Research Article

Impact of Road Network Topology on Public Transportation Development

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1. Introduction

Amid rapid urbanization, the development of Chinese cities has received much attention from researchers worldwide. Looking back at the development of major cities around the world, we find that urbanization was often accompanied by the continuing centralization of population towards cities and the expanding scale of cities, which has resulted in such “urban malaise” as traffic congestion and environmental degradation [1]. With an increasing number of automobile ownership, China has surpassed the United States and risen to be the world’s largest automobile market. As the supply-demand conflict in urban transportation grows stronger, traffic congestion has become one of the most serious social problems, and this also holds true in big cities in China. As a result, unsustainable development of urban transportation takes a toll on both economic growth and people’s life.

It was reported that in the first quarter of 2021, about 59% Chinese cities were in a state of slow passage and 1.66% in a state of congestion in rush hours [2]. Traffic congestion not only wastes time but increases energy consumption and pollutes the environment. Optimizing public transportation is widely conceived as an effective solution to urban congestion, which can alleviate the energy pressure and environmental pollution caused by wide use of cars and help build a sustainable urban transport system. Under this circumstance, “prioritizing public transportation” has been proposed as the national strategy by the Chinese government. In addition, it was found that giving priority to the development of public transportation can effectively reduce the dependence of citizens on cars and achieve the purpose of alleviating traffic congestion [3].

However, it remains to be explored how to promote the development of urban public transportation, especially to increase the number of per capita public transportation. There are many factors affecting the frequency of per capita public transportation. Based on the travelers’ pursuit of utility maximization of the choice, some scholars analyzed factors such as bus travel time, walking time, and waiting time [4]. Some explored the impact of factors related to
2. Methodology

2.1. Definition of Road Network Layout Structure. Urban road networks often present regular structural characteristics, which reflect the combination of various interrelated factors in the network, including the overall form, arrangement mode, connection mode, and hierarchy of roads. Major structures of road networks include the hierarchical structure, the functional structure, and the layout structure [18]. The layout structure refers to the overall pattern and topology of a road network. The overall pattern describes the structure more intuitively. The main overall patterns include grid, radial, circular and radial, linear, and freestyle patterns, which are usually described with specific pictures or qualitative language; the topology of a road network shows the interrelationship between the basic units that make up the network and the relationship between the units and the whole network [19]. Road network topology is an abstract notion that describes the layout structure of a road network, which considers only the connection rules among the constituent parts and the structural characteristics of the entire network instead of the specific geometric attributes such as width and gradient of lane. Besides, the connection condition of roads such as connectivity, clustering, and centrality are considered. Connectivity describes the number of paths that can be selected between two points and the degree of direct reach. The clustering represents the aggregation degree of vertices in a road network. The centrality can describe the relatively important points and road segments in the road network. For example, the betweenness centrality can reflect the skeleton of the road network and the development model of the city [20]. The road network can be abstracted into a topological structure. The layout structure of the road network can be well quantified by constructing a topological structure model of the urban road network and then performing analysis with the help of the graph theory and complex network theory.

2.2. Relationship between structure of road network and public transportation system. In 1977, Machu Picchu Charter published by the International Institute of Modern Architecture proposed that “private cars should be subordinate to the development of the public transportation system in accordance with the policy of future urban transportation” [21]. In 1996, the New Urbanism Association proposed in New Metropolitan Charter that in contemporary metropolises, the motor vehicle traffic should adapt to the principle of encouraging public transportation and slow passage, which can only be achieved by the way of showing respect for walking and public space [22]. In order to flesh out the concept of new urbanism, many modes have been proposed, and the most representative one is transit-oriented development (TOD) [23]. Relevant studies have focused on the benefits of TOD in transport, environment, and other aspects through case analysis and summarized lessons from real-world cases, which prove the sustainability of the TOD development mode in urban development [24, 25]. After being introduced into China, public transportation planning strategies have greatly promoted the development of the road network structure because of the strategies' emphasis on urban road connection and coordinated development of land use and transport, which is conducive to alleviating the problem caused by expanded urbanization. As a result,
traffic and environmental issues can also be solved through urban development guided by public transportation. At present, studies on the layout structure of road networks mainly focus on structure optimization of road network, which is called the network design problem (NDP), involving two subproblems: continuous network design problem (CNDP) and discrete network design problem (DNPD). The latter is to improve existing sections of roads, whereas the latter is to add new sections. Considering the road selection behavior of the traffic network users and budget constraints, these two types of optimization problems are aimed at minimizing the cost of entire transport system. Based on this, mathematical models and many algorithms describing the network design are proposed, among which are the linear optimization model and the random user equilibrium optimization model [26–28]. From the perspectives of the traffic function of the road network structure model and its adaptability to the traffic model, the road network structure model suitable for public transportation guidance is determined.

