



## Evaluating factors Influencing Active Transportation in Developing Metropolises

Shumaila Saleem<sup>1</sup>, Anuj Jaiswal<sup>2</sup>, Bivina G R<sup>3</sup>

[Shumailasaleem8@gmail.com](mailto:Shumailasaleem8@gmail.com)<sup>1</sup>, [Jaiswalanuj10@gmail.com](mailto:Jaiswalanuj10@gmail.com)<sup>2</sup>

1. Research scholar, Department of Architecture and Planning, MANIT, Bhopal, India, 462003;
2. Assistant Professor, Department of Architecture and Planning MANIT, Bhopal, India, 462003;
3. Assistant Professor, Department of Civil Engineering MANIT, Bhopal, India, 462003;

Received: 06/10/2024

Revised: 17/12/2024

Accepted: 15/01/2025

### Abstract

Urban transportation is facing challenges of heavy traffic of two-wheeler, cars, and trucks due to motorized transport. In order to sustain the environment as well as stop environmental degradation, active mobility was introduced as sustainable urban transportation. It indicates any mode of transport that involves physical activity, e.g., cycling, walking, skateboarding, skiing, etc. The study aims to prioritize the factors influencing active mode choices using the FAHP (fuzzy analytical hierarchy process) model, which incorporates expert opinions. To verify the robustness of results, a sensitivity analysis is also performed on the results. The study highlights that major performance indicators belonged to the infrastructure category, with network continuity, width of track, and separate tracks amongst the most influencing factors. To promote active mode choices, it is essential for decision makers and planners to consider factors that matter the most while planning to reduce the impact of transportation on the environment. By prioritizing infrastructure improvements and service provisions that directly address these key factors, cities can encourage a shift toward active mobility, mitigating the negative impacts of transportation on the environment and public health.

**Keywords:** active mobility; FAHP; Sensitivity; walking; cycling.

## 45 1. Introduction

46 The growing dependence on motorized transportation in developing cities has led to significant  
47 environmental degradation. In response, active mobility is increasingly recognized as a crucial  
48 component of sustainable urban transportation systems. The shift toward active mobility may  
49 help in achieving social sustainability objectives as it offers environmental benefits and  
50 contributes to improved public health and an enhanced quality of life (Rainieri G. et al.2024).  
51 But in order to effectively promote sustainable transportation practices, a thorough  
52 understanding of the factors influencing the adoption of active modes is required (Pisoni et  
53 al.,2022).

54 Active mobility, often referred to as active travel, active transport, or active transportation, is the  
55 movement of people via non-motorized means that is based on physical activity on the part of  
56 the passengers, which aids in health improvement and environmental benefits (López & Wong,  
57 2017; Vich et al., 2019). The most prime forms of active mobility include walking and cycling  
58 (Montoya-Robledo et al., 2020; Pajares et al., 2021; Müllers et al., 2022). Despite some research  
59 on active mobility, there is still a significant gap in understanding the relative importance of  
60 various contributing factors, especially in the context of developing metropolises. This study  
61 aims to address this gap by prioritizing the factors influencing active mode choices, providing  
62 valuable insights for policymakers and urban planners.

63 The primary objective of the study is to review the various aspects of active mobility for  
64 sustainable urban transportation. We found a substantial gap in the assessment of factors  
65 influencing active mode choice due to a lack of complete understanding of how infrastructure,  
66 the physical and built environment, safety, and security influence the choice to walk or cycle.  
67 The study emphasizes the necessity for a nuanced approach in policy-making to promote active  
68 modes of transportation. This paper assesses these aspects using fuzzy logic and sensitivity  
69 analysis to quantify the list of factors affecting active mobility. The fuzzy analytic hierarchy  
70 process integrates fuzzy set theory to address the inherent imprecision and subjectivity in human  
71 judgments. This approach is particularly beneficial when evaluating criteria that are challenging  
72 to quantify precisely, such as "comfort" or "aesthetics" in active transportation (Vaidya & Kumar,  
73 2004). Traditional multi-criteria decision-making methods often struggle with such subjective  
74 factors. While the pairwise comparison process can be time-consuming, it can also facilitate a  
75 deeper understanding of the trade-offs involved (Tanwar & Agarwal,2024). In contrast, the  
76 outputs of methods like TOPSIS, COPRAS, and PROMETHEE provide rankings but may not  
77 offer the same level of transparency regarding the relative importance of different criteria.

78 The Fuzzy Analytic Hierarchy Process can effectively address the challenges associated with  
79 subjective and imprecise judgments when evaluating factors influencing active transportation.  
80 Its ability to handle linguistic variables and incorporate fuzzy set theory is advantageous for  
81 capturing the nuanced perceptions of diverse stakeholders. In contexts where multiple  
82 stakeholders with differing perspectives are involved, FAHP can more accurately aggregate  
83 preferences by accounting for the inherent fuzziness in individual judgments. Furthermore, when  
84 expert opinions are central to transportation planning, as is often the case, FAHP can effectively  
85 encapsulate the uncertainty and ambiguity inherent in such expert assessments (Tripathi et al.,  
86 2021). The fuzzy sets provide a mechanism to represent the range of plausible interpretations of  
87 qualitative factors.

88

89 **2. Role of factors for active mobility**

90 In recent years, many researchers have actively studied the role of active mobility in sustainable  
 91 urban transportation (Hackl et al., 2019; Möllers et al., 2022; Pajares et al., 2021; Pisoni et al.,  
 92 2022). Walking and cycling are the primary modes of transportation that promote active mobility  
 93 (López & Wong, 2017). Researchers identified mindset and infrastructural facilitation as key  
 94 supporting attributes for addressing urban mobility issues (Pisoni et al., 2022; Markvica et al.,  
 95 2020; Vich et al., 2019; Zhang et al., 2018). Although the use of cycling has increased, the  
 96 availability of bikes tends to discourage walking as an active mobility transportation mode  
 97 (Scorrano & Danielis, 2021). Gender-based behavior also plays a significant role in the choice  
 98 of transportation modes for active mobility (Montoya-Robledo et al., 2020). Different statistical  
 99 models have been implemented for evidence-based decision-making to assess behavioral  
 100 patterns for walking and cycling to promote active mobility (Hackl et al., 2019).

101 Adlakha and John (2021) promoted cycling as a means of active mobility for India's urban  
 102 streets, emphasizing the need for better urban transportation policies. The lack of active mobility  
 103 in South Asian countries was attributed to government policies favoring motorized transport and  
 104 the failure of local authorities to develop walking and cycling infrastructure. Poor infrastructure  
 105 for bicycling and pedestrians was also identified as a prime reason, especially in Indian regions.

106 Therefore, it was recognized that factors influencing active mobility were vital for the proper  
 107 planning of urban transportation (Hiremath et al., 2013). To maximize the relevance and  
 108 usefulness of factors, practitioners must take into account the intended aims of these factors  
 109 (Hiremath et al., 2013; Saleem & Jaiswal.,2024). Table 1 illustrates the various factors and  
 110 methodologies employed by prior researchers for assessing active mobility in urban  
 111 transportation.

