner (2018)	64	16-18	Austria	Attention test	Staying in green space	Stanine value comparison and t-test (higher score associated with large park)	Spatial (activity)
Wang (2023)	6220, 2623	14.4±0.5, 48±5.0	Australia	Cogstate tests	30-m NDVI within 1600 m	Mixed-effects generalized linear models, adjusted for PM _{2.5} and NO ₂ (no mediated percentage provided, adjustment only found in adolescents but not in mid-life adults)	Functional (air)
Binter (2024)	2725; 95	9-12	Netherlands, France	white matter microstructure outcomes including average of fractional anisotropy (FA) and mean diffusivity (MD) from imaging data	access to major parks from land use data; area surface of green space,	interquartile range linear regressions controlling for PM _{2.5} , NOx, BMI, etc.; road-traffic noise mediated 19% and 52% of these associations; air pollution no association	Functional (noise)
Buczylowska (2024)	689	10-13	Poland	ADHD diagnosis	percentage of grass and tree cover in 500 m and 1 km buffers from 20-m land cover data, measured from life-long residential addresses	Structural equation modeling, with mediators including social cohesion (8%), perceived green space (7%), and physical activity (6%)	Spatial (social, activity); Perceptual

138 4.1. Study design

139 The 20 selected articles can be classified into three categories. Twelve research works are clas-
 140 sic population health studies that examined a specific population or cohort, measuring individual

Table 3: Summary of 8 of 20 selected articles directly examining the pathways between green space and cognition (cognitive impairment group).

Study	Sample	Age	Region	Outcome	Exposure	Model	Pathways
Astell-Burt (2023)	109,688	45+	Australia	dementia	Tree canopy cover from 2-m land use data within 1.6 km	Marginal structural models with mediators including physical activity, social support, and sleep duration	Spatial (activity, social)
Dzhambov (2019)	112	45-55	Bulgaria	cognitive tests (main), cortical thickness from MRI (25 samples)	30-m NDVI within 100 m	Multivariate linear regression models, adjusted for lower waist circumference (no mediated percentage provided) and NO ₂ (not significant)	Spatial (activity)
Hu (2023)	375,342	57.1±8.0	UK	Cases of dementia, Alzheimer's, and vascular dementia	10-m green space within 300 or 1000 m	Cox proportional hazard regression models, adjusted for PM _{2.5} and PM ₁₀ (explained 25-66% of the association)	Functional (air)
Yu (2018)	3240	65+	HK	Improvement in frailty status	1-m NDVI within 300 m	Ordinal logistic regression, adjusted for physical activity (explained 45% of the association)	Spatial (activity)
Yuchi (2020)	678,000	45-84	Canada	Dementia, Alzheimer's and Parkinson's disease	30-m NDVI within 100 m	Greenness mediated the association between road proximity (air pollution) and odds ratios of Alzheimer's disease	Functional (air)
Zhu (2023)	29,025	63.3±9.4	China	Cases of neurodegenerative diseases	500-m NDVI within 1000 m	Linear model and Cox proportional hazard model, adjusted for PM _{2.5} , PM ₁₀ and NO ₂ (PM ₁₀ explained 33.9% of the association between neurodegenerative disease and NDVI)	Functional (air)
Besser (2024)	162	50+	USA	Cognitive tasks and biomarkers, including number symbol coding task, Montreal Cognitive Assessment and white matter hyperintensity volume	Perceived greenspace access and time spent in greenspaces	Multivariable linear regression models controlling for age, gender, race, ethnicity, education; examined interpersonal discrimination	Spatial (social); Perceptual
Cerin (2024)	1160	60+	Australia	mild cognitive impairment via a series of cognitive tests including Wechsler Memory Scale, Symbol Digit Modalities Test, California Verbal Learning Test, and Mini-Mental State Examination	1-km buffer along road networks, air pollution (PM _{2.5} , NO ₂), land use data (tree cover, parks, blue space)	Generalised additive mixed models controlling with directed acyclic graphs	Functional (air); Spatial (activity)

141 cognitive test performance and individual green space exposure (Almeida et al., 2022; Asta et al.,
142 2021; Astell-Burt et al., 2023; Dadvand et al., 2015; Hu et al., 2023; Yu et al., 2018; Yuchi et al.,
143 2020; Zhu et al., 2023; Besser et al., 2024; Cerin et al., 2024; Binter et al., 2024; Buczyłowska
144 et al., 2024). One study examined both adolescents and middle-aged adults (Wang et al., 2023).
145 Within the classic study design, two studies focused on unique populations—infants and very
146 young children under age 5 (Binter et al., 2022; Liao et al., 2019). In both studies, children and
147 their mothers were paired to assess cumulative exposure to green space and air pollution prior to
148 birth. One research work used a case-control study design in a school setting to test whether expo-
149 sure to natural green space benefited math and attention test scores (Mason et al., 2022). In another
150 case-control experiment, Wallner et al. (2018) tested whether staying in green space improved at-
151 tention tests among adolescents aged 16-18. Two research works used a case-control study design
152 in a lab environment, employing equipment to capture brain activity signals via fMRI machines
153 (Chang et al., 2021). Two studies also obtained biomarker imaging data (Dzhambov et al., 2019;
154 Binter et al., 2024). Additionally, there is one ecological study by Claesen et al. (2021).

155 4.2. Populations

156 Of the 20 research works, four studies focused on older adults aged 60+ (Hu et al., 2023; Yu
157 et al., 2018; Zhu et al., 2023; Cerin et al., 2024), one on middle-aged adults (Dzhambov et al.,
158 2019), and three on adults aged 45-85 (Astell-Burt et al., 2023; Yuchi et al., 2020; Besser et al.,
159 2024). Three studied very young children and traced their exposure prior to birth (Binter et al.,
160 2022; Liao et al., 2019; Binter et al., 2024). Seven studies focused on school-aged children and
161 adolescents (Almeida et al., 2022; Asta et al., 2021; Claesen et al., 2021; Dadvand et al., 2015;
162 Mason et al., 2022; Wallner et al., 2018). One study combined middle-aged adults and adolescents
163 (Wang et al., 2023), and one examined young adults with an average age of 23.7 years (Chang
164 et al., 2021).

165 4.3. Outcomes

166 The most common outcome is cognitive tests, which are often designed locally to accommo-
167 date local language and culture. The Wechsler Intelligence Scale for Children is the most popular
168 cognitive test for children, including three dimensions of cognitive function: verbal, performance,
169 and global (Watkins et al., 1997). Two studies in Portugal and Italy, respectively, used this cog-
170 nitive test (Almeida et al., 2022; Asta et al., 2021). Other cognitive tests include the McCarthy
171 Scales of Children's Abilities in four European countries (Binter et al., 2022), the Bayley Scales
172 of Infant Development (BSID) (Liao et al., 2019), the Stroop Test and Neurobehavioral Evalu-
173 ation System 3, or school scores (Claesen et al., 2021). Some studies used biomarkers, including
174 human posterior cingulate (Chang et al., 2021) and cortical thickness (Dzhambov et al., 2019) via
175 MRI machine. While most studies focused on an end-point outcome in a cross-sectional fashion,
176 a few studies examined the change in cognitive functions over time, such as the 12-month change
177 in working memory and attention (Dadvand et al., 2015), the d2-R test (Wallner et al., 2018),
178 improvement in frailty status (Yu et al., 2018), and math and attention tests before and after expo-
179 sure to green space (Mason et al., 2022). Studies on older populations also used the incidence of
180 dementia, Alzheimer's, Parkinson's disease, and neurodegenerative diseases as outcome variables
181 (Astell-Burt et al., 2023; Hu et al., 2023; Yuchi et al., 2020; Zhu et al., 2023).

Table 4: The mechanisms by which green space could affect cognition. Those marked with an asterisk (*) did not directly examine the pathways, but we may infer such a mechanism by comparison of types of green space. The studies are listed by the last name of the first author and year of publication.

		Positive protective effects	No association or inconclusive
Functional Air		Asta (2021), Astell-Burt (2020)*, Binter (2022), Bijnens (2022)*, Cerin (2024), Claesen (2021), Crous-Bou (2020)*, Dadvand (2015), Godina (2023)*, Hu (2023), Jarvis (2022)*, John (2023)*, Liao (2019), Maes (2021)*, Subiza-Perez (2023)*, Tallis (2018)*, Wang (2023), Wu (2014)*, Yuchi (2020), Zhu (2023)*	Almeida (2022)
Noise		Binter (2024)	Garkov (2024)*
Heat		None	None
Spatial	Physical Activity	Astell-Burt (2023), Almeida (2022), Brown (2024)*, Buczylowska (2024), Cerin (2024), Cherrie (2019)*, Dadvand (2017)*, Dzhambov (2019), Yu (2018), Zhang (2022)*	Astell-Burt (2020)*, Maes (2021)*, Wu (2020)*
	Social Interaction	Astell-Burt (2023), Buczylowska (2024), Hu (2024)*	None
	Interaction with Nature	Mason (2022), Wallner (2018)	None
Perceptual Perception, relaxing and landscape		Besser (2024), Bijnens (2022)*, Buczylowska (2024), Chang (2021)	None

182 4.4. Exposures

183 The measurement of exposures is limited compared to the measurement of outcomes. The
 184 majority of studies (17 out of 20) used either NDVI from satellite images for greenness or extracted
 185 urban green space from land use data. The spatial resolution of NDVI or land use data varies,
 186 from 1 m land use to 30 m to 250-500 m NDVI (Landsat and MODIS, respectively). In terms
 187 of the effective range of green space, a buffering radius of up to 1.6 km (one mile) was common,
 188 assuming it matches a walking distance of 15-20 min. Other common buffering radii included
 189 300-500 m (15 min walking distance for older adults) or a search range from 100 to 2000 m. A
 190 few exceptions include one study based on visual images of green space (Chang et al., 2021) and
 191 two studies examining before-after exposure to nature (Mason et al., 2022; Wallner et al., 2018).

