

PEDRO AUGUSTO BORGES DOS SANTOS

TITLE: SUBTITLE

Thesis presented as a partial requirement for the obtaining of a degree in Master of Sciences by the Programa de Pós-Graduação em Planejamento Urbano of Setor de Tecnologia of Universidade Federal do Paraná.

Supervisor: Prof. Dr. Jorge Tiago Bastos Co-supervisor: Prof. Dr. Oscar Oviedo-Trespalacios

CURITIBA

ABSTRACT

Resumo

Palavras-chaves: Palavras-chave

ABSTRACT

Abstract

Key-words: Key-words

LIST OF FIGURES

FIGURE 1 - ROAD TRAFFIC RATES IN OECD COUNTRIES	9
FIGURE 2 - ROAD TRAFFIC DEATHS ON BRAZIL	11
FIGURE 3 - ROAD TRAFFIC MORTALITY RATE ON BRAZIL	11
FIGURE 4 - ROAD TRAFFIC FATALITY RATE ON BRAZIL	12
FIGURE 5 - ROAD TRAFFIC DEATHS ON CURITIBA	13
FIGURE 6 - MOTORIZATION RATES ON BRAZILIAN STATES CAPITAL CITIES,	
IN 2019	13
FIGURE 7 - MORTALITY RATES ON BRAZILIAN STATES CAPITAL CITIES,	
IN 2019	14
FIGURE 8 - FATALITY RATES ON BRAZILIAN STATES CAPITAL CITIES, IN	
2019	15
FIGURE 9 - KERNEL FUNCTIONS PLOT	17

LIST OF TABLES

SUMMARY

1	INTRODUCTION	7
1.1	OBJECTIVES	7
1.2	JUSTIFICATION	
1.3	THESIS STRUCTURE	8
2	LITERATURE REVIEW	9
2.1	ROAD SAFETY SCENARIO	Ĝ
2.2	SPEEDING	14
2.3	ROAD SAFETY PERFORMANCE AND THE BUILT ENVIRONMENT	14
2.4	NATURALISTIC DRIVING STUDY	15
2.5	GEOGRAPHICALLY WEIGHTED REGRESSION	15
3	METHODS	19
3.1	NATURALISTIC DATA COLLECTION	19
3.2	DATA PROCESSING	19
3.3	GEOGRAPHICALLY WEIGHTED MODEL	19
4	ANALYSIS AND RESULTS	20
5	CONCLUSIONS	21
	BIBLIOGRAPHY	22

1 INTRODUCTION

Very general introduction why the topic of the thesis is relevant.

Road Safety Scenario in the World and in Brazil.

Urban planning having a crucial role on the road safety in cities.

1.1 OBJECTIVES

In the context of the urban planning practices and road safety management, the main objective of this research is to investigate the influence of demographic, socioeconomic, land use and constructed environment on the speeding practices as a risk factor on the frequency and severity of road crashes. The scenario of the study is the city of Curitiba, capital of the state of Paraná, Brazil. The correlation (or lack of) will be analyzed with the use of the Geographically Weighted Regression statistical model.

With the use of speeding data collected from drivers that participated in a Naturalistic Driving Study (NDS) performed in Curitiba, this investigation aims to contribute to the knowledge of its relationship with variables from the built environment (BE). These BE variables are categorized by Ewing and Dumbaugh (2009) in five groups called '5D': density, diversity, design, destination accessibility and distance to transit. In addition to these variables, this study aims to relate the income as a socioeconomic variable to the speeding as well.

As co-benefits in the investigation, this thesis aims to offer some level of insight regarding the development and update of speed control in cities, considering the operational and structural planning guidelines for the road systems and land use. With these factors in mind, it could be possible to present new ideas to the planning practices in Curitiba, regarding mobility plans, master plans and zoning laws.

