

Towards a city congestion index: methodological explorations using Google's Distance Matrix API

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Abstract. Sustainable transport systems are a necessary requirement to achieve efficient economic performance, enhance urban quality of life and diminish environmental costs. Congestion, a negative externality of mobility, is responsible for urban pollution, inefficiency and has adverse effects over individuals facing this problem. For these reasons, transport and city planning agencies have developed interests in defining and measuring transportation congestion. Although, different definitions and metrics have been used, congestion measurements are found aggregated at a city level or for particular road segments. This study proposes a methodology that produces information from a web traffic service to map traffic congestion within an urban area. The method is simple and generalizable enough to be adopted in different urban areas. This paper presents the analysis of four european cities (Amsterdam, Glasgow, Goteborg and Lisbon) and show that the conclusions are consistent with the results obtained from internationally recognized organizations such as INRIX and TomTom.

1. Introduction

Although United Nations (UN) has not identified efficient, clean and equitable mobility as one the 18 Sustainable Development Goals (SDGs), transportation indicators are present within several Goals (CITE UN SDGS). Sustainable transport systems are acknowledged to be necessary pre-conditions to enhance socio-economic opportunities, diminish environmental costs and improve road safety conditions (CITE IMPORTANCE OF TRANSPORT).

Within cities over 60% of the total wealth is produced and the positive externalities of living in urban areas has been extensible registered. However, as the population density increases a set of negative externalities such as pollution, crime or congestion erode the benefits of urban life (UN - OECD).

The definition of congestion in transportation facilities has evolved over the years (XXXXXXX) and there is no universal accepted definition for the problem. According to LOMAX 1997, different actors are interested in understanding different parts of traffic congestion; depending on the what the objective of the study is, different definitions can be adopted and as a result different measurments used.

In a survey deployed to reveal the importance of traffic congestion, (Bertini, 2005) shows the different approaches that can be taken to define an measure congestiotn. The results reinforce the idea that multiple definitions co-exist and that they can be used for different purposes

In this paper traffic congestion in cities will be considered a *'by-product of their success in attracting people to jobs and other amenities, and the inability of cities to improve/expand transportation capacity to keep pace with this growth.'* (Falcocchio & Levinson, 2015) continue

stating that *'the cities' challenge is to keep congestion manageable as their population and economies grow. To be helpful in congestion management decisions, the definition of congestion should be based on a comparison of "actual travel times" with "expected travel times" for peak hour and off-peak conditions.'*

1.1. Research Aim

As mentioned above, traffic congestion is an urban problem that deteriorate the positive benefits of urban life, therefore is imperative to understand the problem in depth and look at it from different angles. Traditional congestion measurements are found to be costly and through practice or research different methods have been explored to overcome this issue.

One avenue of research understands congestion as a time and location specific problem, which occurs in a road segment and that its consequences are suffered from those in the immediate surroundings. A second understanding of the problem deals with the individuals who are affected by traffic.

Focusing on this second interpretation of the problem, the aim of this research is to provide a methodology to spatialize congestion to expose how traffic congestion is distributed across an urban area. The map of congestion can provide useful information to local authorities and planners since more disaggregated spatial information will be provided.

2. Theoretical Background

Congestion definitions can be categorized into the following groups

2.1. Traffic congestion measurements

What congestion indices exist out there?

2.2. Characteristics of the metric

2.3. Data collection

Depending on the how is congestion defined. by one or two authors only.

The components of congestion: the route between OD pair, the distance and travel time (consequently, the speed of travel), the capacity of the route. inherently a travel time difference problem. The components of a broader understanding of the phenomenon: a person travelling with a purpose, a date and time of day, two locations (origin and destination). This brings it closer to definitions of accessibility, but measuring the convenience of travel (refs on this?)

What congestion indices exist out there?

How it is measured (absolute and relative indices)

Practice/literature

How are others mapping congestion (or real travel time) with geodata? (edited)

The focus on the link (flow and speed), the focus on visualisation.

As the list of traffic congestion indicators As the list of At the same time new methodologies to measure traffic congestion were developed, researchers

At the same time different

some researcher were

For instance in Lomax (1993)

In parallel, as the list of traffic congestion measurements increased

3. Methodology

At its core the methodology described in this section details a process of collecting and processing data from an internet service that provides traffic estimations and routing optimization. The method presented here correspond to data extracted from the Distance Matrix API from Google Inc., but the exact same exercise could be done with other services such as HERE or TOMTOM.

Given the quantity and nature the methodology that will be described above, was coded and executed using R in Rstudio and the functions written can be found in GitHub.

