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# **Deciphering the stroke–built environment nexus in transitional cities: conceptual framework, empirical evidence, and implications for proactive planning intervention**

## **Abstract**

Adverse lifestyle-associated health outcomes, and stroke in particular, have been aggravated in transitional countries under high-speed urbanisation. Against this backdrop, deciphering the nexus between built environments (BEs) and lifestyle-associated health outcomes is of importance for crafting proactive interventions. The existing literature on this topic, however, fails to sufficiently capture the multiplicity of health-related BEs and, in turn, the complexity of such a nexus, largely challenging the applicability of established frameworks and the reliability of relevant findings.

Looking at the case of stroke in Wuhan, China, this research aims to flesh out the understanding of the nexus between multidimensional BEs and lifestyle-associated health outcomes in transitional cities, with regards to conceptual framework and empirical evidence. To this end, we clarified stroke-related BE elements and integrated them into one conceptual framework. We then visualised stroke risk and examined its BE determinants using the Bayesian conditional autoregressive model. The visualisation results showed that stroke risks exhibited significant clustering in the high-density urban core. The statistical analysis found that, after the data were controlled for sociodemographic characteristics, net population density and building density were positively associated with stroke risk. In contrast, an abundance of public parks and institutional land use and access to medical care facilities have presented negative correlations with stroke risk, regardless of urban density. Our research reveals that compact urban developments might not be a silver bullet for health promotion in transitional cities, calling for an urgent need to scrutinise their applicability. Moreover, providing better access to these identified salubrious resources is crucial for offsetting the adverse effects of increasingly dense urban environments. Furthermore, we argue that the establishment of comprehensive conceptual frameworks that connect BEs and lifestyle-associated health outcomes deserves to be highlighted in further research, planning intervention schemes, and health impact assessment projects.

**Keywords:** Built environment; Stroke; Lifestyle-associated health outcome; Transitional city; China

1 ***1. Introduction***

2 Since 1978 and the establishment of the free market in China, the country has been  
3 experiencing rapid urbanisation ([Gu, Kesteloot, and Cook 2015](#)). Similar to the most  
4 transitional countries (see [Eckert and Kohler \(2014\)](#) and [Cyril, Oldroyd, and Renzaho  
5 \(2013\)](#) for reviews), the accelerating urbanisation in post-reform China has  
6 dismayingly created numerous representative health problems. During this period,  
7 major changes to demographic structures and lifestyles have been catalysed, which in  
8 turn raise nonnegligible health challenges. The ageing population has  
9 disproportionately grown due to decades of sub-replacement fertility and the escalation  
10 of life expectancy ([Xie et al. 2018](#)). In the context of increasing prosperity, people  
11 (urban residents in particular) have been living more sedentary lifestyles ([Monda et al.  
12 2007](#)); the transition of nutrition dietary patterns, characterised by the increased  
13 consumption of processed food, has also become prevalent ([Wilkinson 2004](#)).

14 In addition, built environments (BEs), which are defined as manmade elements of the  
15 physical environment, e.g., land-use patterns, transport routes, open spaces, and  
16 buildings ([World Health Organization 2009](#)), have been radically reshaped. The  
17 considerable transformation to BEs in China, e.g., the explosive growth in density  
18 ([Shi et al. 2017](#)), the shrinkage of urban green space ([Ren et al. 2011](#)), and the  
19 homogenisation of land use, has largely contributed to the deteriorated air quality  
20 ([Yuan, Ng, and Norford 2014](#)), aggravated automobile dependency ([Jiang et al. 2016](#)),  
21 and limited leisure-time environments ([Zhang et al. 2014](#)).

22 Against this backdrop, China has seen a remarkable epidemiological transition from a  
23 predominance of infectious diseases to one of adverse lifestyle-associated health  
24 outcomes, such as obesity, cancers, and cardiovascular diseases ([Su et al. 2016](#), [Zhu et  
25 al. 2011](#)). Of the various diseases, stroke — a type of acute cardiovascular disease —  
26 has become China’s biggest killer. More than 1.5 million Chinese residents die from  
27 stroke each year ([Liu et al. 2011](#)), and its prevalence is growing rapidly at an annual  
28 rate of 8% ([NCCD. 2017](#)).

29 It should be noted that these alarming figures represent, to a large extent, the epitome  
30 of the exacerbated burden of stroke and of adverse lifestyle-associated health  
31 outcomes in transitional countries over recent decades. Stroke is the second leading  
32 cause of death worldwide, and three out of four stroke deaths occur in developing

33 countries ([Feigin et al. 2015](#)). From 1990 to 2010, the stroke incidence increased by  
34 12% in low- and middle-income countries, whilst this figure exhibited a significant  
35 downward trend in their higher income counterparts ([Feigin et al. 2014](#)). Within this  
36 context, scholars have speculated that the aggravated stroke burden in transitional  
37 countries is potentially associated with urbanisation (e.g., [Matenga \(1997\)](#); [Lin et al.](#)  
38 [\(2007\)](#); [Truelsen and Bonita \(2008\)](#)). However, based on the existing epidemiological  
39 evidence, they predominately attributed this association to lifestyle transitions, and the  
40 role of BE transformations — an essential step of urbanisation — is overlooked.

41 Insights into risk factors for stroke can be drawn from early epidemiological studies  
42 on demographic characteristics ([NCCD. 2017](#)), lifestyles (e.g., [Larsson, Virtamo, and](#)  
43 [Wolk \(2011\)](#); [Bhat et al. \(2008\)](#); [Patra et al. \(2010\)](#)), and physiological factors (e.g.,  
44 [Bi et al. \(2010\)](#); [Hu and Sun \(2008\)](#); [O'Donnell et al. \(2016\)](#)). In recent years, there  
45 has also been a surge of interest in the association between stroke (and its precursors)  
46 and toxic environmental exposure (e.g., [Yin et al. \(2015\)](#)). However, to the best of our  
47 knowledge, there is no research explicitly examining the association between stroke  
48 and manmade (built) environments. Considering the effects of BEs on the  
49 aforementioned risk factors (e.g., [Wang et al. \(2016\)](#); [Ouyang et al. \(2018\)](#)), it is  
50 reasonable to hypothesise the stroke-BE nexus. Additionally, whilst previous studies  
51 conducted at the similar lifestyle-associated health outcome level (e.g., cardiovascular  
52 diseases, diabetes, and cancers) provide a basic empirical basis for deciphering such a  
53 nexus (e.g., [Su et al. \(2016\)](#); [Chum and O'Campo \(2015\)](#); [Ouyang et al. \(2018\)](#);  
54 [Malambo et al. \(2016\)](#); [Xie et al. \(2018\)](#); [Kan et al. \(2008\)](#)), only limited BE elements  
55 and exclusive/relatively narrow BE dimensions have been considered in each study.  
56 To wit, the *multiplicity* of health-related BEs and the *complexity* of relationships  
57 between BEs and lifestyle-associated health outcomes have not been sufficiently  
58 captured in these studies, which challenges the applicability of established conceptual  
59 frameworks and the reliability of relevant findings.

60 This study aims to better understand the relationship between lifestyle-associated  
61 health outcomes and multidimensional BEs in transitional cities by looking at the  
62 representative case of stroke in Wuhan, central China. To this end, we narrow our  
63 focus to the following questions:

- 64 • Where do strokes occur in transitional cities, and what are the BE

65 characteristics of these areas?

- 66 • To what extent is stroke risk affected by BEs in transitional cities?

67 Our research addresses the above issues in the following ways: (1) it clarifies stroke-  
68 associated multidimensional BE elements and integrates them into one conceptual  
69 framework; (2) it visualises stroke risk in Wuhan using the Bayesian conditional  
70 autoregressive (CAR) method; and (3) it examines the BE determinants of stroke risk  
71 in the focal area. The approach and conceptual framework are applicable to other  
72 cities. The framework and results will contribute to a more comprehensive  
73 understanding of the effects of BE in transitional cities on lifestyle-associated health  
74 outcomes and on stroke in particular.

## 75 **2. Literature review**

76 This section elaborates on two aspects: studies on the nexus between BEs and  
77 lifestyle-associated health outcomes and the mechanism by which stroke is potentially  
78 affected by BEs. The focus of the current research lies within the latter, but a review  
79 of the broader context of the topic helps provide a critical overview of the established  
80 frameworks, elucidate the multiplicity of stroke-related BEs, and discuss the  
81 complexities of the relationship between BEs and lifestyle-associated health outcomes  
82 through the case of stroke.