1.2 JUSTIFICATION

The excess of traffic speed performed by vehicles on urban environments are a main risk factor in the chance and severity of road crashes, considering there is great volume of interaction between motorized and vulnerable users (ELVIK et al., 2009). Considering the direct relationship between the mobility, the land use patterns and the built environment (DE VOS; WITLOX, 2013), it is important to investigate these factors in search of improvements in the management process of the road safety. Speeding, which is on the primary causes for traffic crashes (WHO, 2013) can be better studied and analyzed by the collection of naturalistic data.

Having in mind that most of the traffic crashes and conflicts happens in cities (WHO, 2018), as a consequence of the fast process of urbanization in the last decades, it is crucial to create safer conditions in these localities. In Brazil, the management of the mobility is attributed to the municipalities (BRASIL, 1997). Therefore, the management of road safety, as a task inherent to traffic management, is also an attribution of the municipalities. It is necessary to have in mind the issues of road safety in cities when conducting the process of urban planning. Having this in mind, it was established laws in which this integration between the planning process by the municipalities and the sustainable mobility are clearly defined: The Statute of the Cities (BRASIL, 2001) and The Law of the Urban Mobility (BRASIL, 2012).

The built environment can affect the traffic safety through three main mediators: the traffic volumes, traffic conflicts and traffic speeds - which can directly affect the crash safety and severity in urban environments (EWING; DUMBAUGH, 2009). Traffic crashes are a final outcome of the problems involving the road safety. Therefore, in order to have a method to analyzed the operational conditions of the traffic safety in cities, it can be beneficial to use a intermediate outcome - or a safety performance indicator (SPI) (BASTOS, 2014)- which can be the amount of speeding that occurs across its territory, in addition to the studies that uses the traffic crashes as dependent variables.

1.3 THESIS STRUCTURE

General composition of the thesis, giving which chapters the reader should expect

Write this part after completion of chapters 2 and 3

2 LITERATURE REVIEW

2.1 ROAD SAFETY SCENARIO

Road traffic crashes around the world claims more than 1.3 million lives in each year, representing the eighth leading cause of deaths and causing up to 50 million injuries. Low and middle income countries (LMICs), including Brazil, suffers from traffic crashes death rates three times higher when compared to developed countries (WHO, 2018). The increasing number of traffic crashes deaths on LMICs is a consequence of a intense process of motorization that has been occurring in the last decades. To Bhalla and Mohan (2016), it is important to understand the evolution of road safety management on OECD developed countries in order to overcome the road safety problems in LMICs.

Over the last century, the road safety performance of the OECD countries showed a consistent pattern. The road traffic death rate (per 100,000 population) in these countries were rising until the 1960s, as seen in FIGURE 1. After this period, all countries showed a declining pattern, whilst the LMICs still had a rising pattern due to the rapid growth in their motor vehicle fleets. This behavior of rising and declining trend on the road traffic death rate could be explained by three phenomenons: economic determinism, risk substitution and a political shift in the road safety paradigm (BHALLA; MOHAN, 2016).

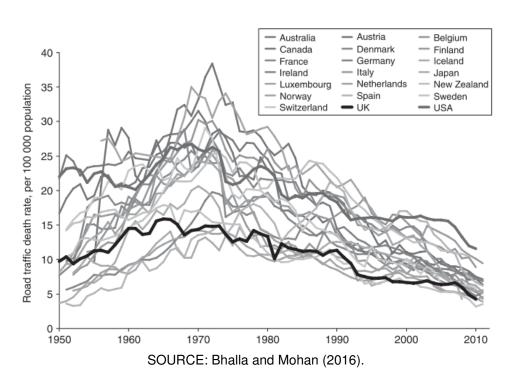


FIGURE 1 - ROAD TRAFFIC RATES IN OECD COUNTRIES

In the scope of economic determinism, the road traffic deaths are defined as a process related to the country development. The rising pattern is associated with the increase in motorization, and the falling happens after a certain level of development, where the countries have the means to start investing in road safety. But this hypothesis has some flaws. This creates an impression that LMICs are not able to invest in road safety before becoming full developed, which is inaccurate. Also, this idea shifts the focus on investment in direct interventions, encouraging the countries to focus on income growth as a strategy for road safety (BHALLA; MOHAN, 2016).