112 **Table1. Summary of factors and methodology involved in South Asian countries for active mobility**

113

Author	Factors considered	Method
Zhang et al., (2018)	Connectivity, Closeness and Spatial distribution, Comfort, Safety	Form based code
Zahraei et al., (2019)	Demographic factors, travel behavior, transportation technologies, macro factors, global drivers	Environmental scanning, Expert interviews, Technology scanning, Focus group discussions, Drivers of change
Wang & Wong (2020)	Walking behavior, built environment, area wise locations, urban heritage	Mixed methods approach (thematic analysis and contingency table analysis)
Das and Banarjee (2023)	Accessibility to urban space, children’s independent mobility and parental perception, Road accidents and child traffic safety	Pilot survey study
Semple & Fountas (2023)	Demographic factors, land use aspect, travel behavior, environmental aspects, economic aspect	Transit oriented development analysis
Chia et al. (2022)	Demographic data, site and nature of the injury, and historical trends	Retrospective chart review of children data; statistical analysis (regression analysis)

114

115 Furthermore, there is a need to examine the interplay between various socio-economic and  
 116 cultural factors influencing active transportation. The factors like individual, physical and built

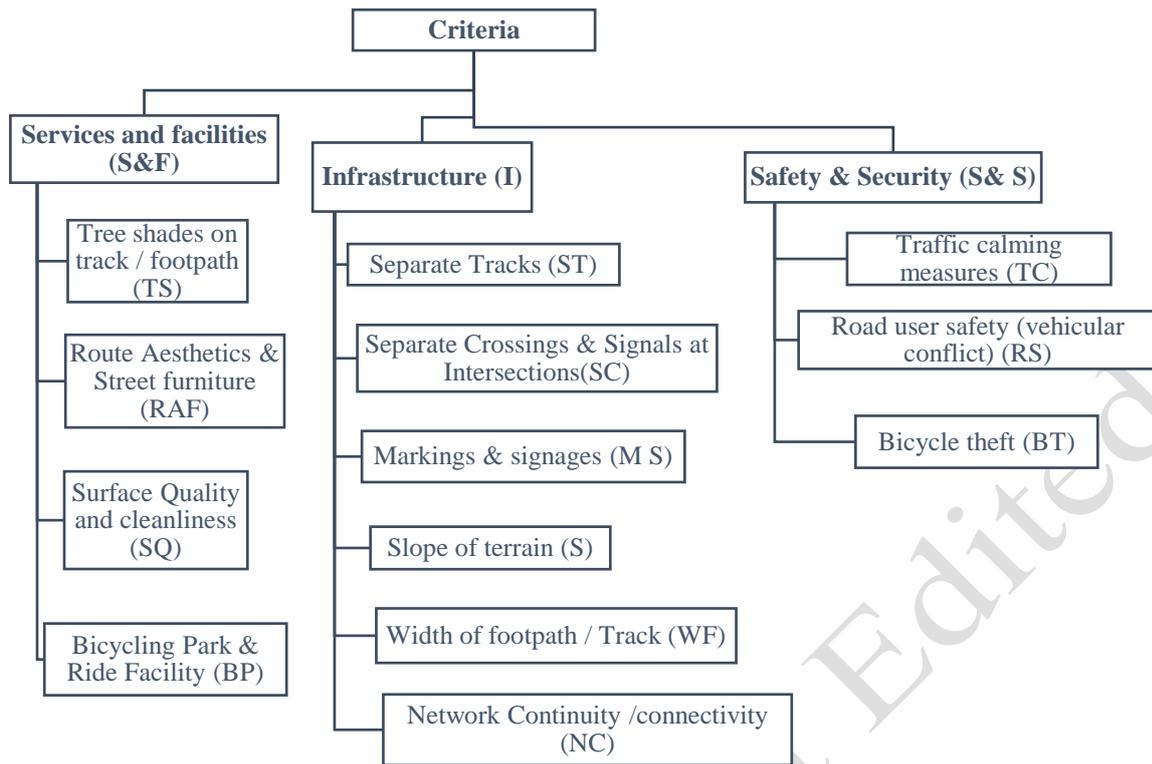
117 environment, neighborhood design and government policies aimed at promoting active travel  
118 impacts differently in varied settings (Winters et al.,2017). This study aims to bridge a crucial  
119 gap in the existing knowledge base by thoroughly examining the multifaceted factors that shape  
120 active transportation within the context of developing urban centres. Recognizing the complex  
121 nature of active travel behaviour, the research at hand centres on three key dimensions: safety  
122 and security, services and facilities, and infrastructure. A well-structured questionnaire survey  
123 was employed to gather the relevant data (Parishad et al.,2021)

124 To address the shortcomings of previous research that often overlooked the nuances of  
125 developing urban environments, this study employs a robust mixed-methods approach. By  
126 systematically gathering and analysing expert opinion, a valuable source of contextualized  
127 knowledge, the relative importance of various factors within each dimension is determined. This  
128 approach offers crucial insights into valuable insights into the unique challenges and  
129 opportunities associated with promoting active transportation in developing metropolises, where  
130 infrastructural limitations and socio-cultural norms often differ significantly from those observed  
131 in developed cities (Parishad et al.,2023).

### 132 3. Methodology

133 The present study aims to prioritize the factors that enable cities to encourage the use of active  
134 modes of travel in order to maintain a balance between their daily commute and health, thereby  
135 promoting sustainable transportation practices. This survey aims to capture the nuanced  
136 understanding of on-the-ground realities and practical challenges associated with promoting  
137 active travel in these contexts. A questionnaire was distributed to a panel of 40 experts. The  
138 expert panel comprised practitioners actively involved in urban transportation planning and  
139 implementation, researchers, and academicians specializing in transportation studies, urban  
140 planning & engineering, and related fields. A structured questionnaire was developed based on  
141 the analytical hierarchy process to elicit expert preferences and judgments regarding the relative  
142 importance of different criteria in a pairwise comparison matrix; it included sections on  
143 infrastructure, services & facilities, safety & security factors. The experts were asked to rate the  
144 importance of one element compared to another on a predefined scale (Vaidya & Kumar, 2004).  
145 The collected expert opinions are then analyzed using the fuzzy analytical hierarchy process  
146 model. This approach allows for the incorporation of inherent uncertainties and ambiguities  
147 associated with expert judgments, leading to a more robust and reliable prioritization of factors  
148 (Miyamoto & Ximenes.,2021)

149 The FAHP model also facilitates sensitivity analysis, examining the robustness of the  
150 prioritization results under different decision-making scenarios. The major categories of factors  
151 were infrastructure, safety, travel behavior, demography, geographical conditions, and weather.  
152 Based on literature review we have segregated into three major categories, Infrastructure (I),  
153 Services and facilities (S&F) and safety and security (S&S) called the criteria. These categories  
154 are further divided into factors. The hierarchic structure adopted for classification is as depicted  
155 in figure 1.



156

157

**Figure 1:** The Hierarchic Structure Adopted for the prioritization of factors

158 The ranking of factors is determined as discussed in section 8 following a work flow  
 159 methodology as described in figure 2.