### 83 **2.1. BEs and lifestyle-associated health outcomes**

84 Along with the increasing interest in healthy urban planning in recent decades, a  
85 growing body of evidence endorses the effects of BEs on lifestyle-associated health  
86 outcomes. One overwhelming stream of such studies has focused on the obesity-BE  
87 nexus (See [Sallis et al. \(2012\)](#) and [Durand et al. \(2011\)](#) for reviews). Given the  
88 relatively self-evident behavioural causes of obesity, this set of studies is primarily  
89 inspired by the accessibility/presence of (1) (in)salubrious food choices in food  
90 environments (e.g., [Ford and Dzewaltowski \(2008\)](#); [Morland, Roux, and Wing \(2006\)](#);  
91 [Anderson et al. \(2011\)](#)) and (2) environments that support physical activity (PA) (e.g.,  
92 [Zhang, Liu, and Liu \(2015\)](#); [Rundle et al. \(2007\)](#); [Wen and Kowaleski-Jones \(2012\)](#)),  
93 which are generally built upon classic travel theories, such as the 3 Ds ([Cervero and](#)  
94 [Kockelman 1997](#)) and 6 Ds frameworks ([Ewing and Cervero 2010](#)).

95 Despite the extensive literature on the obesity-BE nexus, the literature on health

96 outcomes that involve more complex pathogenesis, e.g., cardiovascular diseases,  
97 metabolic syndrome, and cancers, remains relatively sparse. **Table 1** compares the  
98 international literature on this topic. In summary, both the findings and conceptual  
99 frameworks support the need for future works on the effects of BE on lifestyle-  
100 associated health outcomes. However, the insufficient consideration of BEs in terms  
101 of their dimensions and elements has challenged the contributions of these studies.  
102 For example, research by [Ouyang et al. \(2018\)](#), [Xie et al. \(2018\)](#), and [McLafferty and](#)  
103 [Wang \(2009\)](#) has focused on the effects of exclusive BE domains (i.e., land uses, open  
104 spaces, and healthcare facilities, respectively). Although several studies have  
105 incorporated multiple BE dimensions into integrated conceptual frameworks, few  
106 domains constituted by exclusive/limited BE elements have been considered (e.g.,  
107 [Sundquist et al. \(2015\)](#); [Su et al. \(2014\)](#); [Chum and Patricia \(2013\)](#); [Chum and](#)  
108 [O'Campo \(2015\)](#); [Kan et al. \(2008\)](#)). Given the multiple pathways between BEs and  
109 health-related determinants and the (geographical) correlations among health-related  
110 BEs ([Moeller 2013](#)), methodologically, omitting BE elements/dimensions linked to  
111 modifiable health factors might contribute to the endogeneity problem and, in turn,  
112 biased results. More importantly, due to incomprehensiveness, the established  
113 frameworks might not effectively capture the complexity of relationships between  
114 BEs and lifestyle-associated health outcomes, raising questions of their applicability  
115 in further studies, intervention schemes, and health impact assessment projects.

116

**Table 1** Literature of BEs and lifestyle-associated health outcomes.

Study area and sample	Methods	Health outcomes	BE variables	Main findings
700 older adults in Wuhan, China ( <a href="#">Xie et al. 2018</a> )	Logistic regressions	Cardio-cerebral vascular diseases; joint diseases; digestive diseases; endocrine diseases; urological diseases; nervous system diseases; respiratory diseases	Open spaces: accessibility to parks	Negative correlation: accessibility to parks and risks of cardio-cerebral vascular diseases, joint diseases, and endocrine diseases
121 samples of cancer registry areas in the Pan-Yangtze River Delta, China ( <a href="#">Ouyang et al. 2018</a> )	Structural equation modelling (SEM)	Total adjusted cancer; lung cancer; stomach cancer; mammary cancer; liver cancer; colorectal cancer; oesophageal cancer; pancreatic cancer; renal and urinary cancer	Land-use mix: Shannon's Diversity Index (SHDI); landscape shape index (LSI) Land-use type: proportion of built-up area; proportion of open spaces	Positive correlation: LSI and colorectal cancer, lung cancer, mammary cancer, renal and urinary cancer
57 districts in Shenzhen, China ( <a href="#">Su et al. 2016</a> )	Spatial regressions; SEM	Cardiopathy; obesity; hypertension; type 2 diabetes; chronic pneumonia; chronic hepatitis; chronic nephritis; physical fitness; liver cancer; new cancers; thyroid diseases	Land-use abundance; land use; land use; morphology, land-use proximity; land use mix Land-use type: residential; industrial; commercial; institutional; transportation; green; blue; forest; agricultural land; public facilities Walkability: integrated index measured by distances to POIs Street connectivity: intersection density	Negative correlation: proximity to institutional land use and public facilities and cardiopathy, chronic hepatitis, chronic nephritis, chronic pneumonia, liver cancer, new cancers; green land abundance and hypertension, obesity, new cancers, type 2 diabetes, chronic pneumonia; land-use mix and hypertension, physical fitness, obesity, type 2 diabetes, chronic hepatitis; walkability and cardiopathy, obesity, type 2 diabetes, physical fitness, chronic pneumonia; street connectivity and hypertension, physical fitness, liver cancer Positive correlation: industrial land-use morphology and cardiopathy; proximity to industrial land use

and chronic pneumonia

2,411 adults in Toronto, Canada ( <a href="#">Chum and O'Campo 2015</a> )	Multilevel logistic regressions	Cardiovascular diseases	Open spaces: proportion of parkland Food environments: fast food/food store density	Positive correlation: fast food/food store density and myocardial infarction (MI)/any cardiovascular diseases; proportion of parkland and MI
1,626 adults in Toronto Canada ( <a href="#">Chum and Patricia 2013</a> )	Unweighted analysis; time-weighted analysis	Cardiovascular diseases	Open spaces: proportion of parkland Food environments: food store density; presence of fast-food restaurant  Exposure to traffics: whether lives/works within 100-metre buffer of high traffic	Positive correlation: cardiovascular diseases and fast-food density, presence of fast-food restaurant, lives/works within 100-metre buffer of high traffic
512,061 adults in Stockholm, Sweden ( <a href="#">Sundquist et al. 2015</a> )	Multilevel logistic regressions	Type 2 diabetes	Neighbourhood walkability (characterised by residential density, land-use mix, and street connectivity)	Negative correlation: walkability and type 2 diabetes
13,309 middle-age adults in the U.S. ( <a href="#">Kan et al. 2008</a> )	Cox proportional hazards regressions	Coronary heart disease	Traffic exposure: traffic density; distance to major roads	Positive correlation: traffic density and coronary heart disease
671 adults in Cape Town and Mount Frere, South Africa ( <a href="#">Malambo et al. 2016</a> )	Logistic regressions	Hypertension	Perceived land-use mix; perceived land-use access  Perceived street connectivity; perceived infrastructure for walking/cycling	Positive correlation: perceived land-use mix and self-reported hypertension
166,289 cases of cancers in Illinois, the U.S. ( <a href="#">McLafferty and Wang 2009</a> )	Multilevel logistic regressions	Breast cancer; colorectal cancer; lung cancer; prostate cancer	Accessibility to healthcare	Negative correlation: accessibility to healthcare and late-stage cancer risk for breast and lung



## 119 **2.2. Potential stroke-BE nexus**

### 120 **2.2.1. Stroke risk factors**

121 Epidemiological research on stroke risk factors has primarily focused on the  
122 demographic, physiological, and lifestyle (behavioural) domains. It is widely accepted  
123 that one of the most important demographic determinants of stroke is age ([NCCD.](#)  
124 [2017](#)); [Liapis et al. \(2009\)](#) estimated that the risk of stroke increases with each decade  
125 of life.

126 Four physiological determinants (i.e., precursors) of stroke have been identified:  
127 hypertension, cardiac disease, diabetes, and obesity. Approximately half of all strokes  
128 occur in individuals with a history of hypertension ([Bi et al. 2010](#)), and lower  
129 hypertension awareness contributes to greater stroke mortality ([Joffres et al. 2013](#)).  
130 The risk of stroke is also significantly higher among individuals with cardiac disease,  
131 especially atrial fibrillation ([Hu and Sun 2008](#)). Additionally, a global case study of  
132 32 countries has suggested that the occurrence of stroke in people with diabetes and  
133 the highest tertile of the waist-to-hip ratio were 1.36 and 1.44 times higher than in the  
134 control group, respectively ([O'Donnell et al. 2016](#)).

135 Various lifestyles have been connected to stroke in either direct or indirect (i.e., via  
136 the modification of physiological determinants) ways. Evidence has supported the  
137 independently protective effect of moderate-vigorous PA on strokes and its  
138 physiological determinants ([Lee, Folsom, and Blair 2003](#), [Sallis et al. 2012](#)).  
139 Additionally, higher risks of stroke and its precursors are more prevalent among  
140 individuals with unbalanced diets, e.g., diets high in fat, cholesterol, salt, and  
141 processed food ([Larsson, Virtamo, and Wolk 2011](#), [Larsson et al. 2009](#), [Feng,](#)  
142 [Pomborodrigues, and Macgregor 2014](#)). Additional studies have demonstrated that  
143 cigarette smoking and alcohol consumption are closely linked with ischaemic and  
144 haemorrhagic strokes, respectively, with a strong dose-response relationship ([Bhat et](#)  
145 [al. 2008](#), [J et al. 2010](#)).