2.3. Application of Road Network Structure Agglomeration in the Layout Structure of Road Network

2.3.1. Brief Introduction to Complex Network Theory. Most existing works on structural analysis of road networks employ the graph theory-based network models, in which the complex network model is more commonly used [29]. There are many nodes in the road network of a city, and in this logic, the urban road network can be considered a complex network. To study the structural characteristics of complex networks, a mathematic tool called “graph” is often used to describe such networks. A complex network is a system in which several nodes are combined into a connected whole. The graph theory uses some abstract points to represent the nodes in the specific network. The connection relationship between the nodes in the network is represented by lines connecting the nodes. When the graph theory is applied to the urban road network, each intersection is regarded as a node, and each road as a line in the network. In 1936, the first monograph on the graph theory was published, which led to rapid development of this theory. In the 1960s, Paul Erdos and Alfred Renyi proposed the random graph theory, which started a systematic study of the complex network theory. At the end of the 20th century, as the small world model and the scale-free network were proposed, the complex network theory was no longer limited to mathematics. Empirical research has been performed to analyze topological characteristics of real networks, including networks in fields of physics, biology, information management, and transport [30, 31]. Among them, the research on road networks covers many aspects, such as the geometrical properties, structural characteristics, formation mechanism, and evolution law of the network. There are many basic statistical indicators for road network research, the most important of which include the shortest path, average distance, aggregation coefficient, and betweenness [32]. It was found that the urban road network shares the many characteristics with the small world network and scale-free network [33, 34].

2.3.2. Construction Method of Road Network Topology. By the abstraction way, the construction methods for road network topology are mainly divided into the primal approach and the dual approach [35, 36]. The former abstracts the nodes of the road network (such as road intersections and interchanges) into vertices and the connection lines between nodes into edges, whereas the dual approach abstracts the connection line between nodes into points and abstracts the nodes into edges [37].

2.3.3. Betweenness Centralization and Its Calculation Method. Betweenness can also be called intermediateness, which is an indicator to measure the centrality of nodes in a network. The betweenness of a node refers to the proportion of the number of shortest paths passing through this node to the total number of shortest paths in the network. It is a global variable that reflects the role and influence of the corresponding node or edge on the entire network. A larger value of betweenness indicates more importance of the node in the network. Thus, calculating the betweenness of a node in a road network is of practical significance. Freeman [38] used the index of betweenness centralization of a road network to evaluate the aggregation degree of the road network topology. The network centralization includes closeness centralization, degree centralization, and betweenness centralization [39]. Among them, betweenness centralization ($C_B$) can better describe the degree of road network agglomeration, and hence, it is selected as the index of the road network structure agglomeration.

The betweenness centralization can reflect the centralization level of the overall network. It is the ratio of the sum of the betweenness centrality of the most core point and the difference values of the betweenness centrality of other points in a network to the most possible difference value. It is widely used in studies on social networks, industrial clusters, transport, and other fields [40–42]. The calculation formula is as follows:

$$C_B = \frac{\sum_{i=1}^{N} (C_i - C_{i^*})}{N^2 - 4N^2 + 5N - 2}. \quad (1)$$

In the formula, $i$ refers to the serial number of a node, $N$ refers to the quantity of nodes in the complex network, $C_{i,B}$ refers to the betweenness centrality of the node, and $i^*$ refers to the node with the largest value of the betweenness centrality. Betweenness centrality ($C_{i,B}$) represents the possibility of a node located in the shortest path of any two other nodes in the network, that is, the extent to which a node can be located in the “middle position” of other nodes in the network. The calculation formula of $C_{i,B}$ is as follows:

$$C_{i,B} = \frac{1}{(N - 1)(N - 2)} \sum_{j,k \in N, j \neq k \neq i} \frac{n_{jk}(i)}{n_{jk}}, \quad (2)$$

where $N$ refers to the quantity of nodes in the complex network, $n_{jk}$ refers to the quantity of the shortest paths between node $j$ and node $k$, and $n_{jk}(i)$ refers to the quantity of the
shortest paths between node $j$ and node $k$ passing through node $i$ [42].