160

161

162

163

164

165

166

167

168

169

170

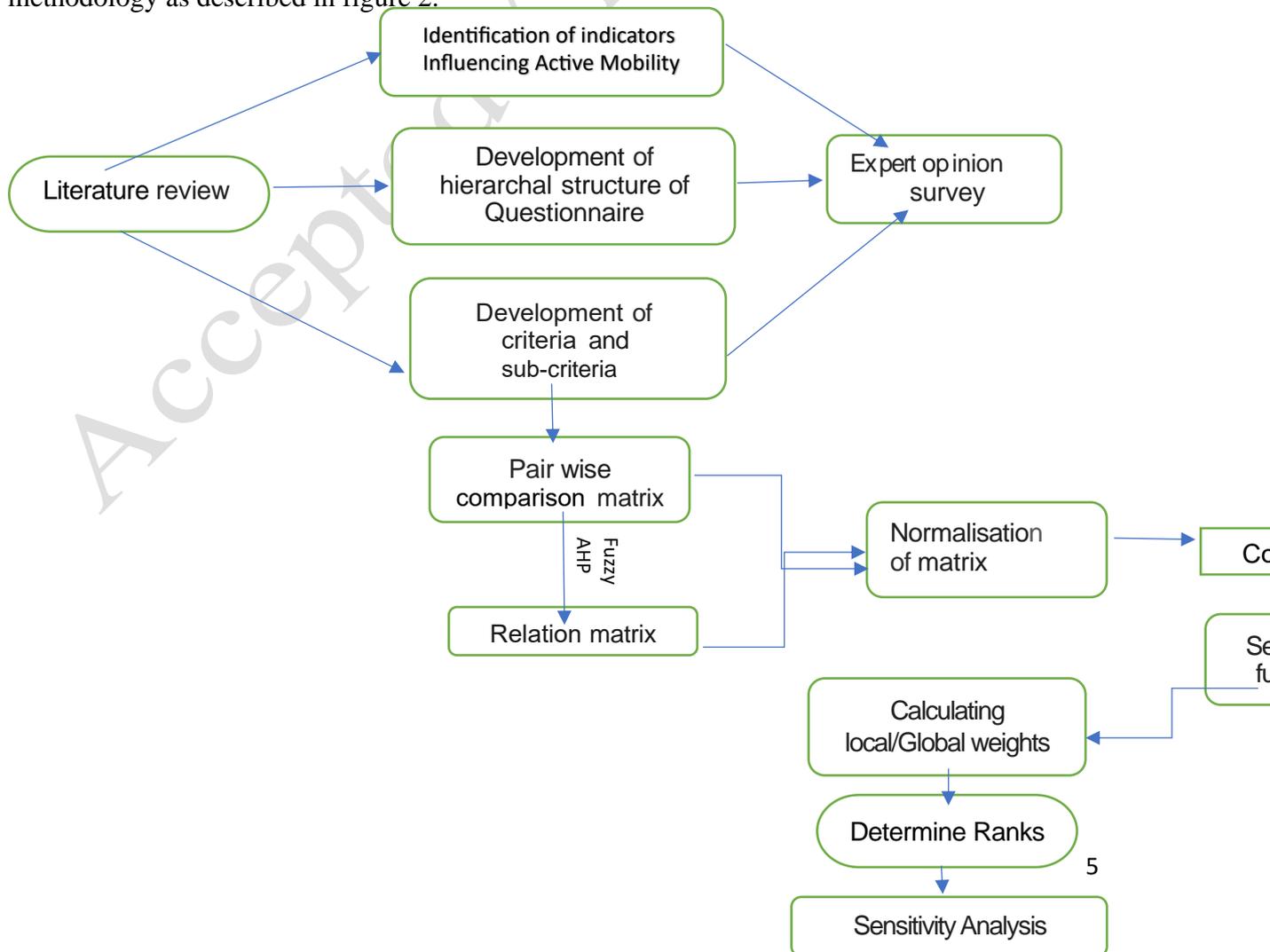
171

172

173

174

175



176

177 **Figure 2:** The work flow process for prioritization of factors for active mobility

178

179 The reliability of a questionnaire is measured by Cronbach's alpha ( $\alpha$ ) value (Eq1) and generally  
180 a questionnaire with a  $\alpha$  value of 0.8 is considered reliable. A value between 0.70 and 0.90 is  
181 considerable (Tavakol & Dennick, 2011). The reliability of the designed questionnaire was  
182 found to be 0.826 which satisfied the minimum criteria. It shows that the rankings are consistent.  
183 Here  $N = 13$  equal the number of study attributes,  $c = 0.30$  corresponds to the average inter-item  
184 covariance between items, and  $v = 0.812$  corresponds to average variance.

185

$$\alpha = \frac{NC}{v+(N-1)c} \quad \text{Eq 1}$$

186

187 In order to evaluate the level of agreement among specialists, Kendall's W coefficient, commonly  
188 known as the coefficient of concordance, was also examined (Eq 2). Finally, the result is 0.78.  
189 Strong levels of unanimity are indicated when the Kendall's coefficient  $W > 0.7$ .

190

$$w = \frac{12R}{m^2(k^3 - k)} \quad \text{Eq 2}$$

191

192

193 Here total  $k = 13$  attributes,  $m = 38$  judges and  $R = 1.61 * 10^5$  is sum of squared deviations and  
194  $W$  is Kendall's coefficient.

195

#### 196 **4. Fuzzy Analytic Hierarchy Process (FAHP)**

197

198 The fuzzy logic theory was used to build the FAHP approach. It allows for the consideration of  
199 multiple criteria and sub criteria providing a comprehensive analysis of various factors  
200 influencing the process. Herein the analytic hierarchy process (AHP) scale is converted into  
201 fuzzy triangular numbers (FTN) scale to access the priority. This is due to the inability of AHP  
202 to handle the randomness in pairwise comparisons (Saaty, 2016). Chang (1996) employed the  
203 fuzzy AHP to address this uncertainty in AHP methodology. Formation of fuzzy-relative  
204 importance matrices for each level of criteria using TFN are illustrated in tables 3, 4 and 5. The  
205 ranks of individual sub-criteria are also calculated for better understanding of priority of factors.  
206 The number of elements should be kept to nine in order to provide adequate consistency when  
207 determining priority from paired comparisons. AHP tolerates inconsistency but gives each set of  
208 judgments a measure of it. The consistency ratio (CR), stated in Equation 3, can be used to  
209 determine the consistency of the judging matrix.

210

$$CR = CI/RI \quad \text{Eq 3}$$

211

212 Where CI is consistency index and RI is Random index. In addition, Saaty (2008) provided  
213 average consistencies (RI values) of randomly generated matrices. CI for a matrix of order  $n$  is  
214 defined in Equation 4 as:

215

$$CI = \frac{\lambda_{max} - n}{n - 1} \quad \text{Eq 4}$$

216

217 CR less than equal to 0.1 is acceptable. If the value is greater, the judgements may be  
218 untrustworthy and should be generated again. Table 2 illustrates that value of all the criteria are  
219 in acceptable range.

220

**Table 2: The Average Consistencies of Random Matrices**

Criteria	Infrastructure	Services and facility	Safety and security
CR	0.09	0.06	0.01

221

#### 222 4.1 Determination of the local and global weights for prioritizing the factors

223 The geometric mean approach was used to calculate the local and global weights of the factors  
224 at each level. Equation 5 can be used to find the geometric mean of the  $i^{th}$  row ( $GM_i$ ) of a crisp  
225 matrix of a corresponding row indicator.

$$GM_i = \left[ \prod_{j=1}^M b_{ij} \right]^{\frac{1}{M}} \quad \text{Eq 5}$$

226

$$w_i = \frac{GM_i}{\sum_{i=1}^n GM_i} \quad \text{Eq 6}$$

227

228

229 Equation 6 can be used to find the local weights, where  $b_{ij}$  in the equation 5 is the value found  
230 in the crisp comparison matrix's  $i^{th}$  row and  $j^{th}$  column. The crisp comparison matrix has  $M$   
231 parameters total. The variable's local weight can be estimated by utilizing equation 6. Having  
232 established the local weights, Equation 7 can be used to calculate the fuzzy global weights ( $G_k$ )  
233 from the local weight of the  $k^{th}$  level and the global weights of the  $(k-1)^{th}$  level.

$$G_k = w_k G_{k-1} \quad \text{Eq 7}$$

234

235 The pairwise comparison matrix for the three criteria based on these equations are as shown in  
236 table 3, 4 and 5.