192 4.5. Pathways directly modeled from the 20 studies

193 In Table 4, we summarize and categorize the 36 reviewed papers, including those that directly
 194 and indirectly modeled the pathways, based on their mechanisms. Research works from which we
 195 could infer potential pathways are marked with an asterisk (*) and are discussed later in Section
 196 4.6. In this section, we first review the 20 studies that directly examined the pathways.

197 4.5.1. Functional effects via reduced air pollution

198 Twelve studies analyzed the mediator of functional effects on reduced air pollution via model-
 199 ing. Asta et al. (2021) presented a directed acyclic graph showing potential mechanisms, in which

200 NO_2 was a potential mediator between greenness and cognitive functions. Since the correlation
201 between NDVI and NO_2 was not significant, they proposed estimating the natural direct effect,
202 controlled direct effect, and natural indirect effect to determine NO_2 's role as a mediator, which ac-
203 counted for 35% of the total effects of green space. The other 65% may be attributed to other forms
204 of air pollution (e.g., $\text{PM}_{2.5}$), physical activities, etc., but required further investigation. Binter et al.
205 (2022) showed that $\text{PM}_{2.5}$ mediated 74% of the association between NDVI in a 300 m buffer and
206 children's verbal abilities. Traffic-related air pollution was found to mediate 22-25% of the associa-
207 tion between NDVI in a 300 m buffer and students' academic performance in Australia (Claesen
208 et al., 2021). In a UK study by Dadvand et al. (2015), traffic-related air pollution explained 20-
209 65% of the association between school greenness and 12-month progress in cognitive functions.
210 Liao et al. (2019) examined the association between NDVI and early childhood neurodevelopment
211 mediated by $\text{PM}_{2.5}$ and found that reduced traffic-related air pollution explained 13.6-28.0% of the
212 association. Hu et al. (2023) used the Baron method to study the mediation effects of air pollution
213 and found that $\text{PM}_{2.5}$ and PM_{10} generated from land use regression models might mediate air pol-
214 lution reduction among 375,342 UK Biobank participants. Rodriguez-Loureiro et al. (2022) found
215 that reduced air pollution ($\text{PM}_{2.5}$, NO_2) was directly associated with lower neurodegenerative dis-
216 ease mortality, but green space still played a role in reducing neurodegenerative disease risk in
217 Belgium after controlling for air pollution. Yuchi et al. (2020) showed that greenness mediated
218 the association between road proximity and Parkinson's disease and dementia by 0.3%-28% in a
219 population-based study in Vancouver, Canada (N=678,000). Cerin et al. (2024) used generalized
220 additive mixed models with directed acyclic graphs to directly estimate air pollution ($\text{PM}_{2.5}$, NO_2)
221 as mediators.

222 4.5.2. Functional effects via noise reduction

223 With research continuing to grow, one recent study by (Binter et al., 2024) used interquartile
224 range linear regressions to directly estimate the contributions of noise reduction as a mediator
225 between green space exposure and cognitive outcome measured as white matter volume. The
226 results from the Netherlands and France show that road-traffic noise mediated 19% and 52% of
227 these associations, with no moderation effects from air pollution.

228 4.5.3. Spatial effects via physical activity

229 Five studies showed direct evidence that physical activity may mediate the association be-
230 tween green space and cognitive functions. Almeida et al. (2022) used the self-reported daily
231 time in physical activity and found a positive association between physical activity and IQ test.
232 Yu et al. (2018) showed that green space both directly and indirectly (via physical activity) af-
233 fected frailty status using analysis of variance (ANOVA). Dzhambov et al. (2019) concluded that a
234 greener neighborhood might be associated with better cognitive functions in middle-aged Bulgari-
235 ans, where the association is possibly mediated by lower central adiposity. Recently, Astell-Burt
236 et al. (2023) used marginal structural models to directly model the causal mediation effects and
237 found that the association between urban tree canopy and dementia was mediated partially by
238 physical activity and diabetes. Buczyłowska et al. (2024) used structural equation modeling to
239 directly estimate the contribution of each pathway towards ADHD diagnosis, of which physical
240 activity accounted for the third-largest contribution, being 6%.

241 *4.5.4. Spatial effects via social interaction or interaction with nature*

242 Four studies showed the benefit of interaction with nature. Mason et al. (2022) designed an
243 experiment on two groups of students in Italy and set them in two different classroom settings:
244 a regular indoor classroom and a green school garden. Students were randomized and switched
245 to the other classroom setting one week later. They found that the green school garden could
246 enhance children's attention and math performance. Although they did not identify the specific
247 items (e.g., daylight, breeze) within nature/green space as the mediator, their result is sufficient to
248 conclude that green space provides the spatial function for exposure to nature, increasing cogni-
249 tive performance. Similarly, Wallner et al. (2018) conducted an experiment with three classroom
250 settings (small park, large park, forest) and found higher d2-R test scores after students had stayed
251 in green space after lunch break. The findings from Astell-Burt et al. (2023) also showed that
252 social support partially mediated the association between urban tree canopy and dementia. Social
253 cohesion was the mediator with the largest contribution between green space exposure and ADHD
254 diagnosis (Buczyłowska et al., 2024).

255 *4.5.5. Perceptual effects*

256 Four studies show that the perception of urban green space is enough to benefit people's cog-
257 nition. In a lab environment with 44 participating young adults, viewing images of green urban
258 landscapes was associated with activity in brain regions for spatial processing, spatial and execu-
259 tive attention, and sensory encoding (Chang et al., 2021). In a well-designed study using structural
260 equation modeling, Buczyłowska et al. (2024) showed that subjectively perceived green space was
261 the second-largest moderator between green space exposure and ADHD diagnosis, being 7%.

262 *4.6. Potential pathways inferred from another 17 articles*

263 Table 5 lists another 17 articles that did not directly investigate the mechanisms through mod-
264eling but from which we can infer the potential pathways by comparing green space exposure.
265 This comparison can be drawn from the variation among different types of green space exposure,
266 the comparison between green space and blue space or non-photosynthetic vegetation, or inferred
267 through a co-modeling method (where the third variable is not used as a mediator). Even seasonal
268 comparisons can indirectly suggest such pathways due to the reduced or nonexistent air pollution
269 mitigation effects of trees in winter. From these indirect approaches, we may infer the potential
270 mediators from the results of an additional 17 articles. These research works are also listed in
271 Table 4 with an asterisk (*). We list them here in Table 5.

272 *4.6.1. Functional effects*

273 By distinguishing the types of green space, we may infer some potential pathways. In an
274 Australian study by Astell-Burt and Feng (2020), land cover data were used as a measure of
275 exposure, and more tree canopy within a 1.6 km radius showed protective effects on self-rated
276 memory, but such an association was not seen between open grass and memory. As a result, the
277 benefit of green space is likely due to effects that tree canopy can provide but open grass cannot,
278 such as reducing air pollution. Similarly, John et al. (2023) concluded that areas with over 30%
279 tree cover showed positive impacts on healthy aging, but the same benefits were not seen from
280 grass cover, thus supporting the functional pathways. A British study drew a similar conclusion

Table 5: Summary of 17 articles from which we can infer potential pathways via which green space affects cognition, by comparing different types of green space. They are marked with an asterisk (*) and listed by first author and year of publication.