146 In recent years, researchers have begun to investigate the role of environmental  
147 toxins, especially of air pollution, in the induction of stroke ([Yin et al. 2015](#), [Cevik et](#)  
148 [al. 2015](#), [Shah et al. 2015](#), [Lipsett et al. 2011](#)). Evidence has shown that both short-  
149 term and long-term exposure to air contaminants, including gaseous pollutants and  
150 particulate matter, were significantly associated with stroke prevalence and mortality,

151 *ceteris paribus* ([Shah et al. 2015](#), [Lipsett et al. 2011](#)). Exposure to ambient air  
152 pollution has also been found to be an important contributor to the risk of cardiac  
153 diseases and deteriorating vascular function ([Lundback et al. 2009](#), [Shah et al. 2013](#)).  
154 Recent research by [Roberts, Voss, and Knight \(2014\)](#) revealed that physical inactivity  
155 is more prevalent in communities with poor air quality; the authors suggested that  
156 such findings might be attributed to physiological (e.g., difficulty breathing) and  
157 psychosocial (e.g., antipathy to smoke) effects. Even if these negative effects are  
158 overcome, exercising in polluted communities contributes to adverse lifestyle-  
159 associated health outcomes ([Li et al. 2015](#)).

### 160 **2.2.2. BE elements related to stroke**

161 Despite the lack of research explicitly connecting BEs with stroke risk, evidence has  
162 broadly connected BEs with three identified modifiable risk factors of stroke, namely,  
163 lifestyle, physiological factors, and environmental toxins. Therefore, it is reasonable  
164 to hypothesise the indirect stroke-BE nexus. To this end, the following section  
165 elaborates on the potential mechanism by which stroke is affected by BEs as it relates  
166 to multidimensional elements — land use, transport system, open space, facilities, and  
167 architectural characteristics — followed by the classic definition proposed by the  
168 [World Health Organization \(2009\)](#).

169 Several studies have examined the correlations between stroke-related risks and land  
170 use. Recent research by [Su et al. \(2016\)](#) proposed a theoretical framework to illustrate  
171 how land use delivers health benefits against the worsening public health in China via  
172 the promotion of active lifestyles and the control of air pollution. Specifically, the  
173 authors employed SEMs to construct the association between land use pattern and  
174 typical chronic disease incidences. Mediated by increased PA and decreased PM<sub>2.5</sub>  
175 concentrations, an abundance of institutional land (e.g., schools, community centres,  
176 and universities) is found to be associated with a lower risk of type 2 diabetes,  
177 hypertension, and cardiopathy. Additionally, the role of mixed land use in supporting  
178 active travel behaviour has been highlighted over the decades ([Cervero and](#)  
179 [Kockelman 1997](#)); correspondingly, a higher level of PA and a lower likelihood of  
180 obesity and hypertension have been observed in districts with mixed land use  
181 ([Malambo et al. 2016](#), [Zhang, Liu, and Liu 2015](#)). In contrast, previous studies have  
182 revealed that increased industrial and residential areas tend to exacerbate ambient air  
183 pollution ([Bertazzon et al. 2015](#), [Henderson et al. 2007](#)), in turn potentially increasing

184 stroke risk.

185 The effects of transport on PA and thus on stroke risk have been examined in previous  
186 studies. Better street connectivity could facilitate walking and moderate-intensity PA  
187 ([Frank et al. 2005](#), [Moeller 2013](#)). Moreover, an increased number of sidewalks and  
188 paved streets has been correlated with increased walking as a mode of transport  
189 ([Zhang et al. 2014](#)). Additionally, [Lachapelle and Frank \(2009\)](#) found that public  
190 transit users were more likely to meet the recommendation for daily PA than  
191 automobile users. [Rundle et al. \(2007\)](#) also observed that the densities of bus and  
192 subway stops were negatively associated with body mass index (BMI). Therefore, BE  
193 elements that encourage the ability to choose a mode of active transport might  
194 counteract the risk of stroke.

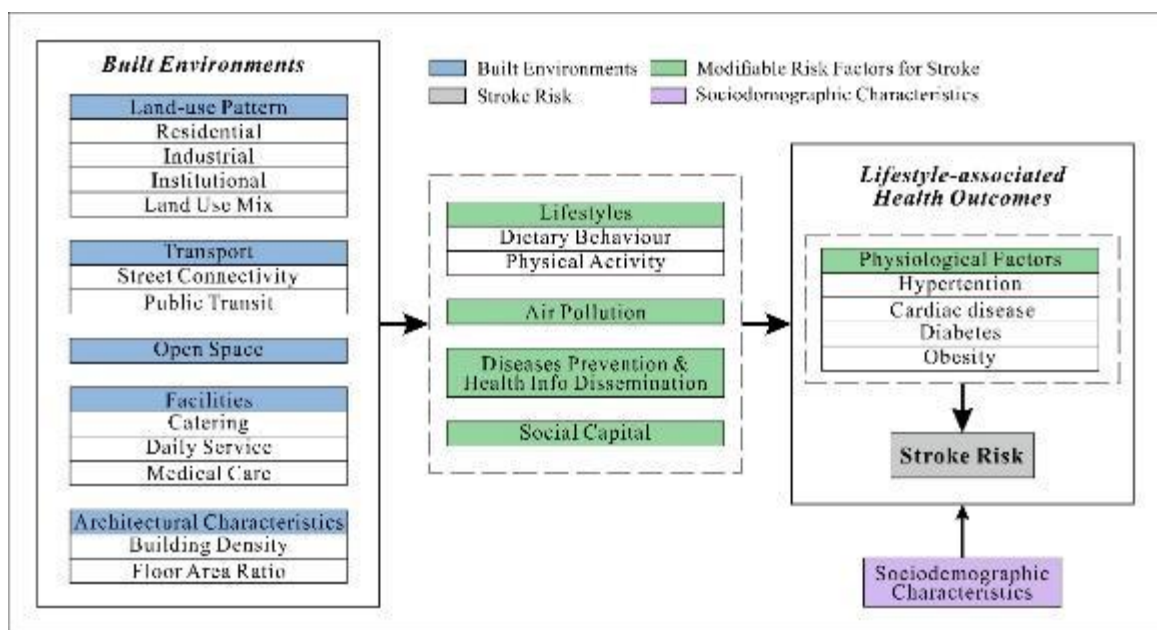
195 Sufficient access to both green space and public space delivers health benefits against  
196 stroke risks, such as better general health status ([Mowen et al. 2007](#)) and a lower risk  
197 of cardiovascular disease and diabetes ([Chum and O'Campo 2015](#), [Xie et al. 2018](#)).  
198 Open space is regarded as the “third place”, after the home and the workplace, in  
199 individual social lives since it provides opportunities for social interaction and access  
200 to social network-based resources ([Mowen and Rung 2016](#)). Accordingly, it largely  
201 supports the accumulation of social capital, promoting better self-management in  
202 terms of health ([Xie et al. 2018](#)). Moreover, a plethora of studies has shown that better  
203 access to open space, e.g., greater proximity, higher abundance, and a better  
204 landscape, is positively associated with various types of PA ([Kaczynski et al. 2009](#),  
205 [Schipperijn et al. 2017](#), [Lu, Sarkar, and Xiao 2018](#)). Public open space could therefore  
206 play an important role in reducing the risk of stroke by facilitating PA and promoting  
207 social capital.

208 It has also been postulated in the literature that access to a variety of facilities plays a  
209 critical role in shaping individual lifestyles, which might impact stroke risk. A recent  
210 study on urban China suggested that having more destinations within walking distance  
211 is positively correlated with more PA ([Zhou, Grady, and Chen 2017](#)). In contrast, a  
212 higher density of catering facilities could lead to an increased risk of cardiovascular  
213 disease and obesity, which can be attributed to excessive fat and calorie intake ([Chum  
214 and O'Campo 2015](#), [Ford and Dzewaltowski 2008](#)). Moreover, prior research has also  
215 emphasised the importance of medical care facilities to disease prevention and health  
216 information dissemination ([Anasi 2012](#)); easy access to medical facilities might help

217 reduce stroke risk.

218 The architectural characteristics of BEs have been closely linked with air pollution  
 219 and PA. Recent research (in the context of high-density urban China) has shown that  
 220 the PM<sub>2.5</sub> concentration was strongly associated with building volume density and  
 221 building coverage ratio ([Shi et al. 2017](#)). Dense urban environments and high-  
 222 intensity urban development not only play important roles in the generation of air  
 223 pollutants but also contribute to the urban heat island effect ([Sharifi 2019](#)), which  
 224 impacts urban airflow dynamics and thereby influences the movement and  
 225 concentration of pollutants ([Agarwal and Tandon 2010](#), [Edussuriya and Chan 2015](#)).  
 226 Additionally, building compactness was associated with PA. For example, by  
 227 separating density indicators into seven subtypes, [Forsyth et al. \(2007\)](#) found that both  
 228 housing and total building density were positively correlated with walking for travel  
 229 purposes but negatively associated with walking for work and leisure purposes.

230 As such, given the multiplicity of stroke-associated BEs, we put forward a conceptual  
 231 framework to capture the potential impact of multidimensional BEs on stroke risk  
 232 (**Fig. 1**). The goal of the current research is to understand the complexity of the  
 233 stroke-BE nexus, thereby fleshing out the current picture on the nexus between BEs  
 234 and lifestyle-associated health outcomes with regards to conceptual frameworks and  
 235 empirical evidence.