237

238

239

240

241

242

243  
244  
245  
246  
247  
248

*Accepted / Not Edited*

249

Table 3: Pairwise comparison matrix for Infrastructure Factors

Criteria	Separate Tracks (ST)			Separate Crossings & Signals at Intersections(SC)			Markings & signages (M S)			Slope of terrain (S)			Width of footpath / Track (WF)			Network Continuity /connectivity (NC)		
	Separate Tracks (ST)	1.0 0	1.0 0	1.0 1	1.1 3	1.1 9	1.8 2	2.0 8	2.3 3	1.0 0	1.1 3	1.3 0	0.7 0	0.7 5	0.8 1	0.7 2	0.7 3	0.7 5
Separate Crossings & Signals at Intersections(SC)	0.8 9	0.9 7	1.0 0	1.0 0	1.0 0	0.8 7	0.9 4	1.0 5	0.8 4	1.0 0	1.2 2	0.4 5	0.5 0	0.5 5	0.4 5	0.5 6	0.6 9	0.0 0
Markings & signages (M S)	0.4 8	0.5 5	0.9 6	1.0 6	1.1 5	1.0 0	1.0 0	1.0 0	1.2 4	1.3 4	1.4 2	0.2 6	0.3 0	0.3 5	0.3 8	0.3 9	0.3 9	0.0 0
Slope of terrain (S)	0.8 8	1.0 0	0.8 2	1.0 0	1.1 9	0.7 0	0.7 5	0.8 0	1.0 0	1.0 0	1.0 0	0.6 1	0.6 7	0.7 2	0.5 3	0.5 6	0.5 9	0.0 0
Width of footpath / Track (WF)	1.3 3	1.4 2	1.8 2	2.0 1	2.2 2	2.8 3	3.3 6	3.9 0	1.4 0	1.5 0	1.6 3	1.0 0	1.0 0	1.0 0	0.8 2	0.8 9	0.9 7	0.0 0
Network Continuity /connectivity (NC)	1.3 6	1.4 0	1.4 4	1.7 8	2.2 2	2.5 4	2.5 8	2.6 2	1.7 0	1.7 9	1.9 0	1.0 3	1.1 3	1.2 2	1.0 0	1.0 0	1.0 0	0.0 0

250

251

Table 4: Pairwise comparison matrix for Services and facilities

Criteria	Bicycling Park & Ride Facility( BP)	Surface Quality and cleanliness	Route Aesthetics & Street furniture	Tree shades on track / footpath
Bicycling Park & Ride Facility( BP)	1	1	1	1
Surface Quality and cleanliness (SQ)	2.5	2.92	3.409	1
Route Aesthetics & Street furniture (RAF)	1.3	1.35	1.442	0.4
Tree shades on track / footpath (TS)	0.9	0.94	1.046	0.42

252

253

254

Table 5: Pairwise comparison matrix for safety and security

Criteria	Road user safety ( vehicular conflict ) (RS)			traffic calming measures (TC)			Bicycle theft (BT)		
	Road user safety ( vehicular conflict ) (RS)	1	1	1.01	1.1	1.2	1.8	2.1	2.3
traffic calming measures (TC)	0.9	0.97	1	1	1	0.9	0.9	1	0.8
Bicycle theft (BT)	0.5	0.6	0.96	1.1	1.2	1	1	1	1.2

255

256

257

Accepted / Not Accepted

258

259

### 260 5. Sensitivity Analysis (SA)

261 Sensitivity analysis is used to account for uncertainties in input data, processing, criterion  
 262 selection, and external factors beyond the decision maker's control (Nyimbili & Erden, 2020).  
 263 SA was done after FAHP to ensure that relative weights were kept when projected to data  
 264 variances, resulting in more accurate decision-making. By adjusting the fuzzification factor, the  
 265 proposed decision-making model underwent a sensitivity study. By altering one input factor at a  
 266 time while holding the other factors constant, it was used to monitor the criterion weight  
 267 sensitivity and analyse the consequences on the model outputs (Saltelli et al., 2006). Equation  
 268 9.1 shows how the weight of the other criteria,  $w_j$ , would change if the weight of the  $i^{th}$   
 269 (criterion/study attribute) was altered from  $w_i^0$  to  $w_i$ .

$$w_j = \frac{(1-w_i)}{(1-w_i^0)} w_j^0 \quad \text{Eq 8}$$

270

271 where  $w_j$  is the new weight value of the other (criterion) to be changed;  $w_i^0$  and  $w_j^0$  were the  
 272 initial weight values of the criteria before being subjected to SA. Table shows an analysis of the  
 273 model output for six sets (0, 0.2, 0.4, 0.6, 0.8 and 1) and the relative importance of the attributes  
 274 was monitored shown Table 6.

275

276

Table 6: Sensitivity Analysis for factors

Criteria	Factors	Original	Set-1	Set-2	Set-3	Set-4	Set-5
Infrastructure (I)	(ST)	0.1217	0.1133	0.1049	0.0965	0.1301	0.1217
	(SC)	0.0903	0.0841	0.0779	0.0717	0.0966	0.0903
	(M S)	0.0750	0.0699	0.0647	0.0595	0.0802	0.0750
	(S)	0.0905	0.0842	0.0780	0.0718	0.0967	0.0905
	(WF)	0.1731	0.1612	0.1492	0.1373	0.1850	0.1731
	(NC)	0.1749	0.1628	0.1508	0.1387	0.1869	0.1749
Services and facilities (SF)	(BP)	0.0291	0.0332	0.0373	0.0414	0.0250	0.0291
	(SQ)	0.0782	0.0892	0.1001	0.1111	0.0672	0.0782
	(RAF)	0.0429	0.0490	0.0550	0.0610	0.0369	0.0429
	(TS)	0.0277	0.0315	0.0354	0.0393	0.0238	0.0277
Safety & Security (SS)	(RS)	0.0571	0.0719	0.0867	0.1015	0.0424	0.0571
	(TC)	0.0248	0.0312	0.0376	0.0440	0.0184	0.0248
	(BT)	0.0147	0.0185	0.0223	0.0261	0.0109	0.0147

277

278 For every set of categories and factors, separate fuzzification factors were used to produce the  
 279 fuzzy pair-wise comparison matrices. Based on global weights, an indicator's rank is  
 280 determined. The rank or priority increases with the global weight of the indicator. Figure 3  
 281 highlights the sensitivity of the decision-making outcomes for all other factors.

282

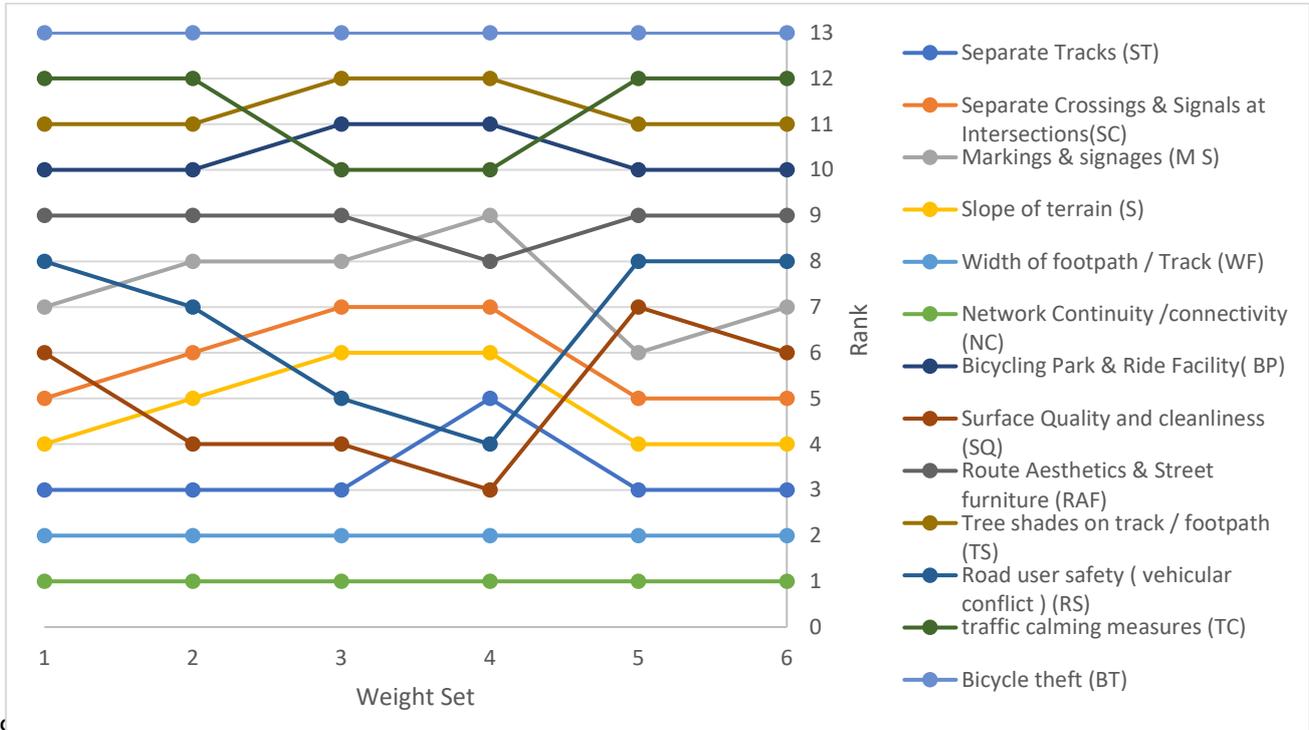
Table 7: Sensitivity analysis for indicator criteria