Study	N	Pathways	Inference	Category
Ahmed (2022)*	936	Spatial (activity)	Proximity to non-photosynthetic vegetation is associated with children's school scores, but residential green space is not	non-green space
Wu (2020)*	4955	Spatial (activity)	Proximity to public parks is associated with lower odds of dementia, but greenness is not	non-green space
Subiza-Perez (2023)*	1,738	Functional (air)	Blue space did not show positive impacts but green space (tree cover) showed positive impacts	green space comparison
Astell-Burt (2020)*	45,644	Functional (air)	Exposure to tree canopy is associated with better memory, but open grass is not	green space comparison
Bijnens (2022)*	596	Functional (air)	Green space higher than 3 m is associated with faster reaction time, but low green is not	green space comparison
Godina (2023)*	2141	Functional (air)	No association between overall greenness and mild cognitive impairment (MCI); forest greenness associated with lower odds of MCI; green space diversity associated with lower hazard of incident dementia	green space comparison
Jarvis (2022)*	27,539	Functional (air)	Stronger positive association for residential exposure to tree cover relative to grass cover (beta coefficient: 0.26 ± 0.11 vs. 0.12 ± 0.10)	green space comparison
Maes (2021)*	3568	Functional (air)	Higher exposure to woodland, not grassland, is associated with higher cognitive score	green space comparison
Tallis (2018)* schools	495	Functional (air)	Urban trees associated with higher elementary school test scores, but not rural trees	green space comparison
John (2023)*	22,715	Functional (air)	Areas with over 30% tree cover show positive impacts on healthy ageing, but the same benefits were not seen from grass cover	green space comparison
Hu (2024)*	422,649	Spatial (interaction)	Domestic gardens have protective effects (better than other green space), a spatial effect for social interaction	green space comparison
Cherrie (2019)*	281	Spatial (activity)	Park availability is associated with better cognitive aging with low traffic density, but not with high traffic density	co-modeling but not mediators
Crous-Bou (2020)*	2743	Functional (air)	Greenness not associated with greater cortical thickness after adjusting for air pollution	co-modeling but not mediators
Zhang (2022)*	16,337	Spatial (activity)	Lower precipitation is associated with slower cognitive decline, precipitation limits usage of green space, and thus physical activity is a pathway	co-modeling but not mediators
Brown (2024)*	230,738	Spatial (activity)	Older adults living in high greenness neighborhoods had lower odds of AD incidence, walkability decreased the odds via co-modeling	co-modeling but not mediators
Garkov (2024)*	12,159	Spatial (noise)	no evidence of an association between PM _{2.5} exposure and cognitive performance in children in England, nor between noise and cognitive performance	co-modeling but not mediators
Wu (2014)* schools	905	Functional (air)	Positive association between greenness in Spring and Summer and academic performance, but negative association between greenness in Autumn and academic performance	seasonal comparison

281 on adolescents' cognition and mental health, finding that woodland—but not grassland and blue
282 space—was associated with higher cognitive development scores (Maes et al., 2021). Godina et al.
283 (2023) found that forest green space (rather than other types of green space) was associated with
284 a lower risk of mild cognitive impairment, which could be due to the strong photosynthesis of
285 forests leading to air pollution reduction. Jarvis et al. (2022) investigated the impacts of early-life
286 green space exposure on childhood development in Vancouver, Canada, and found that tree cover
287 had a stronger association with childhood development than grass cover. Tallis et al. (2018) found
288 that only urban trees, rather than rural trees, were associated with higher elementary school test
289 scores in California, which they argued was due to the higher air pollution and thus greater air
290 pollution reduction by trees in urban school settings. Their findings were also supported by the
291 null association between NDVI/agricultural areas and test scores.

292 Non-photosynthetic vegetation and photosynthetic vegetation's seasonal effects provide an-
293 other way to infer potential pathways. An early ecological study by Wu et al. (2014) showed a
294 positive association between the greenness of schools in March (Spring) and July (Summer) and
295 academic performance but a negative association with the greenness of schools in October (Au-
296 tumn) in Massachusetts after adjusting for socioeconomic factors. The authors claimed this might
297 be due to the inaccurate coarse-resolution NDVI from MODIS (250–500 m resolution). Another
298 possible explanation may be that the photosynthesis process by plants is strong in Spring and Sum-
299 mer but weak in Autumn, leading to differences in the functional (biochemical) effects of green-
300 ness on cognitive development. However, more investigation is needed to confirm one or both
301 assumptions. Ahmed et al. (2022) found an association between exposure to non-photosynthetic
302 vegetation and poor academic performance; since non-photosynthetic vegetation does not provide
303 an air pollution reduction function, it may be viewed as indirect evidence of green space's func-
304 tional effects. Another way to compare green space and non-green space exposure is through blue
305 space (Subiza-Pérez et al., 2023).

306 An interesting study by Bijnens et al. (2022) is also worth mentioning. They treated air pol-
307 lution as a confounding variable instead of a mediator and found that the height of green space
308 was associated with better attention in children. Although they did not report the coefficient of
309 air pollution, from their study, we can infer the functional effects via reduced air pollution. One
310 study (Dadvand et al., 2017) showed a null association between tree cover and better attention
311 among children but a positive association between general residential greenness and better atten-
312 tion. In other words, open grass is better than tree cover. We may infer that air pollution is not
313 the primary mediator between green space and cognitive development. Crous-Bou et al. (2020)
314 showed that green space was not associated with Alzheimer's dementia after controlling for air
315 pollution, from which we may infer that air pollution is the pathway between green space and
316 cognition.

317 Three studies discussed air pollution but did not mention whether it was related to green
318 space (Crous-Bou et al., 2020; Falcón et al., 2021; Yuchi et al., 2020). They examined multiple
319 exposures, including air pollutants and green space individually, but did not analyze the pathways.
320 From their results, we can only infer that, in addition to air pollution, other mechanisms contribute
321 to the association between green space and cognition.

322 4.6.2. *Spatial effects*

323 Spatial effects can be indirectly inferred from various types of green space. Cherrie et al.
324 (2019) showed that park availability in adolescence with low traffic accidents was associated with
325 better cognitive aging in later life, but the association was not significant with high traffic acci-
326 dents. High traffic accidents (and possibly high traffic volumes) may reduce adolescents' usage of
327 parks, thereby limiting their physical activity. Almeida et al. (2022) found a positive association
328 between students' IQ tests and urban green space but not with greenness. Greenness is related to
329 vegetation's photosynthetic function, which is associated with air pollution reduction; the fact that
330 greenness is not a positive factor but exposure to urban green space is suggests that the "space"
331 function likely plays a major role. Zhang et al. (2022) analyzed the association between green
332 space exposure (with physical activity, air pollution) and cognitive test scores from the China
333 Health and Retirement Longitudinal Study. Significant associations were found between higher
334 test scores and lower precipitation as well as higher physical activity. Since precipitation impacts
335 green space accessibility, the authors were confident that physical activity was at least one path-
336 way through which green space affects cognitive functions. Similarly, Hu et al. (2024) showed that
337 domestic gardens, compared to other green spaces, provided better protective effects, and Brown
338 et al. (2024) showed that walkability decreased the odds of AD incidence in co-modeling.

339 Similarly, the opposite association may be inferred from three studies: a positive association
340 was found between memory and tree canopy but not open grass (Astell-Burt and Feng, 2020),
341 between adolescents' cognition and woodland but not grassland (Maes et al., 2021), and between
342 dementia and living close to daily amenities but not local green space (Wu et al., 2020).

343 4.6.3. *Perceptual effects*

344 Similarly, the study by Bijnens et al. (2022) showed that even after controlling for air pollu-
345 tion, greater surrounding green space, such as trees, remained associated with better attention in
346 children, whereas low green space was not. From this result, we may infer that, in addition to
347 functional effects, trees also provide perceptual benefits that enhance cognitive function. A recent
348 study found that while engaging in sports activities in both green outdoor environments and ur-
349 ban indoor environments improves short-term memory, better cognitive function was observed in
350 the green exercise group (Baena-Extremera et al., 2024). The authors suggest that the perception
351 of greenness may preferentially stimulate the right frontal areas and call for further research to
352 explore the underlying mechanisms.

353 5. **Discussion**

354 5.1. *Summary of findings*

355 Despite a large number of studies examining the association between green space and cognitive
356 functions, only a very few have investigated potential pathways and mediators. Of these, most used
357 a framework to examine a single mediator, with air pollution reduction and physical activity being
358 the most common. A comprehensive understanding of the modification effects across functional,
359 spatial, and perceptual pathways is still lacking. In addition, the first study to directly examine
360 noise as a potential mediator was only recently published in 2024 (Binter et al., 2024), despite
361 its established association with cognitive functions (Jafari et al., 2019). Yet, recent studies have

362 introduced more comprehensive frameworks, particularly those using structural equation modeling
363 to estimate the contributions of one or more mediators (Buczyłowska et al., 2024). Our review thus
364 focuses on directly modeling modification effects between green space and cognitive functions.

365 Indirect inference through green space comparisons also provides valuable insights. By analyzing
366 different types of green space, 17 additional studies have contributed to our understanding
367 of potential pathways. However, a comprehensive characterization of land use data remains a
368 significant gap. Implementing globally consistent land use modeling approaches, such as the local
369 climate zone framework, could enhance consistency and improve our understanding of how
370 diverse landscapes contribute to cognitive development and decline.