236

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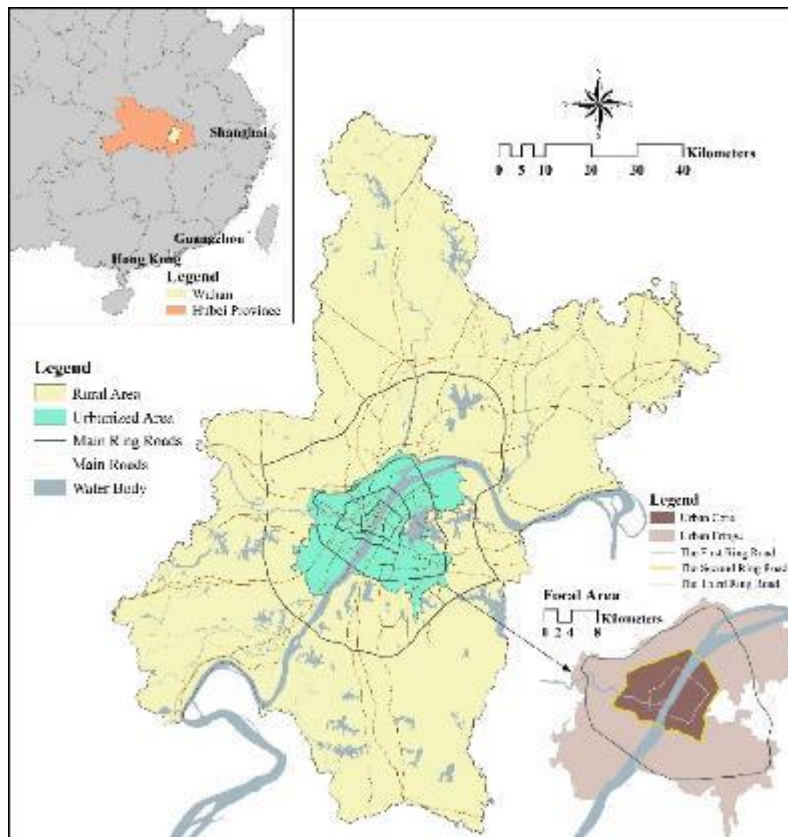
**Figure 1.** Conceptual framework of the stroke-BE nexus.

238 **3. Research design**

239 **3.1. Study area**

240 This study focuses on the urbanised area in Wuhan, China (**Fig. 2**). Wuhan has a  
241 population of 10.7 million (as of 2015) and a total area of 8,494.4 km<sup>2</sup>. It is the most  
242 populated city and the economic and political centre of central China. Similar to other  
243 Chinese metropolises ([Gu, Kesteloot, and Cook 2015](#)), Wuhan has witnessed  
244 remarkable urban expansion, with prosperous industrialisation and rapid economic  
245 development. During this period, lifestyle-associated diseases, especially stroke and  
246 its precursors, have gradually become substantial burdens on public health ([Xie et al.  
247 2018](#)).

248 The geographical distribution of the population in Wuhan is highly centralised ([Xie et  
249 al. 2019](#)). The area of the urban core (i.e., the urbanised area within the boundaries of  
250 the second ring road) and the urban fringe (i.e., the urbanised area outside the  
251 boundaries of the second ring road) is approximately 2% and 8% of that of the rural  
252 district, respectively, but both of these areas cover approximately 30% of the total  
253 population of Wuhan. Our precise focal area covers an area of 734.8 km<sup>2</sup> and has a  
254 population of approximately 6.4 million. The area is characterised by high density, a  
255 rapidly ageing population, and poor air quality, all of which are potentially associated  
256 with stroke risk. In 2015, the gross population density was 8,711 people per km<sup>2</sup> in  
257 this area. This population density is approximately 10 and 5 times that of Toronto,  
258 Canada (954 people per km<sup>2</sup>) and Portland, U.S. (1,684 people per km<sup>2</sup>), respectively,  
259 where prior BE-cardiovascular disease studies have been conducted ([Chum and  
260 O'Campo 2015](#), [Li et al. 2009](#)). Additionally, the ageing population has rapidly grown  
261 and now exceeds 20% of the total population, and the air quality in Wuhan is low;  
262 according to the 2015 Wuhan Environmental Status Bulletin, hazardous pollution  
263 days comprised nearly half the year, and such issues are much more severe in built-up  
264 areas.



**Figure 2.** Map of Wuhan.

### 3.2. Data source

This study used three types of data: medical data, BE data, and community-level sociodemographic data. Medical data on stroke cases in Wuhan from 2015 were collected from the Wuhan Emergency Medical Centre (Wuhan EMMC). Each medical case included the patient's age, gender, disease type, onset time, and geocoded home address. The disease types were diagnosed by professional doctors, and the home addresses were geocoded by an independent commercial company under the supervision of Wuhan EMMC. We successfully extracted 10,971 stroke cases from the dataset according to the International Classification of Diseases (ICD-10: I60-I69). The crude stroke incidence in our focal area was calculated to be 170.1/100,000, which is slightly lower than the average stroke incidence in urban China (203.6/100,000) (NCCD, 2017). Moreover, recent nationwide research has suggested that the stroke incidence in China exhibits a considerably geographical dependency at the province level (Wang et al. 2017). The stroke incidence in Changsha — a twin city with Wuhan in central China — is therefore employed to be compared with our data. As estimated by Sun et al. (2014), based on a self-sampled stroke surveillance network, the stroke incidence in Changsha was comparable, at

284 168.5/100,000 in 2011, although this figure exhibited an upward trend over time.  
285 These figures showed the relatively high representativeness of our dataset in depicting  
286 stroke prevalence in urban China.

287 We obtained the BE data and community-level sociodemographic data, including land  
288 use, facilities, housing, transport (except sidewalks), buildings, and community-level  
289 population, employment, age structure, and gender ratio, from the Wuhan Land  
290 Resources and Planning Information Centre for 2015 (**Table 2**). Due to the data  
291 restrictions on public institutions, sidewalk data were obtained from an open-access  
292 and volunteered GIS platform, namely, OpenStreetMap.org.

### 293 **3.3. BE measurements**

294 The unit of analysis for this study is the individual community (n=1,237). The  
295 community, namely, the *juweihui*, is the smallest administrative unit and the basic  
296 official unit for conducting population statistics and managing public services for  
297 residents in urban China. As previously stated, we adapted the framework from the  
298 [World Health Organization \(2009\)](#) to capture and measure the multidimensional  
299 stroke-related BEs of these communities. The five categories measured were (1) land-  
300 use patterns, (2) transport, (3) open spaces, (4) facilities, and (5) architecture.  
301 Variables for spatial variations in population and employment were also included in  
302 this research due to their roles in shaping urban form. The community-level  
303 sociodemographic characteristics were served as controlled variables in the analysis.  
304 **Table 2** shows the descriptive statistics and descriptions of the variables that were  
305 included in this study.

306 Two metrics were employed to measure land-use patterns: land-use mix and land-use  
307 proportion. The former indicator emphasises the diversity of land use. Given the  
308 development of communities, especially those in urban cores that were dominated by  
309 an exclusive residential pattern, the indicator was measured by Shannon's diversity  
310 index for an all-land-use mix (SAM) and Shannon's diversity index for a residential-  
311 use-excluded mix (SREM). The proportions of variables were used to reflect the  
312 abundance of the specified land-use classes. Based on previous research, three land-  
313 use classes (residential, industrial, and institutional) were eventually included.

314 With regard to transport, intersection density, as suggested by [Moeller \(2013\)](#), was  
315 used to characterise street connectivity. Sidewalk provisioning was measured by the

316 density of the sidewalks. We also considered the accessibility of public transit. To do  
317 this, catchment-based accessibility metrics were employed due to the significant  
318 variation in community sizes and the uneven distribution of residents. **Fig. 3(a-b)**  
319 shows a comparison of the catchment-based and container-based metrics for facility  
320 access. With the container-based metric, facilities within a specified community were  
321 considered accessible. However, this measure did not consider the facilities outside  
322 the community boundaries or the facilities far from residents; thus, this metric might  
323 either underestimate or overestimate the mobility of residents and lead to a modifiable  
324 area unit problem (MAUP) ([Xie et al. 2018](#)). Specifically, we geocoded all residential  
325 building addresses within communities and then measured the number of transit  
326 facilities within a catchment area (i.e., road network distance) of 600 metres (an  
327 approximate 10-minute walking distance) around each residential building. Finally,  
328 we calculated the number of accessible facilities for every residential building in each  
329 community. The median value of the facilities was used to represent accessibility to  
330 public transit within each community, as the standard deviations were higher than the  
331 mean values in most communities.

332 The interaction between facilities and residents was indicated by the accessibility of  
333 catering (i.e., restaurants, fast-food stores, groceries, and markets), daily services (i.e.,  
334 ATMs, banks, telecommunication businesses, and laundromats) and medical care (i.e.,  
335 clinics and polyclinics). The measurement of accessibility to these facilities was  
336 consistent with the metrics for public transit accessibility.