Criteria	Original Set of Weights	Set-1	Set-2	Set-3	Set-4	Set-5
(I)	72.55%	67.55%	62.55%	57.55%	77.55%	72.55%
(SF)	17.78%	20.28%	22.78%	25.28%	15.28%	17.78%
(SS)	9.67%	12.17%	14.67%	17.17%	7.17%	9.67%

	100.00%	100.00%	100.00%	100.00%	100.00%	100.00%
--	---------	---------	---------	---------	---------	---------

283

284 Table 7 indicates the prioritization of the factor’s weights through different scenarios. The  
 285 variations in weight across different factors demonstrated the sensitivity of the decision model  
 286 to change in the relative importance of factors. Infrastructure is of highest priority as the  
 287 variations in weights did not alter the position of the indicator followed by Services and facilities.  
 288



289

290

Figure 3: Sensitivity analysis for the factors

291 **6. Results and discussion**

292 Active mobility is a measure to attain sustainability in the transportation system. To achieve this  
 293 an understanding of its factors and their priority levels is indispensable for planners and policy  
 294 makers. It will help them prioritize the challenges and demands that needs to be met for  
 295 promoting sustainable transportation practices. The findings are helpful in assisting with the  
 296 decision-making process for the creation of an urban bicycle infrastructure plan that incorporates  
 297 multimodal active transportation infrastructures mechanized pedestrian mobility (Zhang &  
 298 Zhou, 2023). The analysis provides an insight into factors that may encourage users to switch to  
 299 active mode of travel. Table 8 Shows the pairwise comparison matrix for indicators of  
 300 Infrastructure(I), Services and facilities (SF) and Safety and security (SS) respectively.

301

Table 8 : Pairwise comparison matrix of Criteria

302

303

304

305

9	I			SF			SS		
I	1	1	1	3.57	4.095	4.58	7.18	7.533	8.12
SF	0.2184	0.244173	0.2804	1	1	1	1.54	1.859	2.16
SS	0.1248	0.13275	0.1437	0.46	0.538	0.65	1	1	1

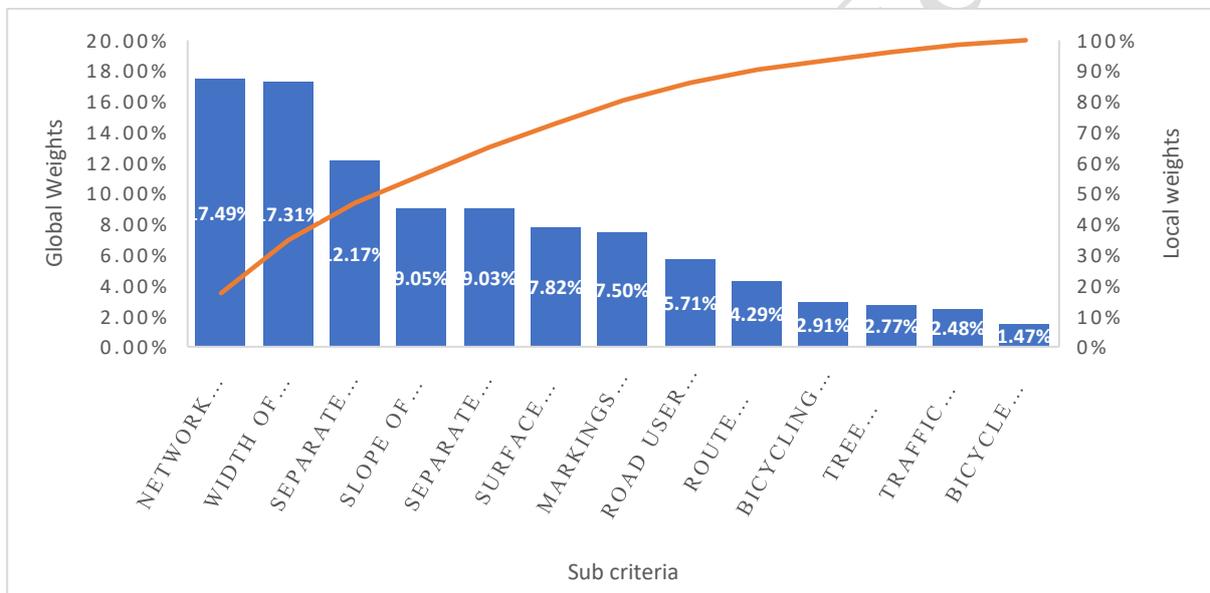
306

307

It was found that the primary factors that affect the mode choice of users depend on infrastructure. Network continuity, width of footpath and separate tracks are the top most

308 priorities under the category of infrastructure as shown in figure 4. As the changes alongside  
 309 the route networks will influence the travel demand. Surface quality & cleanliness and road  
 310 safety are next in the list as reflected in the analysis of expert opinion survey criteria wise as  
 311 depicted in table 9. The safety and infrastructure issue includes concerns about road safety, bike  
 312 lanes, and connection. (Piatkowski, et al., 2015). The continuity of network is regarded as the  
 313 top most priority for commuters this reflects that the association between network  
 314 characteristics and ridership is an indispensable aspect of transportation planning and policy  
 315 and thus they play a significant role in influencing the commuting behaviour (Beck, et al.,  
 316 2023).

317 Under the category of services and facilities, factors such as route aesthetics, street furniture,  
 318 and surface quality are also major considerations. For instance, greenery and barriers are  
 319 positively correlated with increased cycling activity while streetlights and signals show a  
 320 negative correlation with speed and trip density (Wang & Wong, 2020). Safe and secure  
 321 walking environment are among the priority factors for pedestrians (Bivina & Parida, 2020) As  
 322 per the research to create walk/bicycle friendly cities and enhance sustainable transportation  
 323 practices it is essential for planners to understand the relation between various factors.



324

325 Figure 4: Prioritization of factors influencing active mode choices

326

Table 9: Sensitivity Analysis and Prioritization of Factors

Ranks	Original (1)	Set-1 (2)	Set-2 (3)	Set-3 (4)	Set-4 (5)	Set-5 (6)
Separate Tracks (ST)	3	3	3	5	3	3
Separate Crossings & Signals at Intersections(SC)	5	6	7	7	5	5
Markings & signages (M S)	7	8	8	9	6	7
Slope of terrain (S)	4	5	6	6	4	4
Width of footpath / Track (WF)	2	2	2	2	2	2
Network Continuity /connectivity (NC)	1	1	1	1	1	1
Bicycling Park & Ride Facility (BP)	10	10	11	11	10	10
Surface Quality and cleanliness (SQ)	6	4	4	3	7	6
Route Aesthetics & Street furniture (RAF)	9	9	9	8	9	9
Tree shades on track / footpath (TS)	11	11	12	12	11	11

Road user safety (vehicular conflict) (RS)	8	7	5	4	8	8
traffic calming measures (TC)	12	12	10	10	12	12
Bicycle theft (BT)	13	13	13	13	13	13