As the mobility in general becomes more motorized towards the use of the car, pedestrians starts transitioning into car users. This circumstance is known as risk substitution, where the increase in car users occurs at the same time as the number of pedestrians declines, lowering the exposure to more severe road traffic injuries like the collision between cars and pedestrians and lowering the number of road traffic deaths. But this phenomenon can be imprecise when considering the motorization of LMICs, in which mass transit, motorcycles and other non motorized transports have a greater role on the vehicle fleet (BHALLA; MOHAN, 2016).

A factor that explains this pattern in a more precise way is the political shift in the road safety paradigm that happened on the OECD countries. Before the 1950s, the belief that drivers were the only responsible for the road crashes lead the discussions regarding road safety. Therefore, most of the interventions in road safety management ignored the design and development of the build environment and the vehicles. Between the 1960s and the 1970s, OECD countries started to regulate transport in order to tackle the road safety problems that were rising, establishing new laws and road safety management institution on national e local levels. (BHALLA; MOHAN, 2016)

According to WHO (2018), it was predicted that Brazil would had a road traffic mortality rate (number of deaths per 100,000 inhabitants) of 22.5 in 2016, the highest rate between the south american countries. In 2019 (the last entry available by Ministry of Health (2020)), road crashes were responsible for 31.945 deaths. Considering the Decade of Action for Road Safety 2011-2020 (WHO, 2011), Brazil will reach the goal of reducing the road traffic deaths in half (comparing to the number projected for 2020, in case of a rising trend in deaths) by the end of the 2010s. FIGURE 2 has the time series of road traffic deaths in the last decades.

The year of 1979 is the earliest official data entry available. Starting in the 1980s, there is a overall rising trend in the numbers of road traffic deaths that reach its peak in 2012, then it starts declining. In 2010, the year before the Decade of Action, Brazil had 42,844 deaths and reached the max value in 2012: 44,812 road traffic deaths. Almost at the end of the decade the number of deaths declined, reaching 31,945. The next plot (FIGURE 3) shows the mortality rate in Brazil, considering the same time span.

Road traffic deaths Year

FIGURE 2 - ROAD TRAFFIC DEATHS ON BRAZIL

SOURCE: Ministry of Health (2020).

In general, it follows the same pattern of the absolute road traffic deaths.

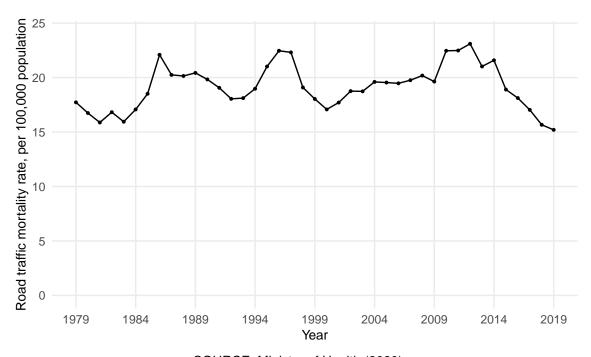


FIGURE 3 - ROAD TRAFFIC MORTALITY RATE ON BRAZIL

SOURCE: Ministry of Health (2020).

The maximum registered mortality rate is 23.1 deaths per 100,000 inhabitants in 2012, the same year which Brazil reached its maximum value for absolute road traffic deaths. Since then the mortality rate has been declining, reaching a rate of 15.2 deaths per 100.000 inhabitants in 2019. In comparison to the OECD countries (FIGURE 1),

Brazil presented a strong declining pattern a few years later, after the 2010s. Another indicator that considers the exposition to traffic hazards is the fatality rate - number of deaths per 10,000 vehicles (FIGURE 4). The oldest entry of fleet size in the DENATRAN (2020) database is 1998.