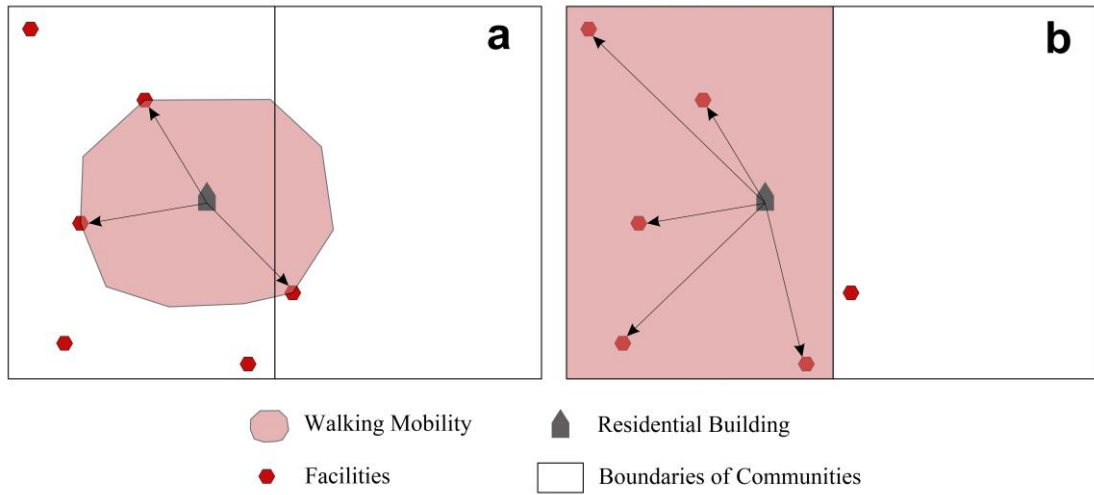
337 Open space was measured using the proportion of area devoted to public parks and  
338 squares. Urban omnibus parks (UOPs), community parks (CPs), and public  
339 squares/plazas are three of the main public open spaces in Wuhan. The UOPs and CPs  
340 are mainly built for social interactions, recreational activities and daily entertainment  
341 activities. These two types of parks are associated with a relatively adequate presence  
342 of green space but have great disparities in terms of scale, amenities, management and  
343 maintenance. Because squares/plazas are mainly established for commemorative  
344 functions, they are poorer in terms of auxiliary facilities and greenery than parks.

345 We employed two types of spatial indicators to reflect architectural characteristics:  
346 intensity and density. The intensity indicator was measured by the average floor area  
347 ratio (FAR) within communities, and the density features were indicated by building  
348 density.



349 Population density and employment density are historically intertwined with zonal  
350 economic development, urban form (e.g., compactness versus sprawling), and BEs  
351 ([Cervero and Kockelman 1997](#), [Wu and Gopinath 2010](#)); correspondingly, such  
352 measurements were also included in the current research. Gross population density  
353 and net population density were initially considered, but given the high correlation  
354 between these variables, the former variable was excluded, as net population density  
355 can capture the concentration and compactness of the population more appropriately.

356 With respect to sociodemographic characteristics, age structure, percentage of males,  
357 and average housing price were included in the current research. The percentages of  
358 adults (residents between 19 and 59 years of age) and older adults (residents over 60)  
359 were used to reflect the age structure. Considering the potential correlations among  
360 socioeconomic status, stroke risk, and BEs within communities, the average housing  
361 prices within communities were employed as a proxy for the wealth values of  
362 residents to reduce endogeneity. [Moudon et al. \(2011\)](#) aggregated the values of land  
363 parcels and individual-level wealth metrics to measure neighbourhood wealth and  
364 property values and found that the housing-price indicator was more predictive of  
365 health status than the individual socioeconomic status (SES). Moreover, several  
366 studies have indicated that area-level housing prices strongly affect the wealth and  
367 debt of households and have significant implications for stroke and related health  
368 outcomes ([Fichera and Gathergood 2013](#)); thus, area-level housing prices could be  
369 used as a proxy for area-level SES in health studies ([Sohn 2013](#)). Additionally, the  
370 inclusion of the housing price variable could help reduce potential residential self-  
371 selection bias, i.e., individuals deliberately opt for where to live according to their  
372 sociodemographic traits and preferences ([Cao, Mokhtarian, and Handy 2009](#)), since  
373 housing affordability plays a key role in choosing residential locations and, in turn,  
374 the community BEs in urban China ([Xiao et al. 2017](#)).



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**Figure 3.** Comparison of (a) catchment-based and (b) container-based metrics for facility access.

**Table 2** Summary and definition of variables.

Domains	Variables	Min	Max	Mean	Std.	Definition
Sociodemographic Characteristics	Male	0.447	0.982	0.516	0.062	Male population/community population. (%)
	Older Adult	0.017	0.658	0.193	0.075	Population older than 60/community population. (n)
	Adult	0.332	0.919	0.722	0.078	Population aged between 19 and 59/community population. (n)
	Housing Price	1162.000	41080.000	11654.275	2395.757	Average housing price within communities. (yuan)
Land-use Pattern	SAM	0.010	2.111	1.009	0.419	Shannon's diversity index for all land-use mix.
	SREM	0.004	1.523	0.878	0.366	Shannon's diversity index for residential-use-excluded mix.
	Residential Land Use	0.089	0.962	0.367	0.133	Proportion of urban residential land use. (%)
	Institutional Land Use	0.000	0.790	0.096	0.235	Proportion of governmental land use, including areas with education and social welfare uses. (%)
	Industrial Land Use	0.000	0.954	0.046	0.187	Proportion of industrial land use, including areas with research, manufacturing, and assembling uses. (%)
Transport	Intersection	0.023	7.594	0.347	0.563	Number of intersections/community land area. (1/ha)
	Sidewalk	0.002	0.313	0.022	0.089	Total sidewalk length/community land area. (km/ha)
	Bus Stop	0.000	11.000	1.283	1.825	Accessibility to bus stops. (n)
	Metro Station	0.000	7.000	0.424	1.398	Accessibility to metro stations. (n)
Open Space	CP	0.000	0.431	0.006	0.018	Proportion of areas devoted to community parks. (%)
	UOP	0.000	0.641	0.017	0.037	Proportion of areas devoted to urban omnibus parks. (%)
	Square	0.000	0.146	0.000	0.000	Proportion of areas devoted to public squares and plazas. (%)
Facilities	Catering	1.000	319.500	122.786	129.883	Accessibility of catering facilities. (n)
	Daily Service	11.500	245.000	57.491	44.891	Accessibility of daily service facilities. (n)
	Medical Care	0.000	32.000	8.554	6.484	Accessibility of medical care facilities. (n)
Architectural Characteristics	Building Density	0.001	0.846	0.138	0.268	Area of building footprint/community land area (%)
	FAR	0.041	7.485	1.830	0.989	Gross floor area/area of the plots within communities. (ratio)
Population Density	Net Population Density	17.028	6373.448	446.860	1095.459	Population/residential land area of the community. (1/ha)
Employment Density	Employment Density	1.050	3398.238	40.281	165.235	Employment population/land area of the community. (1/ha)

### 382 3.4. Bayesian CAR model

383 The Bayesian CAR model was employed to (1) smooth and visualise stroke risk and  
384 (2) examine the association of BE elements with stroke risk.

#### 385 3.4.1. Smoothed and standardised risk of stroke

386 In previous studies, two major indicators were employed to quantify the epidemic  
387 circumstances in a specific area: (1) crude incidence ([Su et al. 2016](#)) and (2)  
388 standardised incidence ratios (SIRs) ([Maheswaran et al. 2012](#)). Crude incidence  
389 represents the number of new cases per population at risk while neglecting the  
390 significant impact of the demographic characteristics of the study area on diseases. As  
391 an alternative indicator, SIRs measure the ratio of observed cases to expected cases,  
392 which is standardised by age and gender and could thus offset the statistical bias  
393 caused by demographic characteristics. However, SIRs are subject to large variation,  
394 especially in areas with small populations, as the number of expected cases can be  
395 small, and a very small change in the number of observed cases might exaggerate  
396 epidemic fluctuations ([Wu et al. 2004](#)). To avoid this problem, a smoothed SIR  
397 (SSIR) measure was used to visualise the risk of stroke in this study.

398 The Bayesian CAR model is a new approach for modelling spatial data with limited  
399 observations. This model incorporates spatially structured heterogeneity with  
400 hierarchical parameters into the estimation of SSIRs, thereby eliminating the  
401 dependence of variance on population size ([Ebrahimipour et al. 2016](#), [Wu et al. 2004](#)).  
402 Recently, the Bayesian CAR model has been increasingly used to explore the  
403 geographical variation in diseases ([Yin et al. 2014](#), [Alegana et al. 2013](#)).