327

328 In heavily populated places, (Din, et al., 2023) advise giving sustainable transportation policies  
329 and solutions top priority. While addressing the issues brought on by population density, policies  
330 like encouraging public transportation and electric cars and making investments in infrastructure  
331 that facilitates active transportation modes like walking and cycling can help reduce the adverse  
332 environmental impacts of transportation. Factors influencing walking and cycling mode vary ,  
333 with trip characteristics and built environment having the highest impact on active mode use (Ton,  
334 et al.,2019) Although The prioritization of factors helps in investment decision and prevents  
335 wastage /underutilisation of available resources (Heidari, et al., 2023) but to promote sustainable  
336 transportation practices educational interventions are necessary as they will help in changing the  
337 mindset and attitude of children and adolescents towards active transportation. This will also  
338 help in eliminating the negative mindsets associated with active mode choices  
339 (Chanpariyavatevong, et al., 2024)

340 Semple & Fountas, (2023) recommend that prior to implementation, pedestrianization should be  
341 studied locally, city by city, or town by town. This enables plans to be customized to the demands  
342 of the community or to the constraints of the current infrastructure. For the majority of instances,  
343 exclusive lanes for pedestrians, and their interactions on MFD (Macroscopic Fundamental  
344 Diagram) are important. While the particular impact of these variables on a particular network  
345 depends on its distinct configuration (Liu, et al., 2024). There isn't a single solution available that  
346 addresses all facet of urban and transportation-related problems. Prior until now, combined  
347 solutions might appear to be the most appropriate for the future (Sarri, et al., 2024). Cities with  
348 similar travel characteristics could be investigated for the effect of land use on travel behaviour  
349 and different active travel scenarios could be modelled.

350 Promoting active transportation in developing metropolises requires a multifaceted approach.  
351 Crucial elements include investing in dedicated cycling infrastructure, enhancing connectivity  
352 between existing routes, and widening pedestrian walkways. This may involve reallocating road  
353 space, creating separated cycling lanes, and improving pedestrian crossings. Additionally,  
354 addressing the mobility challenges faced by vulnerable groups, such as the elderly, people with  
355 disabilities, and lower-income populations, through inclusive design and targeted interventions  
356 is essential.

357 Manageable slopes are particularly important for active travel, especially for cyclists and  
358 individuals with limited mobility. Urban design features, such as street furniture, greenery, and  
359 aesthetically pleasing routes, can enhance the overall walking and cycling experience, thereby  
360 encouraging greater mode shift towards active transportation. Furthermore, education and  
361 awareness campaigns to highlight the benefits of active mobility, coupled with behavioral  
362 nudges, can help foster a culture of sustainable transportation. Cities should also prioritize the  
363 development of secure bicycle parking near transit stations and plant trees along active  
364 transportation routes to provide shade and improve aesthetic appeal. In conclusion, a  
365 comprehensive strategy addressing the built environment, transportation infrastructure,  
366 inclusive design, and behavioral change is necessary to effectively promote active  
367 transportation in developing metropolises.

368

## 369 7.Conclusion

370

371 The research emphasizes that urban streets are a major component of the built environment. This  
372 research provides a nuanced understanding of the factors influencing active travel choices in  
373 developing metropolises, making a unique contribution to the field of sustainable urban  
374 transportation. While prior studies have often focused on developed cities, this research delves  
375 into the unique challenges and opportunities of developing urban environments. By  
376 systematically analyzing expert opinions, this study goes beyond merely identifying relevant  
377 factors; it offers a contextualized understanding of their relative importance, providing valuable  
378 insights for practitioners and policymakers. This perspective is crucial for developing effective  
379 strategies to promote active transportation in developing metropolises, where infrastructural  
380 limitations, socio-cultural norms, and travel patterns often differ significantly from developed  
381 cities. The findings contribute in advancing our understanding of active travel behavior in  
382 understudied contexts and offering actionable insights for creating more sustainable urban  
383 environments.

384

385 The study prioritized the factors that matter for improving the quality of service for active  
386 transportation and thus analyzed the sensitivity in decision-making for the justification of  
387 robustness in the selection of factors using the FAHP model. It was noted that the rank of the  
388 highest priority indicator never alters with a change in the fuzzification factor or the decision  
389 altitude. It can be inferred that the rank of these factors follows a similar pattern irrespective of  
390 the fuzzification factors. The analysis indicated that infrastructure is a major area of focus for  
391 promoting active mobility, followed by services and facilities required for cycling and walking.  
392 The investigations reflect that network continuity in the infrastructure category is the most  
393 influential of all the criteria for encouraging users to switch to active mobility. The top five  
394 factors that majorly influence the active mode choice according to the FAHP model study are  
395 network continuity, width of footpath, separate signals at intersections and crossings, and slope  
396 of terrain.

397

398 There are numerous research possibilities in this area, as very little work has been done, and  
399 little of what has been done focused only on providing ramps without understanding user  
400 behavior and requirements regarding prioritization of various conditions. Such provision of  
401 facilities discourages even the potential users, as they lack safety, comfort, and spatial  
402 integration. Therefore, the existing systems need a proper framework with well-marked factors  
403 to attain sustainability in the transport sector. Longitudinal studies could also be conducted to  
404 track changes in active mobility patterns over time in response to policy intervention and  
405 infrastructural developments. Assessing user behavior and preferences regarding mode choices  
406 and how different demographics influence the mode choices and perceive the level of services  
407 for active modes may help in aiming to achieve sustainability in the transportation sector. This  
408 may lower its subsequent impact on the environment and public health. The available analysis  
409 of prioritization may help formulate a framework to help the planners and policymakers define  
410 the level of service for active transportation and make informed choices about resource allocation  
411 and policy implementation.

412

#### 413 **Data availability**

414 The data that support the findings of this study are available from the corresponding author, upon  
415 reasonable request.

#### 416 **Conflict of Interest**

417 We hereby declare that the authors have no actual and potential conflict of interest.

418

## 41References

420

421 Adlakha, D. and John, F. (2021) 'Push for pedal power: urban mobility and the rise of bicycling in  
422 Indian cities', in Siefken, K., Ramirez Varela, A., Waqanivalu, T. and Schultenkorf, N. (eds.)  
423 *Physical Activity in Low- and Middle-Income Countries*. 1st edn. London: Routledge, pp. 217–224.  
424 <https://doi.org/10.4324/9780429344732>.

425 Bivina, G. R. and Parida, M. (2020) 'Prioritizing pedestrian needs using a multi-criteria decision  
426 approach for a sustainable built environment in the Indian context', *Environment, Development and*  
427 *Sustainability*, 22, pp. 4929–4950. <https://doi.org/10.1007/s10668-019-00381-w>.

428 Chang, D.-Y. (1996) 'Applications of the extent analysis method on fuzzy AHP', *European Journal*  
429 *of Operational Research*, 95, pp. 649–655. [https://doi.org/10.1016/0377-2217\(95\)00300-2](https://doi.org/10.1016/0377-2217(95)00300-2).

430 Chanpariyavatevong, K. et al. (2024) 'The intervention design to promote active travel mode among  
431 children and adolescents: A systematic review', *Heliyon*, 10, p. e26072.  
432 <https://doi.org/10.1016/j.heliyon.2024.e26072>.

433 Chia, A. Z. H., Chong, S. L. and Ganapathy, S. (2022) 'Non-motorised active mobility device use  
434 by children in Singapore: Injury patterns and risk factors for severe injury', *Annals of the Academy*  
435 *of Medicine, Singapore*, 51, pp. 531–539. <https://doi.org/10.47102/annals-acadmedsg.202247>.

436 Das, R. and Banerjee, A. (2023) 'Identifying the parameters for assessment of child-friendliness in  
437 urban neighborhoods in Indian cities', *Journal of Urban Affairs*, 45, pp. 217–235.  
438 <https://doi.org/10.1080/07352166.2020.1863815>.

439 Din, A. U. et al. (2023) 'Green road transportation management and environmental sustainability:  
440 The impact of population density', *Heliyon*, 9, p. e19771.  
441 <https://doi.org/10.1016/j.heliyon.2023.e19771>.