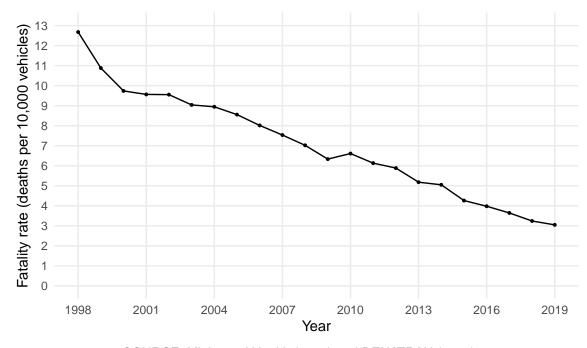


FIGURE 4 – ROAD TRAFFIC FATALITY RATE ON BRAZIL

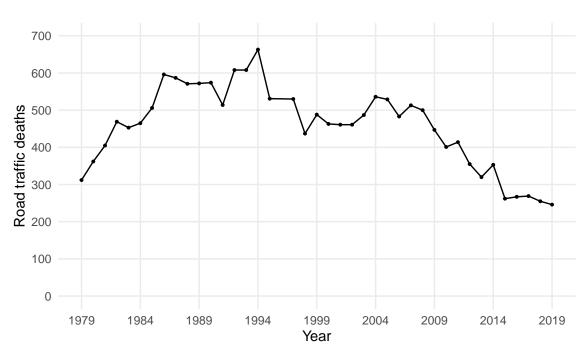
SOURCE: Ministry of Health (2020) and DENATRAN (2020).

Between 1998 and 2019 there is a declining pattern in the fatality rate, starting with 12.7 deaths per 10,000 vehicles in 1998 and ending with 3.0 in 2019, representing a reduction of 76%. This declining pattern can be explained by the continuous rise of the motorization rate in Brazil (ref). The variation in the number of road traffic deaths and road traffic mortality is partially related to the economic landscape in Brazil.

In Curitiba, the road traffic deaths (FIGURE 5) presented a rising pattern between 1979 and 1994, starting with xxx deaths and reaching a maximum of xxx deaths. Overall, the pattern of the time series started declining after 1995, ending with a minimum value of xxx in 2019. Following the brazilian trend, Curitiba's motorization rate kept rising in the last decades. The increase in motorized transit can lead to more exposition to traffic crashes and injuries, in case road safety interventions are not implanted correctly. The plot in FIGURE 6 compares the motorization rate between all brazilian states capital cities.

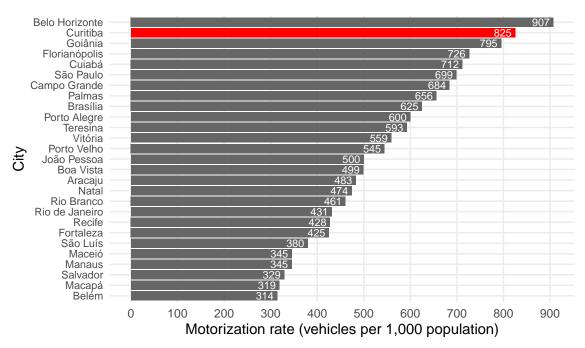
Curitiba was, in 2019, the second most motorized capital in Brazil, with a motorization rate of 825 vehicles per 1,000 population (compare to Brazil), losing only to Belo Horizonte, which presented a motorization rate of 907 vehicles per 1,000 population. When comparing the road traffic mortality rate FIGURE 7, Curitiba stays in

FIGURE 5 - ROAD TRAFFIC DEATHS ON CURITIBA



SOURCE: Ministry of Health (2020).

FIGURE 6 - MOTORIZATION RATES ON BRAZILIAN STATES CAPITAL CITIES, IN 2019



SOURCE: Ministry of Health (2020) and DENATRAN (2020).

the 13th place with 12.7 deaths per 100,000 inhabitants, below the brazilian average of 15.2 in 2019. The lowest value of road traffic mortality rate belongs to Rio de Janeiro, presenting a rate of 5.7 deaths per 100,000 population, and the highest belongs to Palmas, with a alarming rate of 36.4.