371 5.2. *Lack of mechanism studies focusing on older adults*

372 While reading the full text of the candidate articles, we observed that the number of studies on
373 older adults and children/adolescents was relatively similar. However, among the 20 articles that
374 directly examined underlying mechanisms, only three of them focused on older adults aged 60
375 and older. This imbalance has its reasons: (1) case-control studies involving school-aged children
376 provide a natural experimental framework that has drawn researchers' interest, and (2) assessing
377 cumulative lifetime exposure to green space and other environmental factors solely based on
378 the current residential address of older adults is challenging. However, since cognitive impairment
379 diseases such as dementia and Alzheimer's predominantly affect older adults, understanding
380 how cognitive decline unfolds across the life course is crucial. Investigating older populations'
381 lifetime exposure to green space, air pollution, and physical activity through modeling will help
382 achieve this goal. While compiling a comprehensive cohort or dataset would be costly, large lifetime
383 datasets, such as the UK Biobank, could provide valuable insights by enabling near-lifetime
384 exposure assessments.

385 5.3. *Limited methods for measuring green space*

386 Land cover data and NDVI were the only two measures of green space used in studies con-
387 ducted outside of a lab setting. The spatial resolution ranged from 500 m for MODIS to 30 m for
388 Landsat, 10 m for Sentinel, and 1 m for IKONOS or high-resolution land cover data. Surprisingly,
389 seven studies used MODIS data despite the widespread availability of higher-resolution land cover
390 datasets and Landsat imagery worldwide (de Keijzer et al., 2018; Jin et al., 2021; Liao et al., 2019;
391 Markevych et al., 2019; Wu et al., 2014; Xu et al., 2019; Zhu et al., 2020). At this scale, only very
392 large areas of green space can be captured, while many roadside trees remain undetected in NDVI
393 values. In some studies (de Keijzer et al., 2018; Markevych et al., 2019; Jin et al., 2021; Liao
394 et al., 2019; Zhu et al., 2020), the buffer zone was barely one or two pixels wide. These rough,
395 imprecise measurements of green space may explain why some studies report null or inconclusive
396 associations between green space and cognitive function (Markevych et al., 2019; Jin et al., 2021).

397 Surprisingly, despite the growing trend in health studies of shifting from satellite-derived over-
398 head NDVI to street-view greenness using panoramic images (as illustrated in Figure 3)(Larkin
399 and Hystad, 2019; Kang et al., 2020; Xiao et al., 2021), no study has yet adopted this new mea-
400 sure of green space. Horizontal-view assessments of greenness often provide more precise mea-
401 surements. Figure 3 compares NDVI values at 500 m (MODIS) and 30 m (Landsat) resolutions
402 alongside street-view images at locations A (Hexagon) and B (Pentagon). In the coarse-resolution

403 NDVI, both locations share the same value, making it impossible to distinguish differences in ex-
 404 posure. In the high-resolution NDVI, a more accurate measure emerges, revealing that location
 405 A has higher NDVI exposure. Using street-view imagery combined with semantic segmentation
 406 techniques such as Segment Anything(Kirillov et al., 2023), we estimate that location A has 20%
 407 green space exposure, while no greenery is visible at location B. Street-view imagery thus provides
 408 a more precise assessment of green space exposure, particularly in complex urban environments.

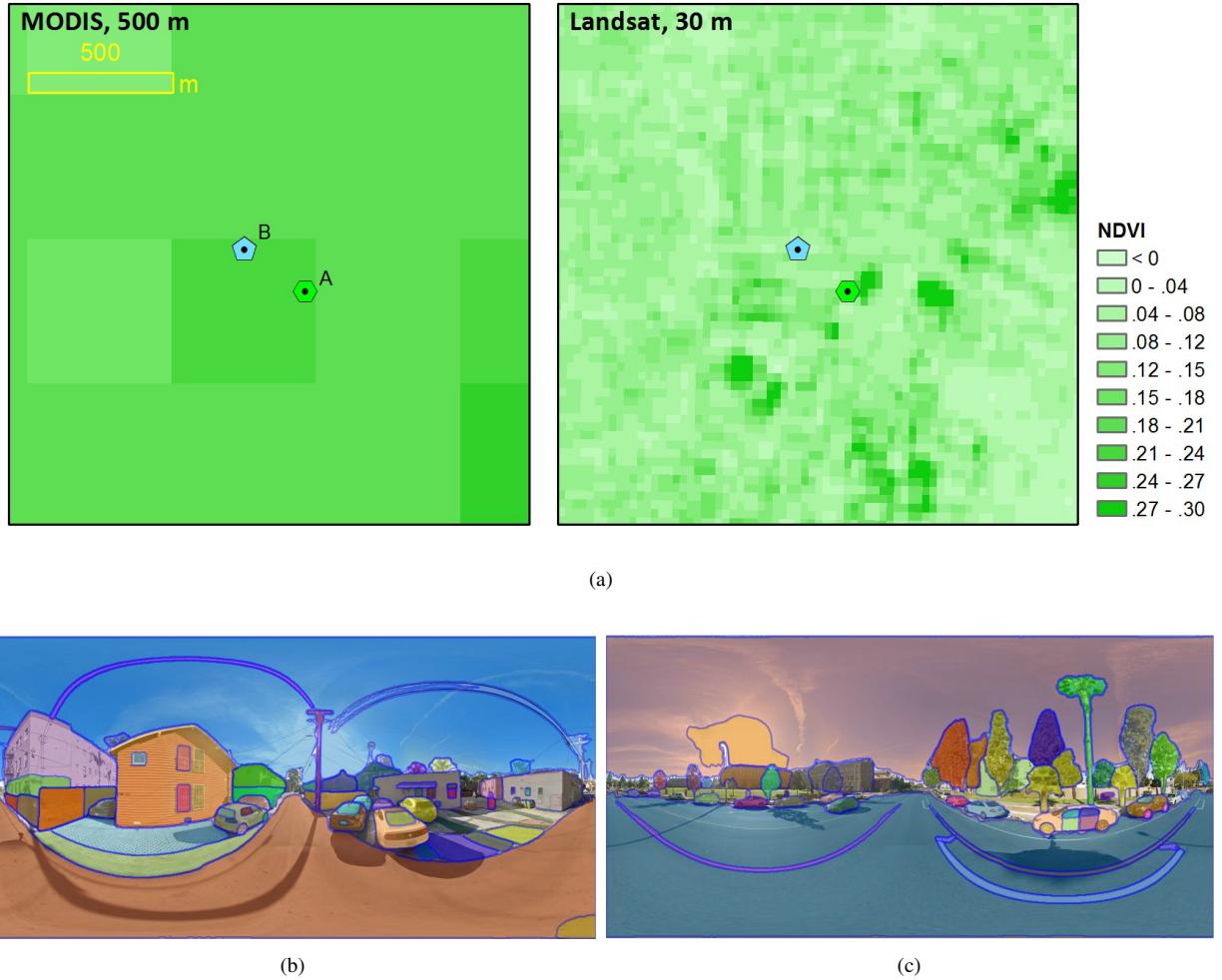


Figure 3: Comparisons of using 500 m satellite data (MODIS), 30 m satellite data (Landsat), and street view imagery to measure green space exposure. (a) Overhead MODIS NDVI at 500 m resolution and Landsat NDVI at 30 m resolution. Hexagon A is an area with high green space exposure, and Pentagon B is an area with low green space exposure. (b) Street view of Pentagon B, with semantic segmentation result showing low green space exposure. (c) Street view of Hexagon A, with semantic segmentation result showing relatively high green space exposure. The street view image segmentation results were generated using the Segment Anything model hosted by Meta AI (Meta AI, 2024)

409 5.3.1. *Effective range of green space: size of buffer zones*

410 The size of buffer zones used to measure green space exposure varied across the reviewed stud-
411 ies, ranging from 100 to 2,000 m. The two most common buffer sizes—500 and 1,000 m—were
412 largely chosen for convenience. A few studies justified the usage of 1,500 or 1,600 m (one mile)
413 buffer zones because this range is teh walking distance within 15 minutes (Astell-Burt et al., 2023;
414 Wang et al., 2023), corresponding to the concept of 15-minute cities (Bruno et al., 2024). The
415 effective range of green space depends on the underlying mechanisms. If functional effects dom-
416 inate, green space needs to be close enough to alter biochemical components of the atmosphere;
417 this range is likely small and within 1 km. If spatial effects dominate, the 15-minute city frame-
418 work becomes relevant, suggesting an effective range of 1.5–2 km. Despite these considerations,
419 existing studies often relied on pre-set values, limiting our understanding of the true effective
420 range.

421 5.4. *Distinguishing varying forms of urban green space*

422 Most studies did not distinguish between different forms of green space, as they relied on
423 NDVI at 30 m or even 500 m resolution. However, distinguishing various forms of urban green
424 space can be achieved in at least three ways.

425 5.4.1. *High-resolution land use land cover data*

426 First, one approach is to utilize existing well-maintained high-resolution land use data, such as
427 the ib1000 dataset with 1 m spatial resolution in Hong Kong (Wang et al., 2021), the Geovision
428 Product with 2 m resolution in Australia (Astell-Burt and Feng, 2019), or the Urban Atlas Land
429 Use dataset with 5 m resolution in Europe (Kolcsár et al., 2021). Beyond regional land cover
430 data, global land use and land cover datasets with resolutions better than 10 m offer various green
431 vegetation categories, including the Esri 2020 Land Cover product (10 m resolution), the ESA
432 WorldCover product (10 m resolution), and the OSM Land Cover dataset (10 m resolution) (Venter
433 et al., 2022). All these products provide a more accurate estimation of green space forms compared
434 to NDVI at 500 m resolution.