442 Pisoni, E., Christidis, P. and Navajas Cawood, E. (2022) 'Active mobility versus motorized  
443 transport? User choices and benefits for the society', *Science of The Total Environment*, 806(2), p.  
444 150627. <https://doi.org/10.1016/j.scitotenv.2021.150627>.

445 Hackl, R. et al. (2019) 'Promoting active mobility: Evidence-based decision-making using  
446 statistical models', *Journal of Transport Geography*, 80, p. 102541.  
447 <https://doi.org/10.1016/j.jtrangeo.2019.102541>.

448 Heidari, I., Eshlaghy, A. T. and Hoseini, S. M. S. (2023) 'Sustainable transportation: Definitions,  
449 dimensions, and factors – Case study of importance-performance analysis for the city of Tehran',  
450 *Heliyon*, 9, p. e20457. <https://doi.org/10.1016/j.heliyon.2023.e20457>.

451 Hiremath, R. B. et al. (2013) 'Indicator-based urban sustainability—A review', *Energy for*  
452 *Sustainable Development*, 17, pp. 555–563. <https://doi.org/10.3390/su13169348>.

453 Liu, H., Xiong, Z. and Gayah, V. V. (2024) 'Quantifying the impacts of right-turn-on-red, exclusive  
454 turn lanes and pedestrian movements on the efficiency of urban transportation networks',  
455 *International Journal of Transportation Science and Technology*.  
456 <https://doi.org/10.1016/j.ijst.2024.02.007>.

457 López, M. C. R. and Wong, Y. D. (2017) 'Attitudes towards active mobility in Singapore: A  
458 qualitative study', *Case Studies on Transport Policy*, 5, pp. 662–670.  
459 <https://doi.org/10.1016/j.cstp.2017.07.002>.

460

- 461 Markvica, K., Millionig, A., Haufe, N., & Leodolter, M. (2020) 'Promoting active mobility behavior  
462 by addressing information target groups: The case of Austria', *Journal of Transport Geography*, 83,  
463 p. 102664. <https://doi.org/10.1016/j.jtrangeo.2020.102664>.
- 464 Möllers, A., Specht, S., & Wessel, J. (2022) 'The impact of the Covid-19 pandemic and government  
465 intervention on active mobility', *Transportation Research Part A: Policy and Practice*, 165, pp.  
466 356–375. <https://doi.org/10.1016/j.tra.2022.09.007>
- 467 Montoya-Robledo, V. et al. (2020) 'Gender stereotypes affecting active mobility of care in Bogotá',  
468 *Transportation Research Part D: Transport and Environment*, 86, p. 102470.  
469 <https://doi.org/10.1016/j.trd.2020.102470>.
- 470 Miyamoto, A. and Ximenes, H. D. C. (2021) 'Development of a road-condition assessment system  
471 and application to road maintenance decision-making', *Civil Engineering Infrastructures Journal*,  
472 54(2), pp. 225-251. <https://doi.org/10.22059/CEIJ.2021.294057.1642>.
- 473 Nyimbili, P. H. and Erden, T. (2020) 'GIS-based fuzzy multi-criteria approach for optimal site  
474 selection of fire stations in Istanbul, Turkey', *Socio-Economic Planning Sciences*, 71, p. 100860.  
475 <https://doi.org/10.1016/j.seps.2020.100860>.
- 476 Vaidya, O. S. and Kumar, S. (2006) 'Analytic hierarchy process: An overview of applications',  
477 *European Journal of Operational Research*, 169(1), pp. 1-29.  
478 <https://doi.org/10.1016/j.ejor.2004.04.028>.
- 479 Parishad, N. et al. (2020) 'Validation of the driver behavior questionnaire in a  
480 representative sample of Iranian drivers', *Civil Engineering Infrastructures Journal*, 53(1),  
481 pp. 161-171. <https://doi.org/10.22059/cej.2019.283780.1593>.
- 482 Parishad, N. et al. (2023) 'Validation of Modified Driver Behavior Questionnaire  
483 Considering Mobile Phone Usage While Driving', *Civil Engineering Infrastructures  
484 Journal*, 56(1), pp. 19-32. <https://doi.org/10.22059/cej.2021.327968.1778>.
- 485 Pajares, E. et al. (2021) 'Accessibility by proximity: Addressing the lack of interactive  
486 accessibility instruments for active mobility', *Journal of Transport Geography*, 93, p.  
487 103080. <https://doi.org/10.1016/j.jtrangeo.2021.103080>.
- 488 Piatkowski, D. et al. (2015) 'Measuring the impacts of bike-to-work day events and  
489 identifying barriers to increased commuter cycling', *Journal of Urban Planning and  
490 Development*, 141, p. 04014034. [https://doi.org/10.1061/\(ASCE\)UP.1943-5444.0000239](https://doi.org/10.1061/(ASCE)UP.1943-5444.0000239).
- 491 Pisoni, E., Christidis, P. and Cawood, E. N. (2022) 'Active mobility versus motorized  
492 transport? User choices and benefits for the society', *Science of the Total Environment*,  
493 806, p. 150627. <https://doi.org/10.1016/j.scitotenv.2021.150627>.
- 494 Rainieri, G. et al. (2024) 'Evaluating active mobility: enhancing the framework for social  
495 sustainability', *TeMA - Journal of Land Use, Mobility and Environment*, (3), pp. 113-128.  
496 <https://doi.org/10.6093/1970-9870/10912>.
- 497 Saaty, T. L. (2008) 'Decision making with the analytic hierarchy process', *International  
498 Journal of Services Sciences*, 1, pp. 83–98. <https://doi.org/10.1504/IJSSCI.2008.017590>.

- 499 Saaty, T. L. (2016) ‘The analytic hierarchy and analytic network processes for the  
500 measurement of intangible criteria and for decision-making’, *Multiple Criteria Decision*  
501 *Analysis: State of the Art Surveys*, pp. 363–419. [https://doi.org/10.1007/978-1-4939-3094-](https://doi.org/10.1007/978-1-4939-3094-4_10)  
502 [4\\_10](https://doi.org/10.1007/978-1-4939-3094-4_10).
- 503 Said, M., Biehl, A. and Stathopoulos, A. (2022) ‘Interdependence in active mobility  
504 adoption: Joint modeling and motivational spillover in walking, cycling and bike-sharing’,  
505 *International Journal of Sustainable Transportation*, 16, pp. 422–440.  
506 <https://doi.org/10.1080/15568318.2021.1885769>.
- 507 Saleem, S. and Jaiswal, A. (2024) ‘Indicators for Active Transportation in Tier II Indian  
508 Cities: A Case of Bhopal, India’, *Journal of Advanced Transportation*, 2024(1), p.  
509 2175645. <https://doi.org/10.1155/atr/2175645>.
- 510 Saltelli, A. et al. (2006) ‘Sensitivity analysis practices: Strategies for model-based  
511 inference’, *Reliability Engineering & System Safety*, 91, pp. 1109–1125.  
512 [https://doi.org/10.1007/978-1-4939-3094-4\\_10](https://doi.org/10.1007/978-1-4939-3094-4_10).
- 513 Sarri, P. et al. (2024) ‘Incorporating Land Use and Transport Interaction Models to  
514 Evaluate Active Mobility Measures and Interventions in Urban Areas: A case study in  
515 Southampton, UK’, *Sustainable Cities and Society*, 105, p. 105330.  
516 <https://doi.org/10.1016/j.scs.2024.105330>.
- 517 Scorrano, M. and Danielis, R. (2021) ‘Active mobility in an Italian city: Mode choice  
518 determinants and attitudes before and during the Covid-19 emergency’, *Research in*  
519 *Transportation Economics*, 86, p. 101031. <https://doi.org/10.1016/j.retrec.2021.101031>.
- 520 Scorza, F. and Fortunato, G. (2021) ‘Cyclable cities: building feasible scenario through  
521 urban space morphology assessment’, *Journal of Urban Planning and Development*, 147,  
522 p. 05021039. [https://doi.org/10.1061/\(ASCE\)UP.1943-5444.0000713](https://doi.org/10.1061/(ASCE)UP.1943-5444.0000713).
- 523 Semple, T. and Fountas, G. (2023) ‘Demographic and behavioural factors affecting public  
524 support for pedestrianisation in city centres: The case of Edinburgh, UK’, *International*  
525 *Journal of Transportation Science and Technology*, 12, pp. 103–118.  
526 <https://doi.org/10.1016/j.ijtst.2021.12.001>.
- 527 Stuchi, S., Paulino, S. and Gallouj, F. (2022) ‘Social Innovation in Active Mobility Public  
528 Services in the Megacity of Sao Paulo’, *Sustainability*, 14, p. 11834.  
529 <https://doi.org/10.3390/su141911834>.
- 530 Sundfør, H. B. and Fyhri, A. (2022) ‘The effects of a subvention scheme for e-bikes on  
531 mode share and active mobility’, *Journal of Transport & Health*, 26, p. 101403.  
532 <https://doi.org/10.1016/j.jth.2022.101403>.
- 533 Tanwar, R. and Agarwal, P. K. (2024) ‘Assessing travel time performance of multimodal  
534 transportation systems using fuzzy-analytic hierarchy process: A case study of Bhopal  
535 City’, *Heliyon*, 10(17), p. e36844. <https://doi.org/10.1016/j.heliyon.2024.e36844>.

