

Assessment of approximation method for TSP path length on road networks: a simulation study

Bachelor Thesis

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Abstract

1 Introduction

The Traveling Salesman Problem (TSP) is an important problem in operations research. It is particularly relevant for last-mile carriers and other logistics companies where efficient routing directly impacts cost, time and service quality. Since the number of parcels worldwide has increased between 2013 and 2022 and is expected to keep increasing [Statista (2025)], the need for fast, scalable route planning methods becomes ever more pressing.

The TSP is an NP-hard problem, it is computationally intensive to find the exact solution for large instances. In many real-world scenarios, the exact optimal routes may not be needed, but instead a rough, reliable estimate of the optimal route length. For instance, consider a postal delivery company. This firm may need to assign a certain amount of deliveries or a certain area to each postman. Reliable estimates for the route length can provide valuable information for making such decisions.

Efficient approximation methods provide a solution for such practical applications where exact solutions are too computationally intensive to conduct or not feasible due to insufficient data. These methods aim at approximating the expected optimal total travel time or distance, while using minimal data and computational effort.

There is extensive research on such approximation methods and how they perform in the euclidean plane. Consider n uniformly drawn locations from some area in \mathbb{R}^2 with area A . Beardwood, Halton, and Hammersley (1959) prove the relation:

$$L \rightarrow \beta\sqrt{nA}, \quad \text{as } n \rightarrow \infty \tag{1}$$

as an estimation for the length of the shortest TSP path measured by Euclidean distance through these random locations, where β is some proportionality constant. This formula is a very elegant result, and it requires very little data. However, its assumptions, uniform random locations and euclidean space differ from real-world applications, which are defined by complex geographic features, such as road networks.

This research investigates how well this approximation method performs when we consider real road networks. Using OpenStreetMap data, we simulate TSP instances in a wide variety of different urban areas in the Netherlands, then solve these for the actual shortest paths using the Lin–Kernighan heuristic. Then we estimate the β from equation 1 and analyze the performance of this formula. Additionally, we compare the results for β and the performance across the selected areas.

In section 2 we dive deeper in the context and previous research in this field. Then, in section 3 we show the methodology ...

2 Literature Review

In this section we review the existing literature on the Beardwood formula and some applications, and on the Lin-Kernighan heuristic and its implementations.

2.1 Applications of the Beardwood formula

We are interested in the performance of formula 1 for reasonable amounts of locations a delivery person can visit in a workday, say $10 \leq n \leq 90$. Table 1 lists values for β for some values of n , where the points were generated uniformly and the L_2 distance metric was used.

Table 1: Empirical estimates of β as a function of n , $20 \leq n \leq 90$ (Lei et al. (2015))

n	$\beta(n)$
20	0.8584265
30	0.8269698
40	0.8129900
50	0.7994125
60	0.7908632
70	0.7817751
80	0.7775367
90	0.7773827

Figliozi (2008) is the first research to apply approximation formulas to real-world instances of TSPs (and VRPs (Vehicle Routing Problems)). An extension of formula 1 that works for VRPs is assessed in a real-world setting. It is found that this model has an R^2 of 0.99 and MAPE (Mean Absolute Prediction Error) of 4.2%. This prediction error is slightly higher than when it is applied to a setting where euclidean distances are considered (3.0%), but the formula still performs well.

Merchán and Winkenbach (2019) use circuitry factors to measure the relative detour incurred for traveling in a road network, compared to the euclidean distance. This circuitry factor is defined as, where p and q are locations:

$$c = \frac{d_c(p, q)}{d_{L_2}(p, q)} \quad (2)$$

By construction, c is greater or equal to 1, a value closer to 1 indicates a more efficient network. Then, β_c is estimated by $\beta_c = c\beta$. This value c , is estimated for three different areas in São Paulo, for which the results are listed in table 2. These values indicate real travel distances are on average 2.76 times longer in area 1 compared to the L_2 metric. These values were obtained by uniformly generating n locations (for n ranging from 3 to 250), computing near-optimal tour lengths under the Euclidean metric, and solving for β , then scaling by the empirical circuitry factor.

Table 2: Estimates of the circuitry factor c and its corresponding β_c (Merchán and Winkenbach (2019))

	Area 1	Area 2	Area 3
c	2.76	2.34	1.82
β_c	2.48	2.10	1.64

It is important to note, however, that the assumptions in this study may limit the generality of the findings. In particular, the use of uniformly distributed locations does not accurately reflect the spatial distribution of delivery points in real urban environments, where locations tend to cluster in residential, commercial, or industrial zones. Additionally, within small urban areas, high-rise buildings and single-family homes may coexist in the same neighborhoods, further challenging the assumption of uniformly distributed delivery points. Furthermore, the circuitry factor c can vary significantly within a single city, depending on local street patterns, infrastructure, and topography. These variations suggest that a fixed circuitry factor may oversimplify the complexity of real-world delivery contexts, especially when applied to smaller subregions or neighborhoods.

