

Chapter 24

Discovering Gathering Pattern Using a Taxicab Service Rate Analysis Method based on Neural Network

Junming Zhang

Beijing University of Posts and Telecommunications, China

Jinglin Li

Beijing University of Posts and Telecommunications, China

ABSTRACT

Moving objects gathering pattern represents a group events or incidents that involve congregation of moving objects, enabling the analysis of traffic system. However, how to improve the effectiveness and efficiency of the gathering pattern discovering method still remains as a challenging issue since the large number of moving objects will generate high volume of trajectory data. In order to address this issue, the authors propose a method to discovering the gathering pattern by analyzing the taxicab demand. This paper first introduces the concept of Taxicab Service Rate (TSR). In this method, they use the KS measures to test the distribution of TSR and calculate the mean value of the TSR of a certain time period. Then, the authors use a neural network based method Neural Network Gathering Discovering (NNGD) to detect the gathering pattern. The neural network is based on the knowledge of historical gathering pattern data. The authors have implemented their method with experiments based on real trajectory data. The results show the both effectiveness and efficiency of their method.

1. INTRODUCTION

The increasing availability of location acquisition has been applied to GPS on vehicles. The technology has enabled tracking almost any moving behaviors of moving objects, which results in huge volumes of spatio-temporal data in the form of trajectories. Movement pattern can be observed from this kind of data. The movement pattern will provide useful information for traffic jam prediction and traffic flow control.

The movement pattern has been well studied for the past few years (L. Huang, Wang, Hsu, Zhang, & Yang, 2015; Jeung, Yiu, Zhou, Jensen, & Shen, 2010; Kalnis, Mamoulis, & Bakiras, 2005; Li, Ding, Han, & Kays, 2010; S. Wang, Hsu, Liang, Sun, & Yang, 2014; Zheng, Zheng, Yuan, & Shang, 2013) (Jeung, Yiu, Zhou, Jensen, & Shen, 2010). In order to support the traffic analysis application in cities, Zheng (Zheng et al., 2013) has proposed a useful movement pattern which name is gathering. It represents a behavior of a set of spatio-temporal moving objects, which move together within a certain period of time. The author has also proposed a gathering pattern discovery method by counting the gathering number. Specifically, it uses the spatial Gird to index the crowd of moving objects, and then uses the Crowd-TAD method to test each crowd to discover the gathering pattern. To improve the effectiveness and efficiency of the gathering pattern retrieving process, we have also proposed a gathering retrieving method based on spatio-temporal graph that forms by moving object clusters (Zhang JM, 2016). The main part of the method is to find the maximal complete graph that meets the spatio-temporal constraints by indexing the graph. But it still took a lot of time to cluster the moving objects and to search graph. Recently, some researchers use the taxicab demand to analyze the traffic condition or find a fast driving route (Liu, Zheng, Chawla, Yuan, & Xing, 2011; Ma, Zheng, & Wolfson, 2013; Rui, Hacker, & Rong, 2014; Yuan, Zheng, Xie, & Sun, 2011). Researcher (Yuan, Zheng, Zhang, Xie, & Sun, 2011) presented a recommender system by using the taxicab drivers' pick-up behaviors. Our previous research has also proposed method to analyze the taxicab demand (Zhou, Zhang, Li, & Wang, 2014). From this point, we could use the taxicab demand to discover the gathering pattern. Even though it need to calculate the taxicab demand from the trajectory data, it will save a huge time comparing with the moving object clustering and graph searching process.

In this paper, we propose a method to retrieve the gathering pattern by analyzing the taxicab demand. This paper first introduces the concept of Taxicab Service Rate (TSR), which can reflect the real demand of taxicab in different regions. The method has two processes: TSR calculating and gathering pattern discovery. In the TSR calculating process, first, we segment the whole map to disjoint regions r bounded by the grid. Second, we use the KS measures to test the distribution of TSR at a certain time period. At last, we use an improved Parzen window method to estimate the Probability Density Function (PDF) of TSR and calculate the mean value of the TSR in a specific time period. In the gathering pattern detecting process, we use a neural network based method NNGD to detect the gathering pattern. Firstly, historical gathering pattern data are used to train the network. Then, we use a four-layer feed-forward network to classify the movement pattern to gathering pattern or non-gathering pattern. In this paper, we make the following contributions:

- This paper introduces a new point of view to discover the gathering pattern. Firstly, we use the Parzen window method to calculate the mean value of the TSR in a specific time period. Then neural network based method NNGD is used to discover the gathering pattern.
- This paper introduces the concept of TSR to measure the taxicab demand instead of the requests number, which is hard to count. Then, we use the KS measures to test the distribution of TSR at

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