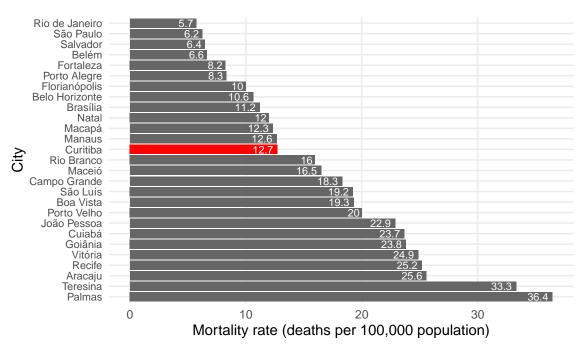


FIGURE 7 - MORTALITY RATES ON BRAZILIAN STATES CAPITAL CITIES, IN 2019

SOURCE: Ministry of Health (2020).

Considering the fatality rate as a road safety performance indicator, FIGURE 8 plot shows that Curitiba have the 6th best performance, with a fatality rate of 1.5 deaths per 10,000 vehicles (compare to brazil). São Paulo is the capital with the lowest fatality rate (0.9) and Recife has the highest one (5.9).

2.2 SPEEDING

2.3 ROAD SAFETY PERFORMANCE AND THE BUILT ENVIRONMENT

Present all the available and used indicators. Guide the literature review based on the indicators (which articles are using the same indicators and why - ewing ...)

add ewing, dumbaugh scheme of BE, mediators and safety.

Add description from urban elements (reports passed form Tiago)

Master plan, mobility planning and zoning laws (try to focus on the mobility and road safety, maybe a brief history). -> relating to the previously described elements

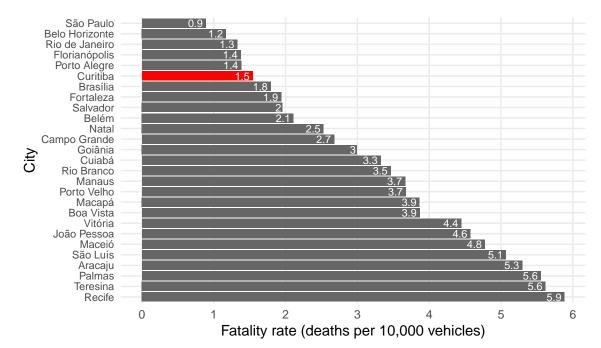


FIGURE 8 - FATALITY RATES ON BRAZILIAN STATES CAPITAL CITIES, IN 2019

SOURCE: Ministry of Health (2020) and DENATRAN (2020).

2.4 NATURALISTIC DRIVING STUDY

- Introduce the Naturalistic Driving Studies as an alternative for the analysis of behavior. Describe multiple international NDS programs, reference the NDS that analyses the speeding. (this part needs to be well described.)
- categorize the nds programs and compared with the method used in this work.
 (maybe construct a table)
- Focus on authors that used NDS to study speeding.

2.5 GEOGRAPHICALLY WEIGHTED REGRESSION

Regression analysis is a tool to investigate the dependency between variables and to predict future parameters based on previous ones. This type of statistical analysis can show the relationship between dependent variables (road accidents or speeding) and independent variables (land use, road design etc.) (LINDLEY, 1987). In investigation of the effects of the constructed environment on the occurrence of crashes in urban areas, negative binomial regression is a traditional model used (WEI; LOVEGROVE, 2013; ZHANG et al., 2014). But the quality of this model is limited by the incapacity of analyzing the spatial dependency and heterogeneity expected to happen on road safety factors related to the urban environment (OBELHEIRO; SILVA; NODARI, 2019).