In order to prioritize clarity over narrative, the necessary steps followed by this methodology are listed below:

- 1) Select and define an urban area to work with: After an urban area is selected a synthetic boundary needs to be drawn. In this case, for simplicity reasons, a circular buffer zone was used to define the 'study area'.
- 2) Choose a zoned cartography of the area and clip it accordingly to the boundaries: Choose from any geographical authority a zoned cartography. Census tracks, neighbourhoods or plots can be chosen. In our case the European Population grids was selected.
- 3) Make centroids from the zones and extract the geographical position of each point
- 4) Generate permutations from centroids to create an empty Origin-Destiny Matrix: Using the GPS location (latitude and longitude) of each centroid, a list of all possible combinations is generated (a total of $n*(n-1)$ routes). For instance, if the area to be studied contains 3 areas (A, B & C), the OD matrix will consider the following routes: (i) A to B, (ii) A to C, (iii) B to A, (iv) B to C, (v) C to A & (vi) C to B.
- 5) Define a *congested* and *non-congested* times: Using the difference between a *congested* and *non-congested* situation as the definition of congestion, a time need to be specified. As information from Google will be used, a future time needs to be specified. Based on previous results, Thursday, October 15, 2020 8:30:00 AM was chosen as a *congested* moment and Wednesday, July 15, 2020 3:30:00 AM, as a *non-congested*.
- 6) Use a routing service to estimate time and distance travelled: The origin to destination list with the *congested* and *non-congested* time specification were used as inputs for the routing service to estimate travel time and distance.
- 7) Calculate descriptive statistics and visualize: Histograms, scatter plots and descriptive statistics were calculated to detect possible problems such as missing data or outliers.
- 8) Deal with data issues: As the centroid of the polygon based on the European Population Grid could fall over a problematic place such as river, park or railway lot (places where google maps cannot match an address), some routes could not be retrieved and consequently were left out of the study.
- 9) Calculate KPIs: Based on research done in the past the following can be derived:
 - Time difference:
 - Travel Time Index (TTI):
- 10) Aggregate data generated and join to spatial grid: With the data retrieved averages by the origins were calculated and joined to these values joined to the original zoned urban area.
- 11) Calculate descriptive statistics and visualize: The results at the Grid level are analyzed using descriptive statistics and basic plotting. This step can be seen as check-up point that can contribute to identify data anomalies such as missing data or outliers.
- 12) Map the results: Finally, once the data clean-up process is done, a map with the processed variables can be generated.

To demonstrate how the method could be used in practice, four European cities were selected as a proof of concept.

4. Data

In order to run the analysis for a desired urban area, a zoned cartography of the area is needed. In this opportunity, as a proof of concept the 1km² population grid from the EuroStat (GEOSTAT 2011) was used. Amsterdam, Glasgow, Goteborg and Lisbon were selected and from their historical centre a buffer zone was used to delimit the *city* boundary. The geographic coordinates of the centroid of each 1km² square was used to create the list of all possible routes.

From these list of origins a synthetic Origin-Destiny matrix was created: all possible origins to all possible destinations was then used to retrieve data for a *congested* and *non-congested* moments.

4.1. Data retrieved

The process generates information for a total $n*(n-1)$ routes (n being the number of zones). For instance, Lisbon is divided into 119 squares and the number of possible routes is $119*118 = 14,042$ (then 28,084 travels were estimated). Amsterdam has 131 zones, Glasgow 136 and Goteborg 123 (17,030, 18,360 and 15,006 routes respectively). A summary of the descriptive statistics for the time in minutes can be found in Table 1.

	Congested scenario (mins)	Routes	Mean	S.D.	Min	Max
1	Amsterdam	17,023	20.26	6.86	0.65	48.57
2	Glasgow	18,355	23.08	8.68	1.42	56.88
3	Goteborg	14,997	18.01	6.82	1.37	57.97
4	Lisbon	14,043	28.16	11.70	1.37	84.63
	Non-Congested scenario	Routes	Mean	S.D.	Min	Max
1	Amsterdam	17,023	11.70	3.71	0.52	24.90
2	Glasgow	18,355	11.93	3.99	0.83	24.85
3	Goteborg	14,997	13.09	5.29	1.10	49.18
4	Lisbon	14,043	12.82	5.77	0.77	46.90

Table 1: Descriptive Statistics of travel times by city

The amount of route information retrieved is consistent with the amount of zones in the city and as expected, the *congested* situation on average takes longer time. Formally an independent-samples t-test was conducted to compare the travel times under a *congested* and *non-congested* situations. For each city, the time difference was found statistically significant with a confidence level of 99%.

5. Results

After capturing, processing and aggregating the data obtained from the API, table 2 on page 5 contains the descriptive statistics for time difference and TTI. In this case, the number of observations corresponds to the number of zones in the map and each zone holds the mean of all the trips departed from that origin. The results show that the methodology successfully captured diversity between cities. For instance, in Lisbon the average time difference is of 15.3 mins, with a maximum of 21.6 mins, meanwhile in Goteborg on average delays are of 4.9 mins with a maximum of 8 mins.

