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#### Тне RELATIONSHIP BETWEEN STREET NETWORK MORPHOLOGY AND PERCENTAGE OF DAILY TRIPS ON FOOT AND BY BICYCLE AT THE CITY-LEVEL

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# Graphical abstract



#### Abstract

There are various factors that affect walking and cycling in urban areas, such as density, diversity and design, but there have been few studies that examine the relationship between urban morphology factors such as street network and green travel modes (e.g., walking and cycling) at the city-level (macro-level). Thus, this paper focuses on this relationship by introducing street network morphology factors, such as blocks per area, nodes per blocks and nodes per area. The street network in this study includes interconnecting lines and points that present streets, roads, motorways, intersections and blocks. The percentage of daily trips on foot and by bicycle data that represents walking and cycling are collected from the International Association of Public Transport's (UITP) database. The blocks per area, nodes per area and the nodes per blocks are estimated by modifying and analyzing Open Street Maps (OSM). The data that are used in this study are from 30 cities in different parts of the world. The strength of the relationship in this study was found using the Pearson correlation coefficient. The results show that increase in daily trips on foot and by bicycle is correlated with increasing number of blocks per area and number of nodes per area while daily trips on foot and by bicycle has negative relationship with number of nodes per blocks. Because the urban street network is the result of macroscale planning decisions, considering the relationship between street network morphology and travel behavior can lead to better planning decisions.

Keywords: Network morphology, daily trips on foot and by bicycle, street pattern, sustainable urban transport planning

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# **1.0 INTRODUCTION**

Malaysia has high rate of obesity and diabetes [1]. Based on National Health and Morbidity Survey, more than 20% of Malaysians who have 30 years old and above suffered from diabetes in 2011. Active lifestyle that includes active travel modes such as walking and cycling are associated with a lower likelihood of diabetes and obesity [2]. Urban areas with high population and a variety of activities (e.g., working, studying, and business) need to have proper plans that address the mobility and accessibility needs such as healthier travel modes [3-4].

Urban structure can affect travel behavior and the effects of urban structure on the travel behavior including walking and cycling has been investigated by various studies [e.g. 5-13]. Street networks are important urban structure indicators that affect cities form and travel behavior [14].

Street length and density are two important factors for urban street network that are considered as effective factors in various studies [9, 15]. In addition to the street length and density, intersections are also important indicators for urban street network [16-18]. Ample studies consider a relationship between intersection density and vehicle miles travelled (VMT) [19-22]. Intersection density also can affect street

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connectivity [23-25]. More intersections that provide more connected roads can reduce travel distance and time for pedestrians and cyclists. Connected roads increase walking and cycling and decrease VMT [8, 24, 26-27].

Street patterns also can affect travel behavior since they shaped block size, block length and block density. For instance, some studies find a relationship between block size and number of walking trips [28-30]. Some studies also find a relationship between street block density and VMT per household [31]. Block length also affects miles walked per person [32]. Smaller blocks can represent more roads and intersections that provide more connected and direct roads.

There are some shortcomings that prevent previous efforts to evaluate the relationship between street network morphology and percentage of daily trips on foot and by bicycle at the city-level sufficiently. Firstly, there are limited studies that focus on city-level and compare different cities in different socio-economic contexts. Secondly, intersections and block density that are important parts of the city morphology have not investigated by previous studies as street network morphology indicators and the majority of the previous studies consider these factors for connectivity indicators. Therefore, block density and intersection density are represented by polygons (blocks) per area and nodes per polygons (blocks) in this study to have network morphology indicators in addition to the connectivity factors.

Overall, this study attempts to evaluate the effects of urban street network indicators that shape street network morphology on the walking and cycling at the city-level. These effects can be used for street network design and strategies that encourage walking and cycling and reduce private car usage at the citylevel. The street network morphology in this study is represented by lines and points that include roads, streets, motorways, blocks and intersections.