To overcome this limitation, the Geographically Weighted Regression (GWR) allows the exploration of relationship between variables on a spatial nonstationarity context. The spatial nonstationarity is a scenario where it is assumed that parameters are not constant across the space. In this thesis context, these parameters are the occurrence of speeding across Curitiba's territory. A global regression model may be incapable of explaining the relationship between sets of variables with a acceptable level of precision in this scenario. The GWR model considers that the nature of the model must alter over space to reflect the structure within the data, allowing the actual parameters for each location in space to be modeled and mapped (BRUNSDON; FOTHERINGHAM; CHARLTON, 1996).

The basic form of GW linear regression is based on the following equation:

$$y_i = \beta_{i0} + \sum_{k=1}^{m} \beta_{ik} x_{ik} + \epsilon_i; \tag{2.1}$$

where y_i is the dependent variable (speeding) at location i, x_{ik} is the value of the kth independent variable at location i, m is the number of independent variables, β_{i0} is the intercept parameter at location i, β_{ik} is the local regression coefficient for the kth parameter at location i and ϵ_i is the random error at location i. The GWR model depends on a spatial weighting function called w_{ij} that controls the contribution of the point j on the calibration of a model for point i. The spatial weighting function represents the idea that for each point i there is a bump of influence around it. Considering this "bump", observations closer to i have more influence in the estimation of i's parameters (BRUNSDON; FOTHERINGHAM; CHARLTON, 1996; GOLLINI et al., 2013).

In the context of a multivariate GW model, these influences are calculated by a weighted least squares approach, described on the following equation:

$$\hat{\beta}_i = (X^T W(u_i, v_i) X)^{-1} X^T W(u_i, v_i) y;$$
(2.2)

where X is the matrix of the independent variables with a column of 1s (ones) for the intercept, y is the dependent variable vector, $\hat{\beta} = (\beta_{i0},...,\beta_{im})^T$ is the vector of m+1 local regression coefficients and W_i is the diagonal matrix denoting the spatial weighting (w_{ij}) of each observed data for regression point i at location (u_i,v_i) (defined by the selected kernel function) (GOLLINI et al., 2013). The spatial weighting function is also known as the kernel function. This function can have multiple configurations, including Gaussian, Exponential, Boxcar, Bisquare and Tricube configurations, detailed on TABLE 1.

The distance between observations i and j is defined by d_{ij} , inside a chosen b bandwidth, which is the key controlling parameter in all kernel functions. The optimal bandwidth can be estimated by a cross-validation function, a process that will be

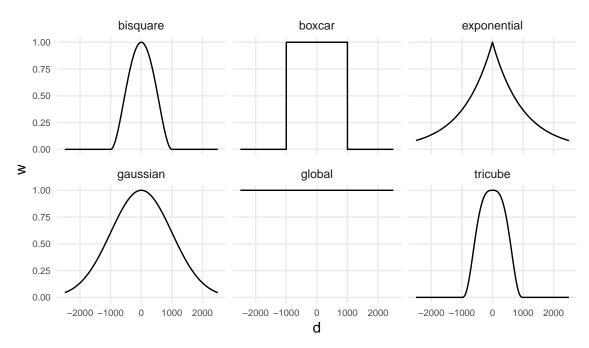
TABLE 1 - POSSIBLE KERNEL FUNCTIONS

Model	Equation
Global Model	$w_{ij} = 1$
Gaussian	$w_{ij} = \exp(-\frac{1}{2}(\frac{d_{ij}}{b})^2)$
Exponential	$w_{ij} = \exp(-\frac{ d_{ij} }{b})$
Boxcar	$w_{ij} = egin{cases} 1 & ext{if } d_{ij} < b, \ 0 & ext{otherwise} \end{cases}$
	0 otherwise
Bisquare	$w_{ij} = \begin{cases} (1 - (d_{ij}/b)^2)^2 & \text{if } d_{ij} < b, \\ 0 & \text{otherwise} \end{cases}$
Disquare	$u_{ij} = 0$ otherwise
Tricube	$w_{ij} = \begin{cases} (1 - (d_{ij} /b)^3)^3 & \text{if } d_{ij} < b, \\ 0 & \text{otherwise} \end{cases}$
iricube	$w_{ij} = \begin{cases} 0 & \text{otherwise} \end{cases}$

SOURCE: Gollini et al. (2013)

detailed in the section 3.3. FIGURE 9 presents the plot of the kernel functions, for a generic bandwidth b=1000, where w is the weight and d is the distance between two observations. As the plot shows, the global kernel function represents a global linear regression, where all the variables are constant across the space.