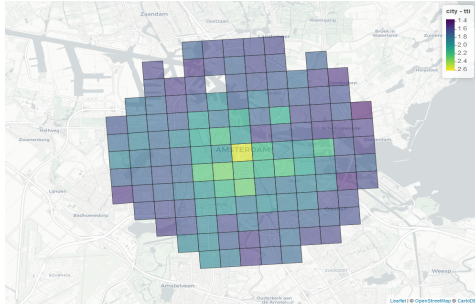
Although, the table above reveals relevant insights about these cities, local authorities or planners would not know which places or who is being affected by traffic congestion. Therefore, in figure 1 the TTI for each city was mapped. By looking at these maps several conclusions about where and how bad congestion is across the city. For instance, in all four cities, although the historical centre is the nearest point to all other destinations, when congestion is taken into

	Time Difference (mins)	Zones	Mean	S.D.	Min	Max
1	Amsterdam	131	8.57	2.39	5.38	19.03
2	Glasgow	136	11.15	1.88	7.66	15.64
3	Goteborg	123	4.92	0.99	3.22	8.06
4	Lisbon	119	15.34	2.67	9.74	21.57
	TTI	Zones	Mean	S.D.	Min	Max
1	Amsterdam	131	1.74	0.23	1.36	2.61
2	Glasgow	136	1.94	0.14	1.60	2.31
3	Goteborg	123	1.38	0.09	1.21	1.69
4	Lisbon	119	2.21	0.16	1.70	2.57

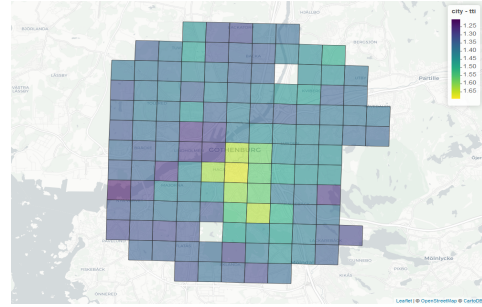
Table 2: Aggregated descriptive statistics by zones

account, these places face the highest TTI. This indicates that in a *non-congested* situation, the city centre is highly accessible but when *congested*, the advantage of being central gets eroded.

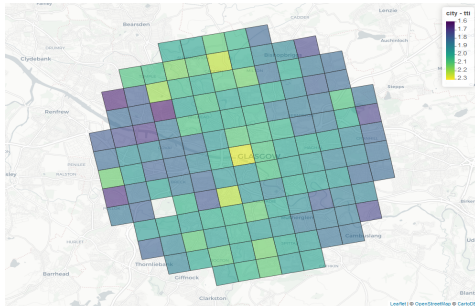
These maps also show that the methodology presented in this study was able to capture differences across cities successfully. In Goteborg, the historical city centre gets mainly congested, meanwhile the other parts of the city remain with similar levels of congestion. In Lisbon, traffic congestion is more scattered across the city, presenting islands of low congestion (where the CBD and commercial areas have moved to).