404 Following [Hegarty, Carsin, and Comber \(2010\)](#), a Poisson CAR model (**Eqs. (1)–(3)**)  
405 was applied to the estimation of SSIRs using the Markov chain Monte Carlo (MCMC)  
406 method.

$$407 \quad O_i \sim \text{Poisson}(E_i R_i) \quad (1)$$

$$408 \quad E_i = \sum \sum N_{ijk} \frac{O_{jk}}{N_{jk}} \quad (2)$$

$$409 \quad \log(R_i) = \alpha + v_i + u_i \quad (3)$$

410 Here, the fitted value of  $R_i$  represents the final SSIR within community  $i$ .  $E_i$  and  $O_i$   
411 represent the expected and observed number of cases within community  $i$ ,  
412 respectively. Specifically,  $E_i$  is defined as the number of age-sex-specific cases.  $N_{ijk}$

413 represents the total population in community  $i$  for age group  $j$  and sex  $k$ ;  $O_{jk}$  represents  
 414 the observed number of cases for age group  $j$  and sex  $k$ , and  $N_{jk}$  represents the total  
 415 population for age group  $j$  and sex  $k$ .  $\alpha$  represents the intercept;  $v_i$  and  $u_i$  represent the  
 416 unstructured and structured spatial correlations, and their prior distributions were  
 417 assigned according to **Eqs. (4)–(5)**, respectively, with  $\tau_h^2$  and  $\tau_e^2$  controlling the  
 418 variance and equalling the prior Gamma distributions (0.5 and 0.0005, respectively),  
 419 as suggested by [Aguero-Valverde and Jovanis \(2006\)](#). In this study, the first-order  
 420 adjacent neighbourhood was used to represent spatial dependency, and  $n_{\delta_i}$  denoted  
 421 the total number of adjacent communities for area  $i$ .

$$422 \quad v_i \sim N(0, \tau_h^2) \quad (4)$$

$$423 \quad [u_i | u_{j, i \neq j}] \sim N\left(\sum_{j \in \delta_i} \frac{u_j}{n_{\delta_i}}, \frac{\tau_e^2}{n_{\delta_i}}\right) \quad (5)$$

#### 424 **3.4.2. Stroke-BE association**

425 The following Bayesian-Poisson CAR model was employed to examine the predictors  
 426 of stroke risk under the domains of BEs and sociodemographic characteristics:

$$427 \quad O_i \sim \text{Poisson}(E_i R_i) \quad (6)$$

$$428 \quad E_i = N_i \frac{O}{N} \quad (7)$$

$$429 \quad \log(R_i) = \alpha + \sum \beta_{ij} x_{ij} + v_i + u_i \quad (8)$$

430 where  $R_i$  represents the stroke risk within a given area  $i$ .  $E_i$  and  $O_i$  denote the expected  
 431 and observed number of cases, respectively, within area  $i$ . More specifically, given the  
 432 introduction of demographic characteristics (e.g., age structure variables),  $E_i$  was  
 433 different from that in **Eq. (2)**. In this model,  $E_i$  is defined as the population-weighted  
 434 average number of cases.  $N_i$  represents the total population in community  $i$ ;  $O$  and  $N$   
 435 denote the total number of observed cases and the total population of the focal area,  
 436 respectively.  $x_{ij}$  represents the predictor, and  $\beta_{ij}$  represents the corresponding  
 437 coefficient. The settings of the other variables and the prior distributions of the  
 438 parameters were assigned according to **Eqs. (1)–(5)**.

439 Additionally, given the richness of our dataset, multicollinearity issues potentially  
 440 emerge, which might contribute to biased estimations. The classic variance inflation  
 441 factors (VIFs) were therefore implemented for classic Poisson regressions (instead of  
 442 the Bayesian-Poisson CAR model due to computing limitations) to determine the

443 input variables. In each round of VIF calculation, a variable with a VIF greater than  
444 10 was excluded from the Bayesian CAR model. Ultimately, three variables — the  
445 residential land-use proportion, bus stop accessibility, and daily services accessibility  
446 — were excluded from the estimation model.

### 447 **3.5. Hot spot analysis**

448 The hot spot analysis was performed to identify the geographical cluster of SSIRs  
449 based on the Getis-Ord  $G_i^*$  statistic ([Ord and Getis 1995](#)). The resultant z-score  
450 indicates the spatial aggregation of stroke risks. Communities with a high z-score (i.e.,  
451  $z > 1.96$ ) and a low  $p$ -score (i.e.,  $p < 0.05$ ) have statistically significant hot spots. The  
452 model was implemented using the fixed distance band tool in ArcGIS 10.3.

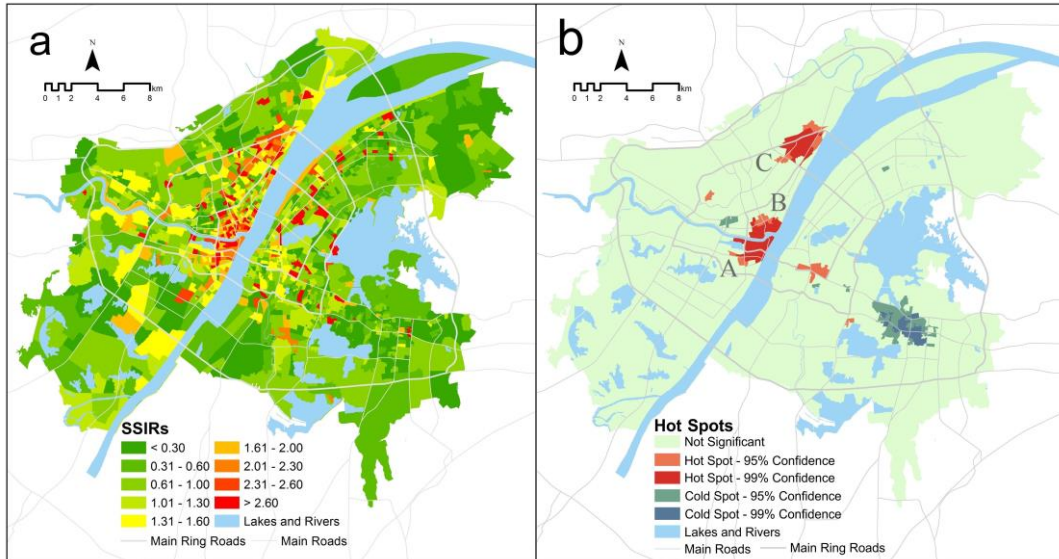
## 453 **4. Results**

### 454 **4.1. Distribution of stroke risk**

455 **Fig. 4-a** depicts the distribution of SSIRs in Wuhan. Strong geographical variation in  
456 the SSIRs of stroke was observed in the city. Residents living in the urban core  
457 appeared to have a higher risk of stroke, whereas residents living along the urban  
458 fringe were less likely to have strokes. **Table 3** shows the comparison of SSIRs  
459 between the two urban regions. Specifically, high-risk communities (i.e., communities  
460 with a stroke risk greater than 2) were broadly concentrated in the urban core. Within  
461 this area, the percentage of high-risk communities was noticeably higher than that in  
462 the urban fringe (23.82% versus 10.09%).

463 The geographical clusters of high SSIRs were identified using hot spot analysis (**Fig.**  
464 **4-b**). All three distinct hot spots included many high-density and overcrowded  
465 neighbourhoods compared to the average development level in the city (**Table 4**).  
466 Specifically, two hot spots were located exactly over the city centre. Hot Spot A was  
467 located to the south of the Han River (**Fig. 5-a**). The communities within Hot Spot A  
468 were mainly composed of old state-owned enterprises with a high concentration of  
469 established enterprise workers, and the majority of residents were retired with a low  
470 income. Housing in the area tends to forgo necessary maintenance due to the  
471 bankruptcy of these state-owned enterprises. Hot Spot B was located in a traditional  
472 and historic commercial district north of the Han River (**Fig. 5-b**). In this area, the  
473 living and commercial spaces were highly mixed, the building density was extremely  
474 high, and the public space was scarce, accounting for only 1.4% of the total area. Hot  
475 Spot C was located along the edge of the city centre and was surrounded by main

476 arterial roads (**Fig. 5-c**). This area included both old and new communities, with high  
477 building and population concentrations. Our results implied that residents living in the  
478 compact urban core tended to have a higher risk of stroke, whereas residents who  
479 lived in the urban fringe were less likely to be diagnosed with stroke.



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**Figure 4.** Distribution and high-risk clusters of stroke risks.

482 **Table 3** Comparison of stroke risks between the urban core and the urban fringe.