435 5.4.2. *Finer-grained classification using local climate zones*

436 Second, scene classification can be used for fine-grained classification of green space, e.g.,
437 the local climate zone classification system (Figure 4). Instead of relying solely on spectral in-
438 formation (greenness), which performs poorly in distinguishing high trees from low bushes, scene
439 classification schemes provide a more detailed categorization of green space by analyzing global
440 image features, such as tree canopy shading. As shown in Figure 4, natural environments can be
441 categorized into seven classes: dense trees, scattered trees, bush or scrub, low plants, bare rock or
442 paved, bare soil or sand, and water. Their combination with low- and high-rise built environments
443 likely influences air pollution reduction and physical activity. For example, cities and populated
444 regions tend to have higher air pollution (Zhang et al., 2020), while high-rise built environments,
445 often accompanied by public transportation, generally promote more walking (Althoff et al., 2017).
446 Leveraging data from this comprehensive framework enables a more precise characterization of
447 different forms of green space within the built environment, thereby improving our understanding
448 of the relationship between green space and cognition.

Local Climate Zones

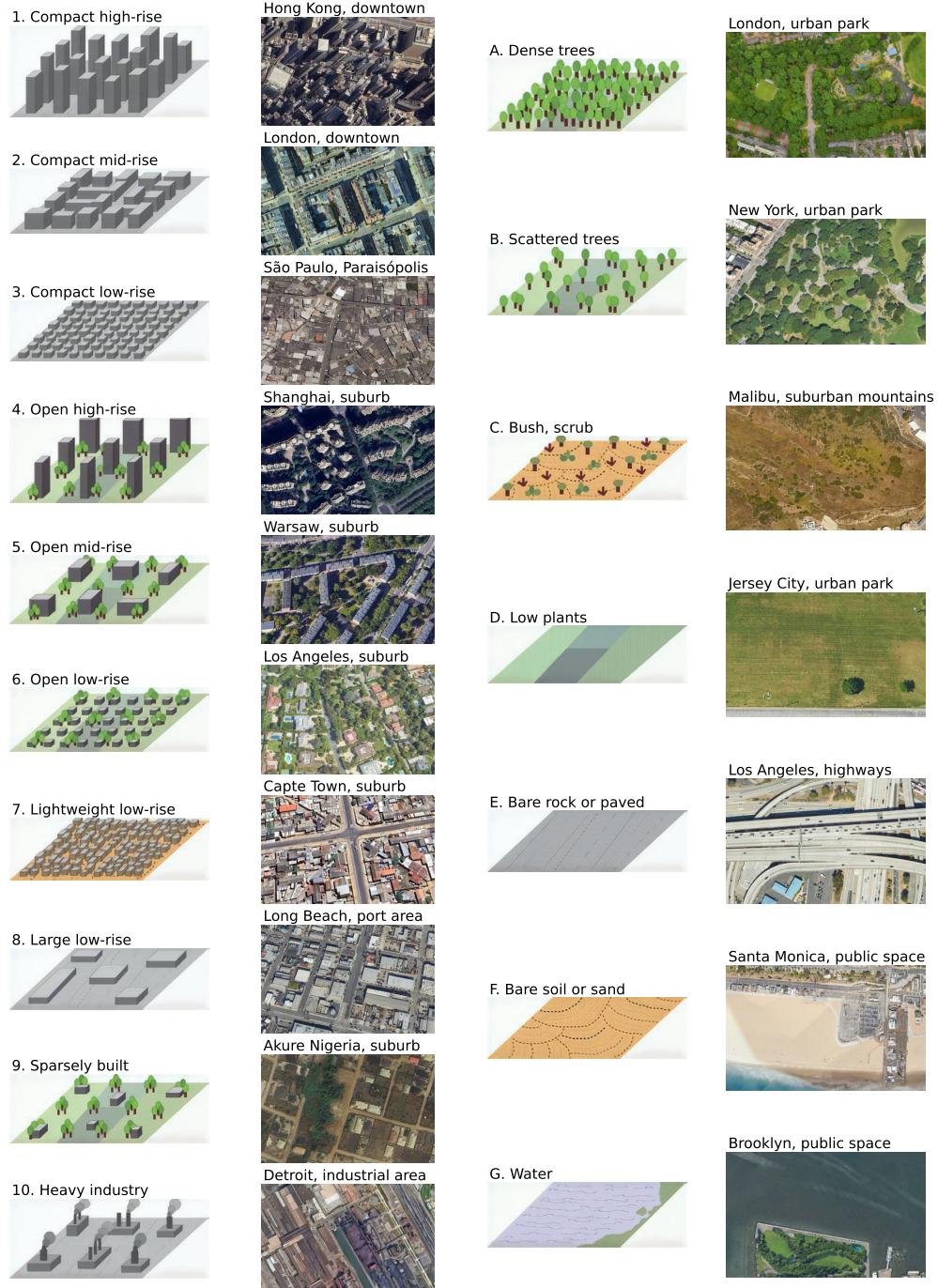


Figure 4: Urban landscape categorization using the local climate zone system. The conceptual models representing local climate zones are based on Stewart and Oke (2012), and the satellite images are from Google Earth with the highlighted cities selected by the authors.

449 *5.4.3. Street view imagery*

450 Third, the use of street view imagery. As mentioned, instead of overhead NDVI, horizontal-
451 view (eye-level) measures of green space provide a more accurate representation of urban forests
452 and open grass areas. Eye-level imagery also better reflects people's daily perception of green
453 space (Larkin and Hystad, 2019; Kang et al., 2020; Xiao et al., 2021).

454 *5.5. Mechanisms from green space to cognitive functions*

455 There are ten articles directly investigating air pollution as a potential mediator in their mod-
456eling, one article demonstrated that green space is a mediator of urban heat on hospitalization and
457 deaths related to Alzheimer's disease, and three articles provided direct evidence that simply per-
458 ceiving green space is enough to bring benefits to the brain. However, some studies suggest a null
459 association between physical activity, air pollution, and noise. More research is needed to explore
460 these mechanisms, particularly with improved categorization of green space.

461 *5.5.1. Limited studies on social interaction*

462 No study has analyzed the potential effects of social interactions on cognitive functions. The
463 reason may be due to the difficulty of capturing social interactions compared to physical properties.
464 However, there may be some viable proxies. In an Irish study (Dempsey et al., 2018), a U-shaped
465 relationship was found between surrounding green space and obesity: both low and high levels
466 of green space were associated with a higher probability of obesity. Although the authors did not
467 provide a clear explanation, one possible theory is that people living in areas with abundant green
468 space tend to reside in remote locations with low population density, where social interaction
469 is limited. If green spaces were categorized based on their surrounding population density—a
470 potential proxy for social interaction—studies on this topic could become feasible.

471 **6. Conclusions and future directions**

472 This scoping review identified 37 studies from the Web of Science examining the association
473 between green space and cognitive function. Among these, 20 studies directly investigated the
474 pathways via modeling, while 17 studies provided indirect evidence by comparing different types
475 of green space. While findings support all three pathways, the functional effect of reducing air
476 pollution is the most commonly studied, followed by the spatial effect of promoting physical
477 activity. Regarding perceptual effects, three case-control studies provided supporting evidence.
478 However, no studies have explored other potential pathways, including heat and social interaction.

479 A key limitation identified in most studies is the coarse measurement of green space. We high-
480 lighted MODIS, Landsat, and street view imagery as three levels of green space measurement,
481 emphasizing the need for more accurate green space measurement for cognition studies. Future
482 studies should refine their definition of green space and adopt appropriate measurement techniques
483 to better understand the functional, spatial, and perceptual pathways linking green space to cogni-
484 tive function. Incorporating local climate zones as a systematic framework can help differentiate
485 green space types, which is essential for elucidating the underlying mechanisms.

486 **Acknowledgements**

487 S.L. was supported by a USC Dana and David Dornsife College of Letters, Arts and Sci-
488 ences/Graduate School Fellowship.