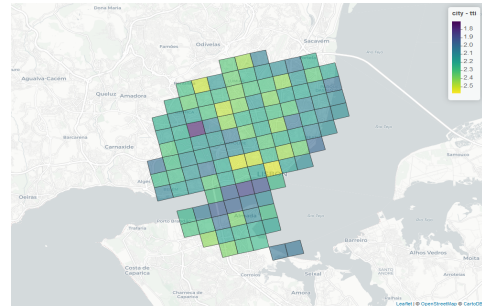
(a) Amsterdam



(c) Goteborg

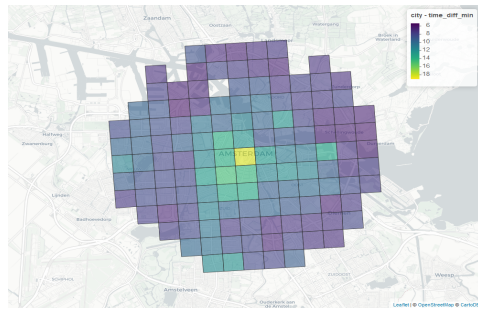


(b) Glasgow

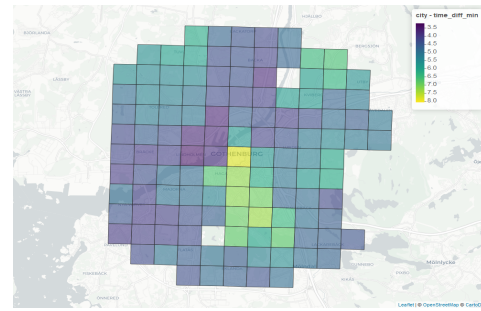


(d) Lisbon

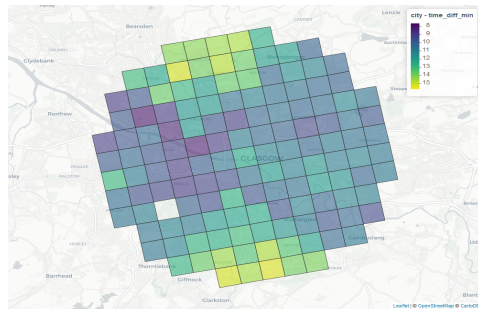
Figure 1: Images extracted from results



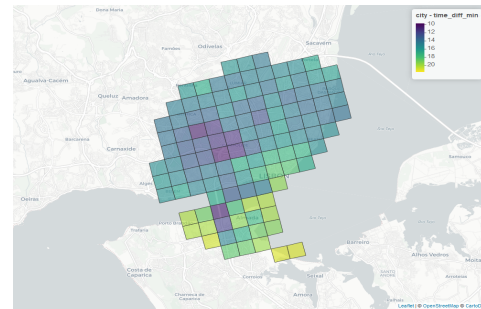
(a) Amsterdam



(c) Goteborg



(b) Glasgow



(d) Lisbon

Figure 2: Images extracted from results

6. Discussion

The methodology presented in this paper exploits a new data source to provide spatial insights about traffic congestion in different cities. The information generated allows planning and transport agencies to reconstruct some of the most popular indexes discussed in the theoretical background section.

With completely different business models, INRIX, HERE and TomTom are two internationally known companies which in the last years started to offer similar transportation insights as a service. Historically, INRIX has been consulted to provide the levels of congestion in several countries in the OECD and U.S. transport agencies, but given the increase in the amount of devices embedded with GPS, other businesses have found themselves in a position of exploiting these new data sources. Nowadays, they all offer transportation consulting services. INRIX and TomTom publish reports of traffic congestion and rank cities according to different criteria. [SHOULD WE SHOW A A GRAPH WITH THIS?] The aggregated results producing this methodology are consistent with those produced by both companies. Moreover, the methodology proposed in this paper allows practitioners to visualize how congestion is spatially distributed across the city.

The results presented here, took the EuroStat population 1km² grid cell to retrieve traffic information. Certainly, this decision was arbitrary and the shape/size of these areas can be subject of debate. Nevertheless, the zones used for the analysis can be easily interchanged allowing the process to be enriched.

Traffic congestion deteriorates different domains of urban life and as a consequence the focus on the problem varies across disciplines. The maps of congestion produced as a result of applying this methodology can be used to understand who suffers from congestion the most and what parts of the city are more vulnerable to the problem.

6.1. Examples to cite

As shown 1.Lomax, Turner, and Shunk 1997

2.(Lomax, Turner, and Shunk 1997)

3.Lomax, Turner, and Shunk (1997)

6.2. Limitations and future work

The methodology presented in this study relays primarily in Google's web service and this fact is a drawback as users have no control over the service standards, usability or costs. For instance, the type request performed during this study had a cost of 10 U\$D/1000 requests, considering a grid of 100 areas, will generate 9,900 travel routes and to estimate congestion a total of 19,800 requests will be needed. This process will involve a cost of 198 U\$D.

This relates to future work; to explore how datasets produced from other data sources such as HERE or TomTom can be compared from the results extracted here. Costs of operation of the method. give details of the specific ones will vary with the granularity and the coverage will vary with the data source

6.3. Applications

As shown Lomax, Turner, and Shunk 1997 Traffic congestion measurements can be used

6.4. Limitations and future work

Grid size and costs In this case

Weights using population using an OD matrix

1km grid size can be too coarse for spatial planning, has impact o

use census tracks or population statistics to correlate with socio-demographics

Reviewer Remarks:This is a very interesting paper proposal. It will be interesting to learn how the impact of different types of car engines (petrol, diesel, electric, etc) will impact CO2 emissions in congested and non-congested traffic flows.

7. Conclusion

The study presented a methodological approach to study the how traffic congestion is spatially distributed within an urban area. Generating a synthetic Origin-Destiny matrix, the method uses an online routing service to estimate different travel routes. The methodology provides a non-expensive, generalizable and systematic to estimate congestion in different parts of a city. The data retrieved can be mapped, thus used by city planners and transportation agencies. The aggregation of the data provide conclusions that are consistent with internationally recognized institutions such as INRIX and TomTom.

References

Lomax, Tim, Shawn Turner, and Gordon Shunk (1997). *NCHRP Report 389: Quantifying Congestion Volume 1 - Final Report*. Tech. rep. Texas Transport Institute (TTI), p. 102. URL: <http://worldcat.org/isbn/0309060710>.