2.3.4. Application of Pajek. For analysis and visualization of large networks with thousands and millions of nodes, Pajek is a common but complex analysis tool widely used in analysis of social networks [43, 44], genealogy [45], network analysis [46], aviation networks [47], power grid [48], etc. One of its data objects is "Networks" data with the expanded name of .net. When generating a "Networks" file, we need to enter the name of each node and edge of the network (represented by the serial number of the nodes at both ends of the edge).

The geometric network constructed by ArcGIS can display the serial number of each node. By inputting the nondirectional connection line between the nodes into Pajek, we can calculate the betweenness centralization ($C_B$) and obtain the analysis chart of the network.

3. Results

Based on the betweenness centralization, an index that reflects the road network layout structure and the development of public transportation, that is, the number of per capita public transportation, this paper establishes a model and performs a case study on 29 cities in China at the prefecture level and above. Development degrees of social economy, public transportation, and the topographical conditions may partly decide people’s choice of travel mode. Therefore, in addition to analyzing the index of road network topology, the empirical analysis should also take into account of the status of economic development, the supply of public transportation, the relevant indicators of the road network, the urban topographic conditions, and the built-up area. However, the errors or disturbances caused by factors such as the size of the cities and investments into public transportation should be excluded, and related variables should be controlled. Thus, the regression of the number of public transportation per capita on the layout structure of the urban road network is obtained, and their relevance is revealed.

3.1. Establishment of the Model

3.1.1. Selection of Explanatory Variables. There are many factors affecting residents’ use of public transportation, such as...
as the time and distance of travel (average travel time), land utilization (land utilization mixedness, building density, etc.), road design (density of road network, average width of road, etc.), bus supply (density of bus station, density of bus line network, etc.), and socioeconomic attributes [49]. Since the research in this paper focuses on the large-scale comparison among multiple cities, the betweenness centralization ($CB$) of the urban road network is selected as an indicator to measure the characteristics of urban street patterns, and per capita GDP ($GDP_{pr}$) is used to indicate the economic development of the city. The numbers of public transportation vehicles per 10,000 population ($m_{bo}$) and 300 m bus stations coverage ratio ($c$) both reflect the bus supply of the city. Per capita urban road area ($A_{pr}$) and urban investment in fixed assets ($f_{ai}$) are used to represent the urban built environment. Considering that the slope of the terrain may have an impact on the travel mode in the city, the topography ($t$) of the city is added into the model as an explanatory variable. Specifically, 0, 1, and 2 are used to represent cities with less undulating terrain, cities with moderately undulating terrain, and cities with highly undulating terrain, respectively. Since the scale ($S$) of the city has a significant impact on the distance of travel, thus significantly influencing the structure of urban transportation, the city scale is therefore introduced into the model (indicated by the area of land used for urban construction). Of course, to some extent, the betweenness centralization ($CB$) will also reflect the built environment characteristics of the urban scale such as urban spatial structure or urban morphology. At the same time, to avoid problems that the multicollinearity problem might incur in regression, the authors avoid autocorrelation in the selection of independent variables.