489 **References**

- 490 Abhijith, K., Kumar, P., Gallagher, J., McNabola, A., Baldauf, R., Pilla, F., Broderick, B., Di Sabatino, S., Pulvirenti,
491 B., 2017. Air pollution abatement performances of green infrastructure in open road and built-up street canyon
492 environments—a review. *Atmospheric Environment* 162, 71–86.
- 493 Ahmed, S.M., Knibbs, L.D., Moss, K.M., Mouly, T.A., Yang, I.A., Mishra, G.D., 2022. Residential greenspace and
494 early childhood development and academic performance: A longitudinal analysis of australian children aged 4–12
495 years. *Science of The Total Environment* 833, 155214.
- 496 Almeida, D.Q., Barros, H., Ribeiro, A.I., 2022. Residential and school green and blue spaces and intelligence in
497 children: The generation xxi birth cohort. *Science of The Total Environment* 813, 151859.
- 498 Althoff, T., Sosić, R., Hicks, J.L., King, A.C., Delp, S.L., Leskovec, J., 2017. Large-scale physical activity data reveal
499 worldwide activity inequality. *Nature* 547, 336–339.
- 500 Asta, F., Michelozzi, P., Cesaroni, G., De Sario, M., Davoli, M., Porta, D., 2021. Green spaces and cognitive develop-
501 ment at age 7 years in a rome birth cohort: The mediating role of nitrogen dioxide. *Environmental Research* 196,
502 110358.
- 503 Astell-Burt, T., Feng, X., 2019. Association of urban green space with mental health and general health among adults
504 in australia. *JAMA network open* 2, e198209–e198209.
- 505 Astell-Burt, T., Feng, X., 2020. Greener neighbourhoods, better memory? a longitudinal study. *Health & Place* 65,
506 102393.
- 507 Astell-Burt, T., Navakatikyan, M.A., Feng, X., 2020. Urban green space, tree canopy and 11-year risk of dementia in
508 a cohort of 109,688 australians. *Environment International* 145, 106102.
- 509 Astell-Burt, T., Navakatikyan, M.A., Feng, X., 2023. Why might urban tree canopy reduce dementia risk? a causal
510 mediation analysis of 109,688 adults with 11 years of hospital and mortality records. *Health & Place* 82, 103028.
- 511 Baena-Extremera, A., Martín-Pérez, C., Catena, A., Fuentesal-García, J., 2024. Green exercise versus indoor urban
512 exercise: Related frontal brain thickness and cognitive performance. *Mental Health and Physical Activity* 27,
513 100649.
- 514 Barboza, E.P., Cirach, M., Khomenko, S., Iungman, T., Mueller, N., Barrera-Gómez, J., Rojas-Rueda, D., Kondo, M.,
515 Nieuwenhuijsen, M., 2021. Green space and mortality in european cities: a health impact assessment study. *The Lancet Planetary Health* 5, e718–e730.
- 516 Besser, L.M., Edwards, K., Lobban, N.S., Tolea, M.I., Galvin, J.E., 2024. Social determinants of health, risk and
517 resilience against alzheimer's disease and related dementias: The healthy brain initiative. *Journal of Alzheimer's Disease Reports* 8, 637–646.
- 518 Bijnens, E.M., Vos, S., Verheyen, V.V., Bruckers, L., Covaci, A., De Henauw, S., Den Hond, E., Loots, I., Nelen, V.,
519 Plusquin, M., et al., 2022. Higher surrounding green space is associated with better attention in flemish adolescents.
520 *Environment International* 159, 107016.
- 521 Binter, A.C., Bernard, J.Y., Mon-Williams, M., Andiarena, A., González-Safont, L., Vafeiadi, M., Lepeule, J., Soler-
522 Blasco, R., Alonso, L., Kampouri, M., et al., 2022. Urban environment and cognitive and motor function in children
523 from four european birth cohorts. *Environment International* 158, 106933.
- 524 Binter, A.C., Granés, L., Bannier, E., de Castro, M., Petricola, S., Fossati, S., Vrijheid, M., Chevrier, C., El Marroun,
525 H., Nieuwenhuijsen, M., et al., 2024. Urban environment during pregnancy and childhood and white matter
526 microstructure in preadolescence in two european birth cohorts. *Environmental Pollution* 346, 123612.
- 527 Bratman, G.N., Hamilton, J.P., Daily, G.C., 2012. The impacts of nature experience on human cognitive function and
528 mental health. *Annals of the New York Academy of Sciences* 1249, 118–136.
- 529 Brown, S., Aitken, W., Lombard, J., Parrish, A., Dewald, J., Ma, R., Messinger, S., Liu, S., Nardi, M., Rundek, T.,
530 et al., 2024. Longitudinal impacts of precision greenness on alzheimer's disease. *The Journal of Prevention of*
531 *Alzheimer's Disease*, 1–11.

- 534 Bruno, M., Monteiro Melo, H.P., Campanelli, B., Loreto, V., 2024. A universal framework for inclusive 15-minute
 535 cities. *Nature Cities* , 1–9.
- 536 Buczyłowska, D., Singh, N., Baumbach, C., Bratkowski, J., Mysak, Y., Wierzba-Łukaszyk, M., Sitnik-Warchulska,
 537 K., Skotak, K., Lipowska, M., Izydorczyk, B., et al., 2024. Lifelong greenspace exposure and adhd in polish chil-
 538 dren: Role of physical activity and perceived neighbourhood characteristics. *Journal of Environmental Psychology*
 539 96, 102313.
- 540 Cerin, E., Soloveva, M.V., Molina, M.A., Schroers, R.D., Knibbs, L.D., Akram, M., Wu, Y.T., Mavoa, S., Prina, M.,
 541 Sachdev, P.S., et al., 2024. Neighbourhood environments and cognitive health in the longitudinal personality and
 542 total health (path) through life study: A 12-year follow-up of older Australians. *Environment International* 191,
 543 108984.
- 544 Chang, D.H., Jiang, B., Wong, N.H., Wong, J.J., Webster, C., Lee, T.M., 2021. The human posterior cingulate and the
 545 stress-response benefits of viewing green urban landscapes. *NeuroImage* 226, 117555.
- 546 Chen, H., Kwong, J.C., Copes, R., Hystad, P., van Donkelaar, A., Tu, K., Brook, J.R., Goldberg, M.S., Martin, R.V.,
 547 Murray, B.J., et al., 2017. Exposure to ambient air pollution and the incidence of dementia: a population-based
 548 cohort study. *Environment International* 108, 271–277.
- 549 Cherrie, M.P., Shortt, N.K., Ward Thompson, C., Deary, I.J., Pearce, J.R., 2019. Association between the activity
 550 space exposure to parks in childhood and adolescence and cognitive aging in later life. *International Journal of
 551 Environmental Research and Public Health* 16, 632.
- 552 Cheung, P.K., Nice, K.A., Livesley, S.J., 2022. Irrigating urban green space for cooling benefits: the mechanisms and
 553 management considerations. *Environmental Research: Climate* 1, 015001.
- 554 Claesen, J.L., Wheeler, A.J., Klabbers, G., Gonzalez, D.D., Molina, M.A., Tham, R., Nieuwenhuijsen, M., Carver, A.,
 555 2021. Associations of traffic-related air pollution and greenery with academic outcomes among primary schoolchil-
 556 dren. *Environmental Research* 199, 111325.
- 557 Crous-Bou, M., Gascon, M., Gispert, J.D., Cirach, M., Sánchez-Benavides, G., Falcon, C., Arenaza-Urquijo, E.M.,
 558 Gotsens, X., Fauria, K., Sunyer, J., et al., 2020. Impact of urban environmental exposures on cognitive performance
 559 and brain structure of healthy individuals at risk for alzheimer's dementia. *Environment International* 138, 105546.
- 560 Dadvand, P., Nieuwenhuijsen, M.J., Esnaola, M., Forns, J., Basagaña, X., Alvarez-Pedrerol, M., Rivas, I., López-
 561 Vicente, M., Pascual, M.D.C., Su, J., et al., 2015. Green spaces and cognitive development in primary schoolchil-
 562 dren. *Proceedings of the National Academy of Sciences* 112, 7937–7942.
- 563 Dadvand, P., Tischer, C., Estarlich, M., Llop, S., Dalmau-Bueno, A., López-Vicente, M., Valentín, A., de Keijzer, C.,
 564 Fernández-Somoano, A., Lertxundi, N., et al., 2017. Lifelong residential exposure to green space and attention: a
 565 population-based prospective study. *Environmental Health Perspectives* 125, 097016.
- 566 Dempsey, S., Lyons, S., Nolan, A., 2018. Urban green space and obesity in older adults: Evidence from ireland.
 567 *SSM - Population Health* 4, 206–215. URL: <https://www.sciencedirect.com/science/article/pii/S2352827317302203>, doi:<https://doi.org/10.1016/j.ssmph.2018.01.002>.
- 569 Diener, A., Mudu, P., 2021. How can vegetation protect us from air pollution? a critical review on green spaces'
 570 mitigation abilities for air-borne particles from a public health perspective-with implications for urban planning.
 571 *Science of the Total Environment* 796, 148605.
- 572 Dzhambov, A.M., Bahchevanov, K.M., Chompalov, K.A., Atanassova, P.A., 2019. A feasibility study on the associa-
 573 tion between residential greenness and neurocognitive function in middle-aged bulgarians. *Arhiv za higijenu rada
 574 i toksikologiju* 70, 173–184.
- 575 Dzhambov, A.M., Dimitrova, D.D., 2014. Urban green spaces' effectiveness as a psychological buffer for the negative
 576 health impact of noise pollution: A systematic review. *Noise and health* 16, 157–165.
- 577 Falcón, C., Gascon, M., Molinuevo, J.L., Operto, G., Cirach, M., Gotsens, X., Fauria, K., Arenaza-Urquijo, E.M.,
 578 Pujol, J., Sunyer, J., et al., 2021. Brain correlates of urban environmental exposures in cognitively unimpaired
 579 individuals at increased risk for alzheimer's disease: A study on barcelona's population. *Alzheimer's & Dementia:
 580 Diagnosis, Assessment & Disease Monitoring* 13, e12205.
- 581 Fan, J., Guo, Y., Cao, Z., Cong, S., Wang, N., Lin, H., Wang, C., Bao, H., Lv, X., Wang, B., et al., 2020. Neighborhood
 582 greenness associated with chronic obstructive pulmonary disease: a nationwide cross-sectional study in china.
 583 *Environment International* 144, 106042.
- 584 Fullerton, K.R., 2011. Street performers and the sense of place: A case study of Third Street Promenade Shopping