SSIRs	<1.0	1.0-2.0	2.0-3.0	>3.0
Proportion of communities within urban core (%)	37.25	38.93	15.42	8.40
Proportion of communities within urban fringe (%)	59.79	30.12	5.90	4.19

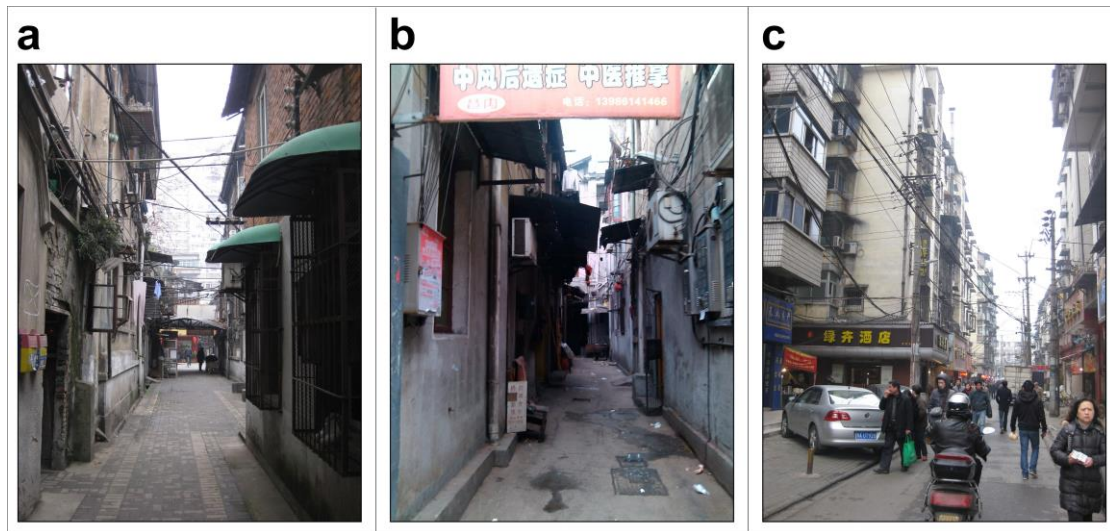
483  
484 **Table 4** Comparison of the hot spots and the overall study area in terms of density.

Area	PD	NPD	BD	FAR
Study Area	87.735	446.860	0.138	1.83
Hot Spot A	210.179	459.089	0.362	2.31
Hot Spot B	886.891	1257.958	0.671	2.63
Hot Spot C	176.361	551.419	0.198	1.74

485 *Abbreviations:* population density (PD), net population density (NPD), building density (BD), floor area ratio (FAR)

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488 **Figure 5.** Typical cityscapes in the high-risk clusters for stroke: (a) Hot Spot A; (b)  
 489 Hot Spot B; (c) Hot Spot C.

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491 **4.2. Impact of BEs on stroke**

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Using the Bayesian CAR analysis, we identified an association between stroke risk and BE variables (**Table 5**). With regard to the sociodemographic characteristics of communities, we found that the percentage of older adults was positively correlated with stroke risk, which implied that a higher risk of stroke was observed in ageing communities. Land-use patterns were also closely associated with stroke. The proportion of institutional land use was negatively associated with stroke risk, indicating that people were less prone to strokes in communities with larger amounts of institutional land use. However, there was no significant correlation between the land-use mix variables and stroke risk. The association between stroke risk and the transport system also remained unclear in our research. Regarding access to open spaces, both CPs and UOPs were negatively related to stroke risk, suggesting communities with greater area devoted to parks were potentially beneficial for stroke prevention. Regarding facilities, only medical care facilities acted as negative predictors of stroke risk, which indicated that better access to medical care facilities could benefit stroke prevention. Additionally, the results showed that building density was positively related to stroke, and such findings were similar to another density indicator: net population density. FAR was not significantly related to stroke risk, which indicated that residents living in crowded communities rather than high-FAR communities were more prone to stroke. In other words, high-rise buildings alone were not correlated with stroke risk in Wuhan, China.

513 **Table 5** BE-stroke association.

Domains	Variables <sup>a</sup>	Mean	2.5% CI <sup>b</sup>	97.5% CI <sup>b</sup>
Sociodemographic Characteristics	Male	0.0271	-0.0612	0.1217
	Older Adult *	0.2502	0.1153	0.3965
	Adult	0.0096	-0.0430	0.0686
Land-use Pattern	Housing Price	0.1026	-0.0419	0.2090
	SAM	-0.0663	-0.1932	0.0525
	SREM	-0.0483	-0.1647	0.0608
	Institutional Land Use *	-0.1422	-0.2592	-0.0177
Transport	Industrial Land Use	-0.0535	-0.2048	0.0804
	Intersection	-0.0901	-0.2118	0.0393
	Sidewalk	-0.0694	-0.1972	0.0612
Open Space	Metro Station	0.0333	-0.0612	0.1261
	CP *	-0.1714	-0.3265	-0.0099
	UOP *	-0.1573	-0.2787	-0.0471
Facilities	Square	-0.0208	-0.1209	0.0813
	Catering	0.0344	-0.0583	0.1313
Architectural Characteristics	Medical Care *	-0.1785	-0.3122	-0.0403
	Building Density *	0.2857	0.1256	0.4516
Population Density	FAR	0.0110	-0.0604	0.0704
Employment Density	Net Population Density *	0.1988	0.0583	0.3347
Model Fit (DIC)	Employment Density	0.0584	-0.1059	0.2326
	4870.805			

514 <sup>a</sup> Independent variables have been standardised.515 <sup>b</sup> CI denotes the credible interval.516 \* Significant at  $p < 0.05$ 517 **5. Discussion**518 **5.1. Health benefits for stroke prevention related to BEs**

519 Our research has identified several community-BE elements that could deliver health  
520 benefits for stroke prevention after controlling for sociodemographic characteristics.  
521 First, a lower stroke risk was observed in communities with greater area devoted to  
522 institutional land use, which might be attributed to the increase in PA (Su et al. 2016).  
523 Similar findings can be found in research by Su et al. (2016) who revealed that the  
524 abundance of institutional land use is negatively correlated with risks of cardiopathy  
525 and hypertension in Shenzhen, China. In contrast, land-use mix measures were not  
526 significantly correlated with stroke risk, which was inconsistent with previous  
527 findings that showed positive health benefits from mixed land use in urban China,  
528 such as lower risk of hypertension, from mixed land use in urban China (Su et al.  
529 2016) and more PA (Gao, Ahern, and Koshland 2016). Second, residents in  
530 communities with sufficient park areas were less likely to suffer from strokes because

531 parks offer valuable opportunities for interpersonal socialisation and participation in  
532 modest daily PA ([Mowen and Rung 2016](#), [Kaczynski et al. 2009](#)). This finding is  
533 consistent with the research by [Chum and Patricia \(2013\)](#) and [Chum and O'Campo](#)  
534 [\(2015\)](#) on cardiovascular diseases in Toronto, Canada. Although CPs were of lower  
535 quality in terms of size, management and amenities than UOPs, our results suggested  
536 that both types of parks were conducive to reducing stroke risk. In contrast, the  
537 association between stroke risk and squares remained unclear. Third, communities  
538 with closer medical care facilities tended to have lower stroke risk, further  
539 emphasising the role of both public primary and tertiary medical care facilities in  
540 stroke prevention. This finding might be attributed to the crucial role of these health  
541 services in disease prevention and health information dissemination ([Anasi 2012](#)).  
542 Despite the difference in subject matter, this finding is partly consistent with the  
543 research by [McLafferty and Wang \(2009\)](#) conducted at a similar lifestyle-associated  
544 health outcome level (i.e., late-stage cancer risk) in Illinois, U.S.

545 Moreover, these elements were identified under the consideration of the multiplicity  
546 of potential stroke-associated BEs. Given the multiple pathways linking BEs and  
547 stroke, the identification is of relatively high robustness. It should also be noted that  
548 these salubrious elements lie within various BE dimensions. These issues, to a large  
549 extent, imply that incomprehensive conceptual frameworks focusing on  
550 exclusive/narrow BE domains might not sufficiently capture the intricate BE-stroke  
551 nexus.

## 552 ***5.2. High-density urban context correlates to stroke***

553 The current research also yielded insight into the association between high-density  
554 development and increased stroke risk. The geographical examination showed that  
555 residents living in dense and compact urban cores tended to have a higher risk of  
556 stroke than residents living in the urban fringe. The statistical analyses revealed that  
557 net population density and building density are positively correlated with stroke risk.  
558 One reasonable explanation is that air pollution and decreased PA might act as  
559 mediators of these associations. Many studies have concluded that high-density  
560 development, especially in the central areas of cities, is positively associated with  
561 ambient air pollution ([Frank and Engelke 2005](#), [Shi et al. 2017](#)). Several major air  
562 pollution sources, e.g., use of personal vehicles, discharge of domestic refuse, and  
563 fuel/gas consumption, are remarkably more prevalent in populated and dense areas  
564 ([Hixson et al. 2012](#)). Increased human activities and building construction in these

565 areas also contribute to the exacerbation of the urban heat island ([Sharifi 2019](#)), in  
566 turn potentially obstructing the air flow condition and diffusion of pollutants ([Agarwal  
567 and Tandon 2010](#)). Moreover, exercising and walking in polluted environments can  
568 expose residents to combustion-related airborne pollutants and amplify the deleterious  
569 effects of harmful air pollution ([Li et al. 2015](#)). [Li et al. \(2015\)](#) argued that such  
570 situations can be severe in urban China due to the worsening air pollution crisis and  
571 the prevalence of outdoor-oriented activities.