## 2.0 MATERIALS AND METHOD

Eurostat database, UITP and some developed cities database are limited urban mobility data sources at the city-level. Although UITP database includes more cities and indicators than others, this database does not have enough street network morphology indicators that represent connectivity and block size. Therefore, open street maps (OSM) are used to have nodes per area, nodes per blocks and blocks per area. A node represents an intersection and a block represents an area that is surrounded by roads. Nodes per area and blocks per area represent nodes and blocks density in the study area (defined metropolitan areas by UITP). UITP database provides percentage of daily trips on foot and by bicycle. Table 1 shows the collected data from UITP and OSM. Nodes per area, nodes per blocks and blocks per area are the results of modifying and analyzing OSM data (refer Figure 1). The size of the blocks can be represented by the number of blocks per area. Nodes per blocks also can illustrate the average road density for the blocks. The type of intersection (nodes) is not considered in this macro-level study (e.g., 4-way, 3-way, interchange, etc.). However, more nodes and more routes can represent connection density and connectivity.

Cities	Percentage of daily trips on foot and by bicycle	Nodes per area	Blocks per area	Nodes per blocks
Amsterdam	51	360	54	7
Barcelona	34	269	55	5
Boloana	29	156	18	9
Brussels	28	252	47	5
Budapest	23	253	28	9
Clermont Ferrand	33	259	29	9
Copenhagen	39	142	22	6
Dubai	16	132	19	7
Geneva	34	223	27	8
Glasgow	24	244	35	7
Graz	35	196	20	10
Lille	31	197	25	8
Lisbon	24	264	38	7
London	31	451	53	8
Lyons	33	179	35	5
Madrid	26	323	59	5
Manchester	23	173	17	10
Marseilles	34	197	23	9
Moscow	24	127	17	8

 Table 1
 Research data

	Percentage of daily trips			
Cities	on foot and by bicycle	Nodes per area	Blocks per area	Nodes per blocks
Munich	38	331	52	6
Nantes	23	336	32	10
Newcastle	27	178	19	9
Oslo	26	175	24	7
Paris	36	282	46	6
Rome	24	400	58	7
Rotterdam	42	365	62	6
Seville	42	305	53	6
Singapore	14	116	10	11
Vienna	30	268	46	6
Warsaw	20	179	20	9



Figure 1 Converting OSM data to nodes per area, nodes per blocks and blocks per area

Since variables in this research have interval scales, the Pearson product-moment correlation coefficient (Pearson's correlation) is used to measure the strength and direction of the relationship between percentage of daily trips on food and by bicycle and each street network morphology indicators. A line best fit through the data is used by Pearson's correlation analysis to calculate Pearson correlation coefficient (r) that shows the strength and direction of the relationship. This coefficient is between -1 and +1. Negative values indicate negative relationship and positive values show positive association while zero means no association. The values that are closer to either +1 or -1 represent stronger association while the values that are closer to zero show weak relationship. The statistical significance of a correlation (p-value) tests the null hypothesis that there is no relationship.

A linear association between the variables is assumed by Pearson's correlation. The scatter plots are used to test this assumption, significant outliers and Homoscedasticity. Since variables also should be approximately normally distributed, the variables are tested by Shapiro-Wilk normality test before Pearson's correlation analysis. The null hypothesis for Shapiro-Wilk normality test indicates that the sample is taken from a normal distribution, thus Sig < 0.05 for Shapiro-Wilk Sig rejects the null hypothesis. This test is the most reliable normality test for small to medium sized samples [33-35]. Natural logarithm transformation is used for the variables that are not normally distributed. However, the normality is also tested after the transformation.

## 3.0 RESULTS

Pearson correlation coefficients are estimated to test the relationship between street network morphology indicators and percentage of daily trips on food and by bicycle. Table 2 indicates street network morphology indicators that are correlated with percentage of daily trips on food and by bicycle (p < 0.05). Linear relationship, outliers and homoscedasticity are tested by scatter plots before Pearson correlation coefficients estimation. All variables except blocks per area are normally distributed after removing outliers (Refer Table 3). Table 4 shows the results of Shapiro-Wilk normality test for blocks per area after natural logarithm transformation. This variable is normally distributed after transformation.