### 3.1.2. Model Construction

In this paper, the frequency of per capita public transportation ($y_p$) is used as the explained variable to construct the multivariate linear regression model. Since the explained variable $y_p$ is relatively larger than the independent variable, in order to assign the independent variable $CB$ (i.e., the betweenness centralization) with a proper meaning, the logarithmic processing of explained

| Table 2: Variable description. |
|--------------------------------|
| Variables | Description |
| $y_p$ | Frequency of per capita transit trip in a year |
| BC | Betweenness centralization |
| $GDP_{pr}$ | Per capita GDP (unit: 10,000 yuan) |
| $m_{bo}$ | Number of public transportation vehicles per 10,000 people |
| $A_{pr}$ | Per capita urban road area (unit: $m^2$) |
| $f_{ai}$ | Urban investment in fixed assets (the generic terms of the workload expressed in monetary terms of building and buying fixed assets in a certain period and related cost) (unit: 100 billion Yuan) |
| $t$ | Topographic index |
| $S$ | City scale (unit: $km^2$) |
| $c$ | Coverage ratio of area within 300 m from bus stations (the area within 300 m from bus stations divided by the urban built-up area) |

| Table 3: Variable data of studied cities. |
|--------------------------------------------|
| City | $y_p$ | $m_{bo}$ | $GDP_{pr}$ | $A_{pr}$ | $f_{ai}$ | $t$ | $S$ | $c$ |
| Beijing | 389 | 18.95 | 15.43 | 7.72 | 6.75 | 0 | 1306 | 0.62 |
| Chengdu | 291 | 18.01 | 11.47 | 13.18 | 3.90 | 2 | 529 | 0.67 |
| Dalian | 348 | 16.72 | 16.31 | 14.48 | 4.31 | 2 | 396 | 0.54 |
| Harbin | 250 | 12.65 | 6.83 | 10.04 | 3.84 | 0 | 391 | 0.69 |
| Haikou | 190 | 9.95 | 5.55 | 8.98 | 0.65 | 0 | 124 | 0.56 |
| Hefei | 281 | 16.01 | 12.77 | 22.72 | 3.12 | 0 | 393 | 0.69 |
| Hohhot | 283 | 29.25 | 15.87 | 17.41 | 0.97 | 2 | 397 | 0.51 |
| Kunming | 308 | 17.76 | 9.77 | 14.26 | 0.41 | 2 | 397 | 0.51 |
| Lanzhou | 307 | 10.91 | 6.04 | 11.74 | 0.74 | 2 | 207 | 0.50 |
| Luoyang | 149 | 8.84 | 6.34 | 11.74 | 0.74 | 2 | 192 | 0.50 |
| Nanchang | 267 | 15.39 | 9.98 | 15.28 | 2.01 | 1 | 250 | 0.46 |
| Nanjing | 166 | 10.80 | 12.46 | 19.84 | 5.09 | 0 | 713 | 0.60 |
| Nanning | 196 | 9.69 | 7.21 | 12.61 | 1.73 | 2 | 283 | 0.60 |
| Ningbo | 210 | 19.75 | 18.90 | 12.61 | 1.93 | 1 | 295 | 0.45 |
| Qingdao | 276 | 16.86 | 14.07 | 21.45 | 2.76 | 1 | 470 | 0.68 |
| Xiamen | 457 | 19.72 | 15.32 | 18.14 | 1.34 | 2 | 282 | 0.61 |
| Shanghai | 199 | 12.25 | 15.64 | 7.28 | 5.64 | 0 | 999 | 0.72 |
| Shenzhen | 848 | 98.53 | 46.77 | 37.03 | 2.50 | 2 | 871 | 0.66 |
| Shenyang | 219 | 10.50 | 11.09 | 14.82 | 5.06 | 0 | 455 | 0.61 |
| Shijiazhuang | 253 | 18.04 | 6.76 | 18.07 | 1.90 | 2 | 217 | 0.64 |
| Suzhou | 191 | 13.50 | 19.88 | 24.05 | 3.07 | 1 | 441 | 0.54 |
| Taiyuan | 193 | 9.91 | 7.79 | 12.53 | 1.48 | 0 | 320 | 0.50 |
| Tianjin | 166 | 11.77 | 15.99 | 15.14 | 8.46 | 0 | 736 | 0.71 |
| Xi’an | 300 | 14.00 | 7.05 | 12.09 | 4.39 | 1 | 424 | 0.68 |
| Xining | 330 | 15.21 | 5.55 | 7.15 | 0.60 | 2 | 85 | 0.52 |
| Yinchuan | 291 | 18.79 | 7.71 | 17.77 | 0.47 | 2 | 149 | 0.48 |
| Changchun | 201 | 12.98 | 9.81 | 18.58 | 2.66 | 0 | 452 | 0.48 |
| Changsha | 247 | 13.89 | 15.07 | 10.01 | 2.69 | 2 | 326 | 0.45 |
| Zhengzhou | 200 | 11.11 | 6.45 | 7.42 | 2.50 | 0 | 383 | 0.68 |
Beijing  
Chengdu  
Dalian  
Hrabin  
Haikou  
Hefei  
Huhhot  
Kunming  
Lanzhou  
Luoyang  
Nanchang  
Nanjing  
Nanning  
Ningbo  
Qingdao  