FIGURE 9 - KERNEL FUNCTIONS PLOT



SOURCE: The Author (2021), based on Gollini et al. (2013).

In addition to the different types of kernel functions, the GWR has multiple configurations based on the type of regression: negative binomial (GWNBR), Poisson (GWPR) and Gaussian, which is the basic form of linear GWR described on EQUA-

TION 2.1. The GWNBR and GWPR models were used by Obelheiro, Silva, and Nodari (2019); Obelheiro, Silva, Nodari, et al. (2020) and Yu and Xu (2017) to study the impacts of the built environment on the occurrence of road crashes inside urban areas.

In conclusion, GWR is a method that enables the exploration of findings that might otherwise be missed if only a global regression method is applied. It is always important to really check if the GWR model will describe the data better than a global regression model comparing the results from both. All the process of constructing a GWR model and its parameters for this work is described in section 3.3.

3 METHODS

- 3.1 NATURALISTIC DATA COLLECTION
- 3.2 DATA PROCESSING
- 3.3 GEOGRAPHICALLY WEIGHTED MODEL

4 ANALYSIS AND RESULTS

5 CONCLUSIONS

BIBLIOGRAPHY

BASTOS, Jorge Tiago. **Análise Estratégica da Segurança viária no Brasil: Pesquisa de índices e Indicadores**. 2014. s. 290. PhD thesis – Universidade de São Paulo / Hasselt University. Cit. on p. 8.

BHALLA, Kavi; MOHAN, Dinesh. Understanding the Road Safety Performance of OECD Countries. In: TRANSPORT Planning and Traffic Safety. Boca Raton: CRC Press, 2016. p. 1–15. DOI: 10.1201/b19737-2. Cit. on pp. 9, 10.

BRASIL. **Lei nº 10.257, de 10 de julho de 2001 - Estatuto das Cidades**. Brasília: Presidência da República, 2001. Cit. on p. 8.

_____. **Lei nº 12.587, de 3 de janeiro de 2012 - Lei da Mobilidade**. Brasília: Presidência da República, 2012. Cit. on p. 8.

Lei nº 9.503, de 23 de setembro de 1997. **Código de trânsito Brasileiro**, 1997. Cit. on p. 8.

BRUNSDON, Chris; FOTHERINGHAM, A. Stewart; CHARLTON, Martin E. Geographically Weighted Regression: A Method for Exploring Spatial Nonstationarity. **Geographical Analysis**, v. 28, n. 4, p. 281–298, Sept. 1996. ISSN 00167363. DOI: 10.1111/j.1538-4632.1996.tb00936.x. Available from: http://doi.wiley.com/10.1111/j.1538-4632.1996.tb00936.x. Cit. on p. 16.

DE VOS, Jonas; WITLOX, Frank. Transportation policy as spatial planning tool; Reducing urban sprawl by increasing travel costs and clustering infrastructure and public transportation. **Journal of Transport Geography**, Elsevier Ltd, v. 33, p. 117–125, 2013. ISSN 09666923. DOI: 10.1016/j.jtrangeo.2013.09.014. Available from: http://dx.doi.org/10.1016/j.jtrangeo.2013.09.014. Cit. on p. 7.

DENATRAN. **Estatísticas - Frota de veículos**. [S.I.: s.n.], 2020. Available from: https://www.infraestrutura.gov.br/component/content/article/115-portal-denatran/8552-estat%7B%5C'%7Bi%7D%7Dsticas-frota-de-ve%7B%5C'%7Bi%7D%7Dculos-denatran.html. Cit. on pp. 12, 13, 15.