The Pearson correlation coefficients are sorted from the highest to the lowest in Table 2. The bivariate correlation between natural logarithm of blocks per area and percentage of daily trips on food and by bicycle is r (30) = 0.530, p < 0.003. This correlation has the highest Pearson correlation coefficient among street network morphology indicators. The bivariate correlation between nodes per blocks and percentage of daily trips on food and by bicycle is r(30) = -0.410, p < 0.024. All of street network morphology indicators except nodes per blocks have positive correlations with daily trips on food and by bicycle. The bivariate correlation between nodes per area and percentage of daily trips on foot and by bicycle is r (30) = 0.404, p < 0.027. This variable has the lowest Pearson correlation coefficient among street network morphology indicators.

Table 2 indicates that cities with higher blocks per area are associated with higher percentage of daily trips on food and by bicycle. More blocks mean more connectivity. Based on previously done studies, it is expected that connectivity produces more direct routes for alternative travel modes and increases trips on food and by bicycle. In addition, activity locations have shorter distances and there are more chances to use non-motorized transport modes when there are more blocks in an area. The Pearson correlation coefficient between percentage of daily trips by foot and bicycle and nodes per blocks shows that more trips by foot and bicycle are associated with fewer nodes per blocks. Nodes per blocks can indicate the morphological characteristic of the blocks and fewer nodes per blocks can be the result of simpler blocks with fewer nodes (intersections). Therefore, more but simpler blocks in an area lead to more trips on food and by bicycle. Table 2 also indicates that cities with higher nodes per area are associated with higher percentage of daily trips on food and by bicycle. Higher nodes per area also can represent connectivity that encourages people to have more walking and cycling trips.

#### Table 2 Pearson correlation coefficients

		Natural logarithm of blocks per area	Nodes per Blocks	Nodes per area
	Pearson Correlation	.530	410	.404
Percentage of daily trips or foot and by bicycle	ີ Sig. (2-tailed)	.003	.024	.027
	Ν	30	30	30

Table 3   Normality test				
		Shapiro-Wilk		
	Statistic	df	Sig.	
Percentage of daily trips on foot and by bicycle	.977	30	.748	
Blocks per area	.913	30	.018	
Nodes per blocks	.938	30	.079	
Nodes per area	.959	30	.287	

Table 4 Normality test after transformation

	Shapiro-Wilk		
	Statistic	df	Sig.
Natural logarithm of blocks per area	.936	30	.073

## 4.0 DISCUSSION AND CONCLUSIONS

The relationship between street network morphology and daily trips on foot and by bicycle is evaluated in this study. The results show that the natural logarithm of block density (blocks per area) and intersection density (nodes per area) have positive relationships with daily walking and cycling trips while more nodes per blocks lead to fewer trips on foot and by bicycle. Among these street network indicators, the natural logarithm of blocks per area has the highest positive coefficient so this variable can be more effective to have more daily trips on foot and by bicycle. The second effective indicator is nodes per area has the least positive coefficient, so it has the least effect on daily trips on foot and by bicycle.

There are some points that make this research significant. Firstly, there are very limited studies that consider city-level and compare various cities in different contexts since the scope of most of the previous efforts includes some cities in a country or some neighborhoods in a city. Therefore, this study tries to investigate the effects of street network morphology on the walking and cycling in various cities in different countries regardless of the context effects. In addition, although street network morphology can affect travel behavior significantly, there are limited studies regarding street network macro-level morphological effects. This study tries to evaluate these effects by considering some street network macro-level morphological indicators such as blocks per area, nodes per blocks and nodes per area.

Further studies can use the same scale (city-level in various countries) for other urban structure indicators and travel behavior variables to find effective factors toward sustainability at the city-level regardless of the context effects. These universal effective indicators can help to have some universal planning strategies. These universal strategies can be modified and localized for various contexts. Future studies also need to update their data sources regularly since some of urban structure indicators change rapidly. Open Street Map data source is used in this study. This data source includes free editable maps. Further studies can enhance the results by including original urban maps.

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