(a)  

Figure 2: Continued.
variable is performed, namely constructing a semielastic model:

\[
\ln y_p = \beta_0 + \beta_1 C_B + \beta_2 GDP_{pr} + \beta_3 m_{bo} + \beta_4 A_{pr} + \beta_5 f_{ai} + \beta_6 t + \beta_7 S + \beta_8 c + \mu,
\]  

where \( \beta_0 \) refers to the intercept parameter of the semielastic model, \( \beta_1 \)-\( \beta_8 \) are the slope parameters of the model, and \( \mu \) refers to the error term of the model.

3.2. Research Data and Processing

3.2.1. Selection of Case Study Cities and Source of Data. The purpose of this paper is to analyze the impact of the layout structure of the urban road network on public transportation. Therefore, the layout structure of the road network of selected research objects should be representative, and the development of urban public transportation should be more balanced. In the present work, a total of 29 municipalities, including Beijing, Shanghai, Shenzhen and Lanzhou, some provincial and subprovincial cities, and prefecture-level cities, are selected for research. Among them, Lanzhou is a typical linear-shaped city due to the terrain limitation, and the layout structure of its road network also shows a typical linear-shaped distribution. Shenzhen features a linear pattern in terms of urban morphology, and the organization and layout of its road network are therefore a linear one. As for Beijing, the main urban area is steeped in history, and the layout structure of road network features a square-grid structure. Chengdu shows a relatively obvious circular-radial road network, whereas Shanghai has a freestyle road network.
Part of the data of the explained variables and explanatory variables of the model are obtained from *China Urban Statistical Yearbook* (2018) (such as $y_p$, GDP$_{pr}$, $m_{bo}$, $A_{pr}$, $f_{ai}$, and urban scale), and some are obtained based on terrain and landform and data processing of Baidu Map POI (Point of Interest). For example, according to the topography, cities can be divided into three types including cities with less undulating terrain, cities with moderately undulating terrain, and cities with highly undulating terrain. The topographic index is represented by the value of 0, 1, and 2 (Table 1). Based on the POI data of the bus stations in each city, we build a buffer zone of 300 m (Figure 1) with the bus station as the center and use the ratio of the area of buffer zone to the area of the municipal district as the coverage rate of 300 m service area of the bus station. Tables 2 and 3 present the descriptions and specific values of variables of each city studied in the present work.

### Table 4: Betweenness centralization of cities studied.

| City    | $C_B$   | City    | $C_B$   | City    | $C_B$   |
|---------|---------|---------|---------|---------|---------|
| Lanzhou | 0.231591| Qingdao | 0.142179| Yinchuan| 0.164094|
| Shenzhen| 0.228655| Hohhot  | 0.140267| Changchun| 0.160323|
| Xining  | 0.225887| Ningbo  | 0.132651| Haikou  | 0.152036|
| Nanchang| 0.189090| Hefei   | 0.131271| Suzhou  | 0.151685|
| Shijiazhuang| 0.188926| Nanning | 0.127353| Shanghai| 0.146765|
| Xiamen  | 0.188101| Nanjing | 0.108786| Changsha| 0.143727|
| Dalian  | 0.184861| Beijing | 0.107189| Luoyang | 0.092967|
| Taiyuan | 0.180529| Chongdu | 0.097408| Tianjin  | 0.090337|
| Kunming | 0.173238| Xi’an   | 0.094943| Zhengzhou| 0.084134|
| Harbin  | 0.079269| Shenyang| 0.076542|         |         |

![Figure 3: (a) The topological structure of Beijing’s main road network (global view). (b) The topological structure of Beijing’s main road network (partial enlarged view).](image)

3.2.2. Construction of Topology of Road Network. Since the primal approach of construction of topology of road network is simple, intuitive, and consistent with the actual experience of people in real life, it is adopted in this paper. Firstly, the satellite images and street maps of cities studied are extracted by Google Earth. The main and secondary roads within are vectorized by the AutoCAD software, and the results can then be imported into the ArcGIS software. The original road map of the road network of cities studied (Figure 2) can be obtained by utilizing utility network analysis to build the geometric networks of the road networks.