ELVIK, Rune et al. **The Handbook of Road Safety Measures**. Second edi. Bingley: Emerald Group Publishing Limited, 2009. ISBN 9781848552500. Cit. on p. 7.

EWING, Reid; DUMBAUGH, Eric. The built environment and traffic safety: A review of empirical evidence. **Journal of Planning Literature**, v. 23, n. 4, p. 347–367, 2009. ISSN 08854122. DOI: 10.1177/0885412209335553. Cit. on pp. 7, 8.

GOLLINI, Isabella et al. GWmodel: an R Package for Exploring Spatial Heterogeneity using Geographically Weighted Models, June 2013. arXiv: 1306.0413. Available from: http://arxiv.org/abs/1306.0413. Cit. on pp. 16, 17.

LINDLEY, D. V. Regression and Correlation Analysis. In: THE New Palgrave Dictionary of Economics. London: Palgrave Macmillan UK, 1987. p. 1–6. DOI: 10.1057/978-1-349-95121-5_1873-1. Available from: http://link.springer.com/10.1057/978-1-349-95121-5 1873-1. Cit. on p. 15.

MINISTRY OF HEALTH. **Number of traffic fatalities in Brazil**. [S.I.]: Ministry of Health, 2020. Available from: http://www2.datasus.gov.br/DATASUS/index.php?area=02. Cit. on pp. 10–15.

OBELHEIRO, Marta Rodrigues; SILVA, Alan Ricardo da; NODARI, Christine Tessele, et al. A new zone system to analyze the spatial relationships between the built environment and traffic safety. **Journal of Transport Geography**, v. 84, p. 102699, Apr. 2020. ISSN 09666923. DOI: 10.1016/j.jtrangeo.2020.102699. Available from: https://linkinghub.elsevier.com/retrieve/pii/S096669231930225X. Cit. on p. 18.

OBELHEIRO, Marta Rodrigues; SILVA, Alan Ricardo; NODARI, Christine Tessele. UMA ANÁLISE DA RELAÇÃO ENTRE AMBIENTE CONSTRUÍDO E ACIDENTES DE TRÂNSITO EM ZONAS DE TRÁFEGO. **33º Congresso de Pesquisa e Ensino em Transporte da ANPET**, p. 3696–3707, 2019. Cit. on pp. 15, 18.

WEI, Feng; LOVEGROVE, Gordon. An empirical tool to evaluate the safety of cyclists: Community based, macro-level collision prediction models using negative binomial regression. **Accident Analysis & Prevention**, v. 61, p. 129–137, Dec. 2013. ISSN 00014575. DOI: 10.1016/j.aap.2012.05.018. Available from: https://linkinghub.elsevier.com/retrieve/pii/S0001457512002023. Cit. on p. 15.

WHO. Global Plan for the Decade of Action for Road Safety 2011-2020. [S.l.], 2011. Cit. on p. 10.

http://journals.sagepub.com/doi/10.1177/0739456X17696035. Cit. on p. 18.

011. Cit. on p. 10.	
. Global status report on road safety 2013 - supporting a decade of	
ction. [S.I.], 2013. p. 1–318. Available from:	
ttp://www.who.int/about/licensing/copyright_form/en/index.html. Cit. on p. 7.	
. Global status report on road safety 2018. Geneva, 2018. Cit. on pp. 8–1	ე.
U, Chia-Yuan; XU, Minjie. Local Variations in the Impacts of Built Environments on	
raffic Safety. Journal of Planning Education and Research, p. 0739456x1769603	,
lar. 2017. ISSN 0739-456X. DOI: 10.1177/0739456X17696035. Available from:	

ZHANG, Cuiping et al. Crash Prediction and Risk Evaluation Based on Traffic Analysis Zones. **Mathematical Problems in Engineering**, v. 2014, p. 1–9, 2014. ISSN 1024-123X. DOI: 10.1155/2014/987978. Available from: http://www.hindawi.com/journals/mpe/2014/987978/. Cit. on p. 15.