- 536 Tripathi, A. K., Agrawal, S. and Gupta, R. D. (2022) ‘Comparison of GIS-based AHP and  
537 fuzzy AHP methods for hospital site selection: a case study for Prayagraj City, India’,  
538 *GeoJournal*, 87, pp. 3507–3528. <https://doi.org/10.1007/s10708-021-10445-y>.
- 539 Tavakol, M. and Dennick, R. (2011) ‘Making sense of Cronbach's alpha’, *International*  
540 *Journal of Medical Education*, 2, p. 53. <https://doi.org/10.5116/ijme.4dfb.8dfd>.
- 541 Vich, G., Marquet, O. and Miralles-Guasch, C. (2019) ‘Green streetscape and walking:  
542 exploring active mobility patterns in dense and compact cities’, *Journal of Transport &*  
543 *Health*, 12, pp. 50–59. <https://doi.org/10.1016/j.jth.2018.11.003>.
- 544 Wang, C. et al. (2023) ‘Exploring the spatially and temporally varying impacts of built  
545 environment factors on rail transit ridership: A case study in Nanjing, China’, *Research*  
546 *Square*. <https://doi.org/10.21203/rs.3.rs-2881084/v1>.
- 547 Wang, Y. and Wong, Y. D. (2020) ‘Repositioning urban heritage for active mobility:  
548 Indications from news coverage in Singapore’, *Cities*, 98, p. 102525.  
549 <https://doi.org/10.1016/j.cities.2019.102525>.
- 550 Winters, M., Buehler, R. and Götschi, T. (2017) ‘Policies to promote active travel: evidence  
551 from reviews of the literature’, *Current Environmental Health Reports*, 4, pp. 278-285.  
552 <https://doi.org/10.1007/s40572-017-0148-x>.
- 553 Woodson, C. et al. (2024) ‘Factors influencing bikeshare service and usage in a rural  
554 college town: A case study of Montgomery County, VA’, *International Journal of*  
555 *Sustainable Transportation*, 4(18), pp. 291-300.  
556 <https://doi.org/10.1080/15568318.2023.2295865>.
- 557 Zahraei, S. M., Kurniawan, J. H. and Cheah, L. (2019) ‘A foresight study on urban  
558 mobility: Singapore in 2040’, *Foresight*, 22, pp. 37–52. <https://doi.org/10.1108/FS-05-2019-0044>.
- 560 Zhang, X. and Zhou, H. (2023) ‘Indicators of Urban Conditions: An Approach to  
561 Understanding the Suitability of the Compact City Paradigm’, *Journal of Urban Planning*  
562 *and Development*, 149(3), p. 04023021. <https://doi.org/10.1061/JUPDDM.UPENG-4150>.
- 563 **Zhang, Y., Azzali, S., Janssen, P. and Stouffs, R.** (2018) ‘Design for walkable  
564 neighbourhoods in Singapore using Form-based Codes’, *IFoU 2018: Reframing Urban*  
565 *Resilience Implementation: Aligning Sustainability and Resilience*, Urban Design and  
566 Management: Infrastructures and Services, National University of Singapore, 10–12  
567 December. MDPI. <https://doi.org/10.3390/IFOU2018-05934>.

568

569

## 570 Annexure

571 The following Annexure highlights the structure of questionnaire survey conducted for this study and  
572 the scale of relative importance.

573

574 Scale of relative importance

1	Equal importance
3	Moderate Importance
5	Strong Importance
7	Very strong Importance
9	Extremely strong importance
2,4,6,8	Intermediate values

575 Example for marking

9 - EXTREMELY IMPORTANT

7 - VERY STRONGLY IMPORTANT

5 - STRONGLY IMPORTANT

3 - MODERATELY IMPORTANT

1 - EQUAL IMPORTANCE

3 - MODERATELY IMPORTANT

5 - STRONGLY IMPORTANT

7 - VERY STRONGLY IMPORTANT

9 - EXTREMELY IMPORTANT

Parameter A Score ← → Score Parameter B

Example showing how to write scores.

If you consider Variable A to be of strong importance compared to Variable B, then write "5" in the box next to Variable A.

Variable A  9 7 5 3 1 3 5 7 9  Variable B

If you consider Variable A to be of equal importance compared to Variable B, then write "1" in the box next to Variable A and B.

Variable A  9 7 5 3 1 3 5 7 9  Variable B

If you consider Variable B to be of extreme importance compared to Variable A, then write "9" in the box next to Variable B.

Variable A  9 7 5 3 1 3 5 7 9  Variable B

576

577

578

579

580 A. Infrastructure

Criteria 1	9	8	7	6	5	4	3	2	1	2	3	4	5	6	7	8	9	Criteria 2
Separate Tracks																		Separate Crossings & Signals at Intersections
Separate Tracks																		Markings & signages
Separate Tracks																		Slope of terrain
Separate Tracks																		Width of footpath / Track
Separate Tracks																		Network Continuity /connectivity
Separate Crossings & Signals at Intersections																		Markings & signages
Separate Crossings & Signals at Intersections																		Slope of terrain
Separate Crossings & Signals at Intersections																		Width of footpath / Track

Signals at Intersections																			
Separate Crossings & Signals at Intersections																			Network Continuity /connectivity
Markings & signages																			Slope of terrain
Markings & signages																			Width of footpath / Track
Markings & signages																			Network Continuity /connectivity
Slope of terrain																			Width of footpath / Track
Slope of terrain																			Network Continuity /connectivity
Width of footpath / Track																			Network Continuity /connectivity

581

582 **B. Services and facilities**

583

Criteria 1	9	8	7	6	5	4	3	2	1	2	3	4	5	6	7	8	9	Criteria 2
Bicycling Park & Ride Facility																		Surface Quality and cleanliness
Bicycling Park & Ride Facility																		Route Aesthetics & Street furniture
Bicycling Park & Ride Facility																		Tree shades on track / footpath
Surface Quality and cleanliness																		Route Aesthetics & Street furniture
Surface Quality and cleanliness																		Tree shades on track / footpath
Tree shades on track / footpath																		Route Aesthetics & Street furniture

584

585 **C. Safety & Security**

Criteria 1	9	8	7	6	5	4	3	2	1	2	3	4	5	6	7	8	9	Criteria 2
Road user safety (vehicular conflict)																		Traffic calming measures
Road user safety (vehicular conflict)																		Bicycle theft
Traffic calming measures																		Bicycle theft

586