2.2 Lin-Kernighan Heuristic

3 Methodology

4 References

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5 Appendix

Province	Neighborhood	Beta
groningen	Hortusbuurt	2.2197
groningen	Binnenstad	2.0428
groningen	Oosterpoort	2.0359
groningen	Rivierenbuurt	1.7779
groningen	De Wijert	1.7561
groningen	Oosterparkwijk	1.7698
groningen	De Hoogte	1.7028
groningen	Korrewegwijk	2.0592
groningen	Schildersbuurt	2.2240
groningen	Paddepoel	1.6704
groningen	Oranjewijk	1.9470
groningen	Tuinwijk	2.8923
groningen	Selwerd	1.5238
groningen	Vinkhuizen	1.4784
groningen	Hoogkerk-zuid	1.5556
groningen	Gravenburg	1.3061
groningen	De Held	1.9358
groningen	Reitdiep	1.6557
groningen	Hoornse Meer	1.5883
groningen	Corpus den Hoorn	1.6074
groningen	Eemspoort	1.7349
groningen	Euvelgunne	1.9058
groningen	Driebond	1.8945
groningen	Winschoterdiep	1.9701
groningen	Eemskanaal	1.7716

Province	Neighborhood	Beta
groningen	Helpman	2.0734
groningen	Lewenborg	1.9133
groningen	Beijum	1.8017
groningen	Maarsveld	1.6757
noord _{holland}	Schrijverswijk	1.7461
noord _{holland}	Stad van de Zon	1.4595
noord _{holland}	Stadshart	1.4584
noord _{holland}	Jordaan	1.8419
noord _{holland}	Slotervaart	1.7458
noord _{holland}	IJburg	1.3445
noord _{holland}	Oostelijke Eilanden	1.7011
noord _{holland}	Oostelijk Havengebied	1.7282
noord _{holland}	Frederik Hendrikbuurt	2.2905
noord _{holland}	Van Lennepbuurt	1.8079
noord _{holland}	Da Costabuurt	2.5017
noord _{holland}	Kinkerbuurt	1.9756
noord _{holland}	Kersenboogerd	1.6555
noord _{holland}	Pax	2.1976
noord _{holland}	Graan voor Visch	2.2232
noord _{holland}	Vrijschot-Noord	2.4097
noord _{holland}	Toolenburg	1.2980
noord _{holland}	Floriande	1.9087
noord _{holland}	Overbos	1.7870
noord _{holland}	Bornholm	1.8147
noord _{holland}	Beukenhorst-Oost	1.6793
noord _{holland}	De Hoek	2.5402
noord _{holland}	West	2.0184
noord _{holland}	Zuid	1.6708
noord _{holland}	Oost	1.8782
noord _{holland}	Noord	1.6478
noord _{holland}	De President	1.4994
noord _{holland}	Graan voor Visch-Zuid	1.7247
noord _{holland}	Zuidwijk	1.3947
noord _{holland}	Buitenveldert-West	1.1970
noord _{holland}	Buitenveldert	1.1434
noord _{holland}	Apollobuurt	1.7353
noord _{holland}	Stadionbuurt	1.4852
noord _{holland}	Prinses Irenebuurt e.o.	1.8953
noord _{holland}	Hoofddorppleinbuurt	1.6765
noord _{holland}	Willemspark	1.9443
noord _{holland}	Schinkelbuurt	1.6466
noord _{holland}	Vondelparkbuurt	1.2894

Province	Neighborhood	Beta
noord _{holland}	Helmersbuurt	1.8133
noord _{holland}	Overtoomse Sluis	1.9465
noord _{holland}	Museumkwartier	1.7108
noord _{holland}	Rivierenbuurt	1.6913
noord _{holland}	IJselbuurt	1.5742
noord _{holland}	Scheldebuilt	1.3664
noord _{holland}	Rijnbuurt	1.6533
noord _{holland}	De Baarsjes	1.8024
noord _{holland}	Landlust	1.7613
noord _{holland}	Staatsliedenbuurt	1.8155
noord _{holland}	Spaarndammerbuurt	2.2497
noord _{holland}	De Pijp	2.2851
noord _{holland}	Grachtengordel	1.7890
noord _{holland}	Oud-Zuid	1.4385