572 Although prior studies have found that high-density and compact communities do  
573 provide residents with opportunities to engage in PA in developed countries (e.g.,  
574 [Frank et al. \(2005\)](#); [Rundle et al. \(2007\)](#)), several findings supporting the opposite  
575 conclusion have been observed regarding high-density developing countries (e.g.,  
576 [Salvo et al. \(2014\)](#); [Reis et al. \(2013\)](#)). A recent study of ultra-dense Zhongshan,  
577 China, found that, after the data were controlled for sociodemographic characteristics  
578 and other BE elements, an increase in population density would hinder both the total  
579 walking frequency and the total walking duration of older adults, who are particularly  
580 vulnerable to stroke ([Zhang et al. 2014](#)). [Lu, Xiao, and Ye \(2016\)](#) argued that there  
581 might be a threshold effect of density on walking, and hence, the excessive high-  
582 density urban context might obstruct participation in PA, at least to some extent.

583 In addition, density is considered to be a primary factor in planning practices and  
584 inevitably has an impact on the layout and formation of various BE elements;  
585 naturally, factors that support stroke prevention might be involved. To wit, the effects  
586 of urban density on stroke might be *inherently* underestimated in our results from the  
587 perspective of practices. Thus, Spearman's correlation coefficient was employed to  
588 examine the associations between density and stroke-related indicators (**Table 6**).

589 As shown in **Table 6**, building density is strongly and negatively correlated with the  
590 proportion of area devoted to public parks within communities (i.e., CPs and UOPs,  
591 with correlation coefficients less than  $-0.3$ ). Similar but weaker trends are also  
592 observed between building density and the proportion of institutional land use and  
593 between net population density and park variables. Higher densities, especially  
594 building density, might decrease the areas devoted to institutional land use and parks,  
595 which potentially hinders recreational activities. Within this context, decreased  
596 leisure-time PA has been generally associated with high building density in urban  
597 China ([Su et al. 2014](#), [Xu et al. 2010](#)). For example, a cross-sectional study in Hong  
598 Kong, where the population density is comparable to that of our study area, found

599 that, compared with medium-density areas, less leisure-time walking was observed in  
600 high-density neighbourhoods (Lu, Xiao, and Ye 2016). Moreover, Day (2016) argued  
601 that excessive density in Chinese megacities might lead to perceived overcrowding,  
602 thereby reducing recreational PA. The crowding barrier reflects, to a large extent, an  
603 imbalanced supply-demand relationship between local residents and space/facilities  
604 for PA; correspondingly, the local population density might have negative  
605 implications for the individual availability of open spaces/facilities due to the barriers  
606 presented by crowds (Ekkel and de Vries 2017). In this vein, recent research by Xie et  
607 al. (2018) on urban China highlighted the role of crowding in inducing chronic health  
608 conditions, particularly cardiovascular diseases, by obstructing participation in park-  
609 based PA. Therefore, even considering the relatively weak association between  
610 population density and the proportion of areas devoted to institutional land use and  
611 open space, highly dense communities potentially hinder leisure-time PA due to the  
612 exacerbation of perceived overcrowding and the supply-demand imbalance.

613

614 **Table 6** Summary of Spearman's correlation coefficients.

Variables	Older Adult	Institutional Land Use	CP <sup>a</sup>	UOP <sup>b</sup>	MCF
Net Population Density	0.198 **	-0.052	-0.267 **	-0.209 **	0.166 **
Building Density	0.320 **	-0.151 *	-0.373 **	-0.380 **	0.109 **

615 *Abbreviations:* community park (NP), urban omnibus park (UOP), floor area ratio (FAR),  
616 medical care facilities (MCF)

617 \* denotes that the correlation reaches statistical significance at the 0.05 level.

618 \*\* denotes that the correlation reaches statistical significance at the 0.01 level.

619

### 620 **5.3. Planning as a tool for proactive health intervention**

621 Planning intervention has emerged as a crucial need for a healthy urban agenda, since  
622 it is deemed a more proactive, fundamental, far-reaching, and longstanding strategy  
623 for health promotion than programmes that solely focus on changing the behaviours  
624 of small groups (Barton and Grant 2013). Some planning implications can be drawn  
625 from the current research. First, our research indicates that high-density development  
626 could be a potential public health concern, at least for stroke, which calls for the need  
627 to scrutinise and reconsider the applicability of the prevailing compact urban  
628 developments in transitional cities. Historically, compactness has been extensively  
629 acknowledged as an effective urban form that supports healthy urban living in  
630 developed countries (Ewing and Hamidi 2015). However, in many transitional  
631 countries such as China, population growth has continuously accelerated due to rapid

632 economic growth and large population size ([Wen and Goodman 2013](#)). Meanwhile, to  
633 pursue economic benefits and meet surging residential requirements, urban housing  
634 provisions are generally dominated by market-oriented programmes that are largely  
635 based on low-mixed/exclusive residential use ([Shin 2014](#)). Consequently, in recent  
636 decades and in the immediate future, the population and building density in  
637 transitional cities has and will continue to overload. Therefore, the applicability of  
638 compact urban development in these ultra-dense cities should be examined via  
639 systematic health impact assessments. Additionally, if such developments are  
640 identified as insalubrious, planners might need to consider de-densifying crowded  
641 urban cores and alleviating the extremely unbalanced distribution of population across  
642 urban areas. A rational way to achieve these goals is to promote the ‘new  
643 suburbanisation’ that aims at developing multifunctional centres to coordinate  
644 residential settlements, commercial areas, and workspaces ([Shen and Wu 2017](#)).

645 Second, as indicated by statistical analyses, several BE elements, i.e., the abundance  
646 of parks and institutional land and access to medical care facilities, still provide  
647 residents with the benefits of stroke prevention regardless of urban density. Therefore,  
648 the guarantee of sufficient opportunities for residents to access these salubrious  
649 resources could be seen as a crucial goal of planning practices for health promotion,  
650 especially considering that urban densities in transitional cities will inevitably  
651 increase in the future. Under such a scenario, planning practitioners should try to  
652 provide adequate urban therapeutic landscapes within communities (e.g., institutional  
653 land and parks). Medical facilities at various levels should also be delicately  
654 configured to ensure residents can access health care services, at least primary health  
655 care services, that are within walking distance.

656 Third, intervention scheme planning for health promotion and the reliability of the  
657 corresponding assessment projects of the schemes largely relies on systematic and  
658 comprehensive frameworks ([Ross et al. 2012](#)). By looking at the BE-stroke relation,  
659 our research highlighted the necessity of considering the multiplicity of health-related  
660 BEs in the development of these frameworks. Although the intricate nexus between  
661 BEs and lifestyle-associated health outcomes might vary with changes in focal cities,  
662 comprehensive conceptual frameworks allow planners to identify relevant BE  
663 elements accurately and to change them in a targeted manner.

664 **6. Conclusion**

665 Deciphering the nexus between BEs and lifestyle-associated health outcomes is of  
666 capital importance for crafting planning interventions. To this end, this article  
667 presented a conceptual framework of BE-stroke relation. Using Bayesian CAR  
668 methods, the research team was able to visualise the spatial distribution of stroke risk  
669 and identify BE correlates of stroke in the study area. The results revealed that BE  
670 variables were significantly associated with stroke risk at the community level after  
671 controlling for sociodemographic characteristics. The following determinants were  
672 identified: net population density, building density, an abundance of parks and  
673 institutional land, and access to medical facilities.

674 The current research has made new achievements as follows: (1) it elucidated the  
675 multiplicity of stroke-related BE elements and established a conceptual framework for  
676 capturing the intricate BE-stroke nexus; (2) inconsistent with prior studies conducted  
677 in developed countries, it indicated that compact urban developments might not be a  
678 panacea for public health promotion, at least not for stroke prevention; and (3) it  
679 identified several salubrious BE elements that act against the prevalence of stroke. As  
680 such, we argue that, given the multiple pathways connecting BEs and lifestyle-  
681 associated health outcomes, the establishment of comprehensive conceptual  
682 frameworks that sufficiently consider the multiplicity of health-related BEs deserves  
683 to be highlighted in further research, intervention schemes, and assessment projects.  
684 Policy makers and planning practitioners should scrutinise the applicability of  
685 prevalent compact urban developments in transitional cities. Additionally, more  
686 efforts should be made to guarantee access to the identified salubrious BE resources,  
687 especially considering that the urban densities in transitional countries will inevitably  
688 increase.

689 We end by acknowledging the limitations of our research. First, only one-year stroke  
690 records are included, while the distribution of stroke risks might vary over time.  
691 Second, detailed information about community-level air pollution and diseases that  
692 are precursors to stroke is not taken into consideration due to data availability and  
693 privacy protections in China. Excluding these variables might exaggerate the  
694 influence of BEs on stroke. Third, this study reveals only the stroke-BE correlations  
695 rather than causal relationships. Future studies could employ longitudinal datasets to  
696 eliminate temporal variation in stroke. The mediators of the association between the  
697 BE and stroke risk could be quantified. Additionally, a more sophisticated research

698 design (e.g., a quasi-experimental design) could be implemented to better understand  
699 the causal relationship between BEs and stroke.

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