3.2.3. Calculation of Betweenness Centralization. The geometric network constructed by ArcGIS displays the serial number of each node. By inputting the nondirectional connection line between the nodes into Pajek, and calculating the betweenness centralization, we can obtain the $C_B$ of the
road network of each city studied (Table 4). The analysis charts of Pajek are shown in Figure 3. The red dots in the figure represents road nodes, and the size of the dot indicates their betweenness centrality ($C_B$), which is also specified by the values in square brackets; values outside the square brackets represent the name of the node.

Figure 3 and Table 4 show that the $C_B$ of linear cities and radial cities (such as Lanzhou, Shenzhen, and Xining) are higher than those of circular and grid cities (such as Beijing and Shenyang).

3.3. Regression Analysis

3.3.1. Summary Statistics. Summary statistics for regression analysis are shown in Table 5.

3.3.2. Result of Regression Analysis. The OLS (ordinary least square) estimation method is employed to estimate the above model, and the estimated values of each coefficient and related statistics are obtained, as shown in Table 6.

The formula is

$$\ln y_p = 4.59 + 2.81C_B - 0.02GDP_{pr} + 0.02m_{bo} - 0.005A_{pr} + 0.04f_{ai} + 0.08t + 0.0001S + 0.43c,$$

$$N = 29, R^2 = 0.59.$$

The $F$ test of the model is significant, and the imitative effect is good, which is of statistical significance. As shown in the result of regression analysis, there is a significantly positive correlation between the betweenness centralization ($C_B$) and per capita public transportation ($y_p$) with a confidence of 95%, that is, the frequency of per capita public transportation increases with the increase of $C_B$ of an urban road network. When the significance level is 5%, the probability $P$ of the $C_B$ is 0.0432, and the coefficient is 2.81, which indicates that under this model, for every 0.1 increase in $C_B$, the $y_p$ will increase by 28.1%.

According to the analysis result, when the variables such as GDP$_{pr}$, $m_{bo}$, $c$, $A_{pr}$, $f_{ai}$, and $t$ remain unchanged, cities with a linear or radial road network have a higher frequency of per capita public transportation than cities with a grid road network or a circular road network. In other words, radial road networks may be more conducive to the development of public transportation than the grid one.

4. Discussion and Conclusion

With 29 cities in China as the study cases, a multivariate linear regression model is established to estimate the impact of the layout structure of urban road networks on residents’ utilization of public transportation. With other factors controlled, the results show that there is a significant correlation between the layout structure of an urban road network and residents’ utilization of public transportation. A greater betweenness centralization means a more centralized road network and more per capita transit trips for urban residents, which contributes to a higher degree of utilization of public transportation by the residents.

There are many factors affecting public transportation, including the development of urban economy and public transportation, vehicle ownership, city scale, and urban topography. In addition, the betweenness centralization is also an important factor affecting the frequency of per capita transit trip. The reasons are as follows. On the one hand, a linear and radial road network facilitates traveling between the central area of the city and the surrounding suburbs, as well as the traffic among the city groups, which is beneficial to the organization of bus routes. Based on several main nodes, building a layout model of a road network makes it more convenient to plan transfers in public transportation, which can reduce the operating cost of public transportation and improve the bus service. In addition, the strip-shaped cities are mostly large cities with and undulating
fluctuations, complex terrain, and large slopes, which are not suitable for bicycle travel. Therefore, the use of public transportation is promoted. As for a grid road network, there is a high repetition rate of public bus routes. Therefore, public transportation is difficult to organize, and passengers have to transfer more times. In addition, excessive intersections are likely to cause congestion and hence result in bad bus-riding experience. On the other hand, compared with a radial road network, a grid road network has better connectivity, accessibility, and flexibility, which promotes the development of private cars. However, as the radial roads show less accessibility for traveling in the tangential direction, private cars may have a higher cost than, or the same cost as, buses, and hence, public transportation is preferred.