- 585 Center, Santa Monica, California. California State University, Long Beach.
- 586 Gaffin, S., Rosenzweig, C., Khanbilvardi, R., Parshall, L., Mahani, S., Glickman, H., Goldberg, R., Blake, R., Slos-
587 berg, R., Hillel, D., 2008. Variations in new york city's urban heat island strength over time and space. *Theoretical
588 and applied climatology* 94, 1–11.
- 589 Godina, S.L., Rosso, A.L., Hirsch, J.A., Besser, L.M., Lovasi, G.S., Donovan, G.H., Garg, P.K., Platt, J.M., Fitz-
590 patrick, A.L., Lopez, O.L., et al., 2023. Neighborhood greenspace and cognition: The cardiovascular health study.
591 *Health & Place* 79, 102960.
- 592 Grahn, P., Stigsdotter, U.K., 2010. The relation between perceived sensory dimensions of urban green space and stress
593 restoration. *Landscape and urban planning* 94, 264–275.
- 594 Ha, J., Kim, H.J., With, K.A., 2022. Urban green space alone is not enough: A landscape analysis linking the spatial
595 distribution of urban green space to mental health in the city of chicago. *Landscape and Urban Planning* 218,
596 104309.
- 597 Hartig, T., Mitchell, R., De Vries, S., Frumkin, H., 2014. Nature and health. *Annual Review of Public Health* 35,
598 207–228.
- 599 Hopkins, R.S., 2015. *Planning the Greenspaces of Nineteenth-Century Paris*. LSU Press.
- 600 Hu, H.Y., Ma, Y.H., Deng, Y.T., Ou, Y.N., Cheng, W., Feng, J.F., Tan, L., Yu, J.T., 2023. Residential greenness and
601 risk of incident dementia: A prospective study of 375,342 participants. *Environmental Research* 216, 114703.
- 602 Hu, X., Wang, J., Yang, T., Jin, J., Zeng, Q., Aboubakri, O., Feng, X.L., Li, G., Huang, J., 2024. Role of residential
603 greenspace in the trajectory of major neurological disorders: A longitudinal study in uk biobank. *Science of The
604 Total Environment* 912, 168967.
- 605 Hunter, R.F., Nieuwenhuijsen, M., Fabian, C., Murphy, N., O'Hara, K., Rappe, E., Sallis, J.F., Lambert, E.V., Duenas,
606 O.L.S., Sugiyama, T., et al., 2023. Advancing urban green and blue space contributions to public health. *The
607 Lancet Public Health* 8, e735–e742.
- 608 Jafari, M.J., Khosrowabadi, R., Khodakarim, S., Mohammadian, F., 2019. The effect of noise exposure on cognitive
609 performance and brain activity patterns. *Open access Macedonian journal of medical sciences* 7, 2924.
- 610 Jarvis, I., Sbihi, H., Davis, Z., Brauer, M., Czekajlo, A., Davies, H.W., Gergel, S.E., Guhn, M., Jerrett, M., Koehoorn,
611 M., et al., 2022. The influence of early-life residential exposure to different vegetation types and paved surfaces on
612 early childhood development: A population-based birth cohort study. *Environment International* 163, 107196.
- 613 Jin, X., Shu, C., Zeng, Y., Liang, L., Ji, J.S., 2021. Interaction of greenness and polygenic risk score of alzheimer's
614 disease on risk of cognitive impairment. *Science of The Total Environment* 796, 148767.
- 615 John, E.E., Astell-Burt, T., Yu, P., Brennan-Horley, C., Feng, X., 2023. Green space type and healthy ageing in place:
616 An australian longitudinal study. *Urban Forestry & Urban Greening* 84, 127903.
- 617 Kang, Y., Zhang, F., Gao, S., Lin, H., Liu, Y., 2020. A review of urban physical environment sensing using street view
618 imagery in public health studies. *Annals of GIS* 26, 261–275.
- 619 de Keijzer, C., Tonne, C., Basagaña, X., Valentín, A., Singh-Manoux, A., Alonso, J., Antó, J.M., Nieuwenhuijsen,
620 M.J., Sunyer, J., Dadvand, P., 2018. Residential surrounding greenness and cognitive decline: a 10-year follow-up
621 of the whitehall ii cohort. *Environmental Health Perspectives* 126, 077003.
- 622 Kirillov, A., Mintun, E., Ravi, N., Mao, H., Rolland, C., Gustafson, L., Xiao, T., Whitehead, S., Berg, A.C., Lo, W.Y.,
623 et al., 2023. Segment anything. *arXiv preprint arXiv:2304.02643*.
- 624 Klompmaker, J.O., Janssen, N.A., Bloemsma, L.D., Marra, M., Lebret, E., Gehring, U., Hoek, G., 2021. Effects of
625 exposure to surrounding green, air pollution and traffic noise with non-accidental and cause-specific mortality in
626 the dutch national cohort. *Environmental Health* 20, 1–16.
- 627 Kohlhase, J.E., 2013. The new urban world 2050: perspectives, prospects and problems. *Regional Science Policy &
628 Practice* 5, 153–166.
- 629 Kolcsár, R.A., Csikós, N., Szilassi, P., 2021. Testing the limitations of buffer zones and urban atlas population data in
630 urban green space provision analyses through the case study of szeged, hungary. *Urban forestry & urban greening*
631 57, 126942.
- 632 Kothencz, G., Kolcsár, R., Cabrera-Barona, P., Szilassi, P., 2017. Urban green space perception and its contribution
633 to well-being. *International journal of environmental research and public health* 14, 766.
- 634 Lachowycz, K., Jones, A.P., 2013. Towards a better understanding of the relationship between greenspace and health:
635 Development of a theoretical framework. *Landscape and Urban Planning* 118, 62–69.