Therefore, as for transit-oriented urban planning, it is not wise to blindly improve the density and connectivity of an urban road network. Based on the layout structure of the road network, it is advisable to lay out an axial and radial road network to facilitate the development of public transportation and hence enhance the frequency of residents’ utilization of public transportation.

Data Availability

The data used to support the findings of this study were supplied by the corresponding author under license and so cannot be made freely available. Requests for access to these data should be made to the corresponding author.

Conflicts of Interest

The authors declare no conflict of interest.

Authors’ Contributions

Conceptualization was done by C.M. and X.J.; methodology was done by C.M.; investigation was done by C.M. and W.F.; writing—original draft preparation—was done by C.M.; writing—review and editing—was done by C.M. and Y.M.

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References

[1] S. Anenberg, P. Achakulwisut, M. Brauer, and D. Moran, "Particulate matter-attributable mortality and relationships with carbon dioxide in 250 urban areas worldwide," Scientific Reports, vol. 9, no. 1, article 48057, 2019.
[2] "Q1 traffic analysis reports for major cities in China," https://report.amap.com/index.do.
[3] "Outline of the 12th five-year plan for economic and social development," http://www.gov.cn/2011lh/content_1825838.htm.
[4] Y. Xue, H. Guan, H. Qin, T. Liu, and L. Gong, “Study on key factors influencing public transport share rate based on disaggregate model — Jinan as an example,” in 14th COTA International Conference of Transportation Professionals, Changsha, China, 4 July 2014.
[5] Y. Pei and D. Xu, “Study on the application of mass transit ridership prediction base on fuzzy inference,” China Civil Engineering Journal, vol. 7, pp. 22–26, 2003.
[6] M. Zhang, “Travel choice with no alternative,” Journal of Planning Education and Research, vol. 25, no. 3, pp. 311–326, 2006.
[7] H. Sung and J. T. Oh, “Transit-oriented development in a high-density city: identifying its association with transit ridership in Seoul, Korea,” Cities (London, England), vol. 28, no. 1, pp. 70–82, 2011.
[8] C. Wong, W. Szeto, L. Yang, Y. Li, and S. Wong, ‘Elderly users’ level of satisfaction with public transport services in a high-density and transit-oriented city,” Journal of Transport & Health, vol. 7, pp. 209–217, 2017.
[9] D. Rodriguez, E. Brisson, and N. Estupinan, "Relationship between segment-level built environment attributes and pedestrian activity around Bogota’s bus rapid transit stations," in Transportation Research Board 88th Annual Meeting, Washington DC, United States, 11 January 2009.
[10] H. Badland, G. Schofield, and N. Garrett, “Travel behavior and objectively measured urban design variables: associations for adults traveling to work,” Health & Place, vol. 14, no. 1, pp. 85–95, 2008.
[11] J. Cai and X. Lu, “Impact of road network density on promoting bus traffic development,” Urban Transport of China, vol. 2, pp. 1–9, 2016.
[12] J. Wu, Studies on the Complexity of Topology Structure in the Urban Traffic Network [Phd Thesis], Beijing Jiaotong University, Beijing, China, 2008.
[13] V. Latora and M. Marchiori, “Is the Boston subway a small-world network?,” Physica a: statistical mechanics and its applications, vol. 314, no. 1-4, pp. 109–113, 2002.
[14] J. Sienkiewicz and J. A. Holyst, “Statistical analysis of 22 public transport networks in Poland,” Physical Review E Statistical Nonlinear & Soft Matter Physics, vol. 72, no. 4, article 046127, 2005.
[15] P. Sen, S. Dasgupta, A. Chatterjee, P. A. Sreeram, G. Mukherjee, and S. S. Manna, "Small-world properties of the Indian railway network," Physical Review E, vol. 67, no. 3, article 036106, 2003.
[16] Z. Tian, Z. Zhang, H. Wang, and L. Ma, “Complexity analysis on public transport networks of 97 large- and medium-sized cities in China,” International Journal of Modern Physics B, vol. 32, no. 9, article 1850108, 2018.
[17] P. Ye, X. Chen, and X. Cui, “Impact on density of public transportation network by urban road network layout,” Journal of Tongji University: Natural Science, vol. 1, pp. 51–56, 2012.
[18] F. Shi and W. Wang, “City planning review, urban road network structure analysis,” City Planning Review, vol. 8, pp. 68–73, 2007.
[19] P. Ye, Network Characteristics of Urban Road Network [PhD Thesis], Tongji University, Shanghai, China, 2008.
[20] S. Porta, V. Latora, F. H. Wang et al., “Street centrality and the location of economic activities in Barcelona,” Urban Studies, vol. 49, no. 7, pp. 1471–1488, 2012.
[21] International Union of Architects, Charter of Machu Picchu, The Fifth Column, Lima, Peru, 1977.
[22] Congress for the New Urbanism, *Charter of New Urbanism*, McGraw-Hill Professional, Charleston, America, 1996.

[23] P. Calathorpe, *Transit Oriented Development in China: A Manual of Land-use and Transportation for Low Carbon Cities*, China Architecture & Building Press, Beijing, China, 1st ed. edition, 2014.

[24] M. Kamruzzaman, D. Baker, S. Washington, and G. Turrell, “Advance transit oriented development typology: case study in Brisbane, Australia,” *Journal of Transport Geography*, vol. 34, pp. 54–70, 2014.

[25] A. Nasri and L. Zhang, “The analysis of transit-oriented development (TOD) in Washington, D.C. and Baltimore metropolitan areas,” *Transport Policy*, vol. 32, pp. 172–179, 2014.

[26] K. Sohn, “Multi-objective optimization of a road diet network design,” *Transportation Research Part A: Policy and Practice*, vol. 45, no. 6, pp. 499–511, 2011.

[27] G. Wang, Z. Gao, M. Xu, and H. Sun, "Joint link-based credit charging and road capacity improvement in continuous network design problem," *Transportation Research Part A: Policy and Practice*, vol. 67, pp. 1–14, 2014.

[28] H. Liu and D. Wang, "Global optimization method for network design problem with stochastic user equilibrium," *Transportation Research Part B Methodological*, vol. 72, pp. 20–39, 2015.

[29] M. Rosvall, A. Trusina, P. Minnhagen, and K. Sneppen, "Networks and cities: an information perspective," *Physical Review Letters*, vol. 94, no. 2, article 028701, 2005.

[30] D. J. Watts and S. H. Strogatz, "Collective dynamics of 'small-world' networks," *Nature*, vol. 393, no. 6684, pp. 440–442, 1998.

[31] A. L. Barabási and R. Albert, "Emergence of scaling in random networks," *Science*, vol. 286, no. 5439, pp. 509–512, 1999.

[32] S. Boccaletti, V. Latora, Y. Moreno, M. Chavez, and D. Hwang, "Complex networks: structure and dynamics," *Physics Reports*, vol. 424, no. 4-5, pp. 175–308, 2006.

[33] F. Shi and Y. Ju, “Analysis on influencing factors of public transportation share: an empirical study of central Nanjing,” *City Planning Review*, vol. 2, pp. 76–84, 2015.

[34] P. Ye, "Complex network characteristics of urban road network topology," *Journal of Transportation Engineering and Information*, vol. 1, pp. 13–19, 2012.

[35] L. Freeman, "Centrality in social networks conceptual clarification," *Social Networks*, vol. 1, no. 3, pp. 215–239, 1978.

[36] Y. He, H. Zhao, and D. Yao, "An integrated approximation algorithm for the betweenness centrality and average shortest-path length," *Complex Systems and Complexity Science*, vol. 3, pp. 44–53, 2011.

[37] H. Jeong, H. Kim, and K. Kim, "Betweenness centralization analysis formalisms on workflow-supported org-social networks," in *International Conference on Advanced Communication Technology*, Pyeongchang, Korea (South), 16 February 2014.