- 636 Larkin, A., Hystad, P., 2019. Evaluating street view exposure measures of visible green space for health research.
637 *Journal of Exposure Science & Environmental Epidemiology* 29, 447–456.
- 638 Lee, A.C., Maheswaran, R., 2011. The health benefits of urban green spaces: a review of the evidence. *Journal of
639 Public Health* 33, 212–222.
- 640 Liao, J., Zhang, B., Xia, W., Cao, Z., Zhang, Y., Liang, S., Hu, K., Xu, S., Li, Y., 2019. Residential exposure to green
641 space and early childhood neurodevelopment. *Environment International* 128, 70–76.
- 642 Liu, X.X., Ma, X.L., Huang, W.Z., Luo, Y.N., He, C.J., Zhong, X.M., Dadvand, P., Browning, M.H., Li, L., Zou,
643 X.G., et al., 2022. Green space and cardiovascular disease: a systematic review with meta-analysis. *Environmental
644 Pollution* 301, 118990.
- 645 Maas, J., Verheij, R.A., Groenewegen, P.P., De Vries, S., Spreeuwenberg, P., 2006. Green space, urbanity, and health:
646 how strong is the relation? *Journal of Epidemiology & Community Health* 60, 587–592.
- 647 Maes, M.J., Pirani, M., Booth, E.R., Shen, C., Milligan, B., Jones, K.E., Toledano, M.B., 2021. Benefit of woodland
648 and other natural environments for adolescents' cognition and mental health. *Nature Sustainability* 4, 851–858.
- 649 Markevych, I., Feng, X., Astell-Burt, T., Standl, M., Sugiri, D., Schikowski, T., Koletzko, S., Herberth, G., Bauer,
650 C.P., von Berg, A., et al., 2019. Residential and school greenspace and academic performance: Evidence from the
651 giniplus and lisa longitudinal studies of german adolescents. *Environmental Pollution* 245, 71–76.
- 652 Mason, L., Manzione, L., Ronconi, A., Pazzaglia, F., 2022. Lessons in a green school environment and in the
653 classroom: Effects on students' cognitive functioning and affect. *International Journal of Environmental Research
654 and Public Health* 19, 16823.
- 655 Meta AI, 2024. Segment anything ai model. <https://segment-anything.com/>. Accessed: 2024-10-15.
- 656 Mueller, N., Rojas-Rueda, D., Khreis, H., Cirach, M., Andrés, D., Ballester, J., Bartoll, X., Daher, C., Deluca, A.,
657 Echave, C., et al., 2020. Changing the urban design of cities for health: The superblock model. *Environment
658 international* 134, 105132.
- 659 Mueller, W., Milner, J., Loh, M., Vardoulakis, S., Wilkinson, P., 2022. Exposure to urban greenspace and pathways
660 to respiratory health: An exploratory systematic review. *Science of the Total Environment* 829, 154447.
- 661 Panduro, T.E., Veie, K.L., 2013. Classification and valuation of urban green spaces—a hedonic house price valuation.
662 *Landscape and Urban planning* 120, 119–128.
- 663 Paul, L.A., Hystad, P., Burnett, R.T., Kwong, J.C., Crouse, D.L., van Donkelaar, A., Tu, K., Lavigne, E., Copes, R.,
664 Martin, R.V., et al., 2020. Urban green space and the risks of dementia and stroke. *Environmental Research* 186,
665 109520.
- 666 Pryor, M., 2022. Everyday urbanism in high-density cities, in: *The Routledge Handbook of Sustainable Cities and
667 Landscapes in the Pacific Rim*. Routledge, pp. 645–656.
- 668 Rodriguez-Loureiro, L., Gadeyne, S., Bauwelinck, M., Lefebvre, W., Vanpoucke, C., Casas, L., 2022. Long-term
669 exposure to residential greenness and neurodegenerative disease mortality among older adults: a 13-year follow-up
670 cohort study. *Environmental Health* 21, 49.
- 671 Ruiz-Gonzalez, D., Hernandez-Martinez, A., Valenzuela, P.L., Morales, J.S., Soriano-Maldonado, A., 2021. Effects of
672 physical exercise on plasma brain-derived neurotrophic factor in neurodegenerative disorders: a systematic review
673 and meta-analysis of randomized controlled trials. *Neuroscience & Biobehavioral Reviews* 128, 394–405.
- 674 Sit, K.Y., Chen, W.Y., Ng, K.Y., Koh, K., Zhang, H., 2025. Unveiling environmental inequalities in high-density asian
675 city: City-scaled comparative analysis of green space coverage within 10-minute walk from private, public, and
676 rural housing. *Landscape and Urban Planning* 253, 105225.
- 677 Stewart, I.D., Oke, T.R., 2012. Local climate zones for urban temperature studies. *Bulletin of the American Meteo-
678 rological Society* 93, 1879–1900.
- 679 Subiza-Pérez, M., García-Baquero, G., Fernández-Somoano, A., Guxens, M., González, L., Tardón, A., Dadvand, P.,
680 Estarlich, M., de Castro, M., McEachan, R.R., et al., 2023. Residential green and blue spaces and working memory
681 in children aged 6–12 years old. results from the inma cohort. *Health & Place* 84, 103136.
- 682 Sugiyama, T., Leslie, E., Giles-Corti, B., Owen, N., 2008. Associations of neighbourhood greenness with physical
683 and mental health: do walking, social coherence and local social interaction explain the relationships? *Journal of
684 Epidemiology & Community Health* 62, e9–e9.
- 685 Tallis, H., Bratman, G.N., Samhouri, J.F., Fargione, J., 2018. Are california elementary school test scores more
686 strongly associated with urban trees than poverty? *Frontiers in Psychology* 9, 2074.

- 687 Thompson, C.W., 2002. Urban open space in the 21st century. *Landscape and urban planning* 60, 59–72.
- 688 Van Renterghem, T., Botteldooren, D., Verheyen, K., 2012. Road traffic noise shielding by vegetation belts of limited
689 depth. *Journal of Sound and Vibration* 331, 2404–2425.
- 690 Venter, Z.S., Barton, D.N., Chakraborty, T., Simensen, T., Singh, G., 2022. Global 10 m land use land cover datasets:
691 A comparison of dynamic world, world cover and esri land cover. *Remote Sensing* 14, 4101.
- 692 Wallner, P., Kundi, M., Arnberger, A., Eder, R., Allex, B., Weitensfelder, L., Hutter, H.P., 2018. Reloading pupils'
693 batteries: Impact of green spaces on cognition and wellbeing. *International Journal of Environmental Research
694 and Public Health* 15, 1205.
- 695 Wang, P., Goggins, W.B., Shi, Y., Zhang, X., Ren, C., Lau, K.K.L., 2021. Long-term association between urban air
696 ventilation and mortality in hong kong. *Environmental Research* 197, 111000.
- 697 Wang, Y., Crowe, M., Knibbs, L.D., Fuller-Tyszkiewicz, M., Mygind, L., Kerr, J.A., Wake, M., Olsson, C.A., Enticott,
698 P.G., Peters, R.L., et al., 2023. Greenness modifies the association between ambient air pollution and cognitive
699 function in australian adolescents, but not in mid-life adults. *Environmental Pollution* 324, 121329.
- 700 Watkins, M.W., Kush, J.C., Glutting, J.J., 1997. Discriminant and predictive validity of the wisc-iii acid profile among
701 children with learning disabilities. *Psychology in the Schools* 34, 309–319.
- 702 WHO, 2020. The top 10 causes of death. World Health Organization URL: <https://www.who.int/news-room/fact-sheets/detail/the-top-10-causes-of-death>. accessed: 2021-12-01.
- 703 Wu, C.D., McNeely, E., Cedeño-Laurent, J., Pan, W.C., Adamkiewicz, G., Dominici, F., Lung, S.C.C., Su, H.J.,
704 Spengler, J.D., 2014. Linking student performance in massachusetts elementary schools with the “greeness” of
705 school surroundings using remote sensing. *PloS One* 9, e108548.
- 706 Wu, Y.T., Brayne, C., Liu, Z., Huang, Y., Sosa, A.L., Acosta, D., Prina, M., 2020. Neighbourhood environment
707 and dementia in older people from high-, middle- And low-income countries: Results from two population-based
708 cohort studies. *BMC Public Health* 20. doi:10.1186/s12889-020-09435-5.
- 709 Xiao, X., Wang, R., Knibbs, L.D., Jalaludin, B., Heinrich, J., Markevych, I., Gao, M., Xu, S.L., Wu, Q.Z., Zeng,
710 X.W., et al., 2021. Street view greenness is associated with lower risk of obesity in adults: Findings from the 33
711 chinese community health study. *Environmental Research* , 111434.
- 712 Xu, Z., Tong, S., Cheng, J., Zhang, Y., Wang, N., Zhang, Y., Hayixibayi, A., Hu, W., 2019. Heatwaves, hospitalizations
713 for alzheimer’s disease, and postdischarge deaths: A population-based cohort study. *Environmental Research* 178,
714 108714.
- 715 Yang, T., Gu, T., Xu, Z., He, T., Li, G., Huang, J., 2023. Associations of residential green space with incident type 2
716 diabetes and the role of air pollution: A prospective analysis in uk biobank. *Science of the Total Environment* 866,
717 161396.
- 718 Yu, R., Wang, D., Leung, J., Lau, K., Kwok, T., Woo, J., 2018. Is Neighborhood Green Space Associated With Less
719 Frailty? Evidence From the Mr. and Ms. Os (Hong Kong) Study. *Journal of the American Medical Directors
720 Association* 19, 528–534. doi:10.1016/j.jamda.2017.12.015.
- 721 Yuchi, W., Sbihi, H., Davies, H., Tamburic, L., Brauer, M., 2020. Road proximity, air pollution, noise, green space and
722 neurologic disease incidence: A population-based cohort study. *Environmental Health: A Global Access Science
723 Source* 19. doi:10.1186/s12940-020-0565-4.
- 724 Zautra, A.J., Hall, J.S., Murray, K.E., 2010. A new definition of health for people and communities. *Handbook of
725 adult resilience* 1.
- 726 Zhang, L., Luo, Y., Zhang, Y., Pan, X., Zhao, D., Wang, Q., 2022. Green space, air pollution, weather, and cognitive
727 function in middle and old age in china. *Frontiers in Public Health* 10.
- 728 Zhang, L., Wilson, J.P., MacDonald, B., Zhang, W., Yu, T., 2020. The changing pm2. 5 dynamics of global megacities
729 based on long-term remotely sensed observations. *Environment International* 142, 105862.
- 730 Zheng, Y., Ren, C., Shi, Y., Yim, S.H., Lai, D.Y., Xu, Y., Fang, C., Li, W., 2023. Mapping the spatial distribution of
731 nocturnal urban heat island based on local climate zone framework. *Building and Environment* 234, 110197.
- 732 Zhu, A., Yan, L., Shu, C., Zeng, Y., Ji, J.S., 2020. Apoe ε4 modifies effect of residential greenness on cognitive
733 function among older adults: a longitudinal analysis in china. *Scientific reports* 10, 1–8.
- 734 Zhu, Z., Yang, Z., Yu, L., Xu, L., Wu, Y., Zhang, X., Shen, P., Lin, H., Shui, L., Tang, M., et al., 2023. Residential
735 greenness, air pollution and incident neurodegenerative disease: A cohort study in china. *Science of The Total
736 Environment* 878, 163173.

738 Zou, H., Wang, X., 2021. Progress and gaps in research on urban green space morphology: A review. Sustainability
739 13, 1202.