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Smart Card clustering to extract typical temporal passenger habits in Transit network. Two case studies: Rennes in France and Gatineau in Canada

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Smart card data are increasingly used for the analysis of urban mobility in particular multimodal and/or intermodal mobility. Their richness from spatial, temporal and quantitative point of view made them good material for mobility analysis. Indeed, they can be used to predict passenger flows, as well as to extract relevant information on transport network operation. They can also be used to better understand passenger behaviours. In the latter, statistical models are built to automatically extract a reduced set of clusters, each cluster contains a subset of passengers sharing the same temporal habits in terms of public transport usage.

In this paper we will investigate a statistical modeling to perform the clustering of passengers based on their ticketing logs in the public transport. The aim is partition the passengers into groups on the basis of their travel hours. The clustering is proposed to be performed in unsupervised way by using advanced data partitioning tools, that is dedicated Gaussian mixture models. Doing so, we will be able to extract typical patterns describing different types of transport usage, namely sporadic usage, typical home-work commute behavior, scholar usage etc.

Considering two case studies of the city of Rennes in France and the city of Gatineau in Canada, three main contributions will be illustrated in this paper:

- The first contribution is related to the statistical modeling based on a two-level generative model. Unlike different clustering approaches proposed in the litterature that consider a discrete representation of time (e.g. trips are aggregated over pre-defined time slots), the proposed model allows to take into account the continuous nature of timestamps. The obtained distributions offer a direct way to interpret the results in terms of times of typical transport usage as well as in terms of variance around these time peaks of usage.
- The second contribution is related to the way to deal with the type of the day in the statistical model. It will be shown that this choice is closely linked to the public transport usage in the city. A categorization weekday/weekend is enough sufficient in Gatineau whereas a model accounting for all the days of the week is more appropriate to highlight the passenger habits in Rennes. This is mainly due to the temporal organization of scholars in France most of which do not study on wednesday afternoon. Moreover, this is also due to the

usage of public transport for nighlife activities or for leisure activities in the weekend.

The third contribution is related to the analysis of different temporal activities that could be revealed by such method and link those activities with the city characteristics. By crossing the clustering results with spatial organization of both cities, it is indeed possible to have a better understanding of passenger activity on the transportation network. It will be shown that passengers, whom city is a centre of activity (Rennes) will be more likely to go out at night or during the weekend than passengers, whom city is located in the suburbs, which is the case of Gatineau. The results also highlight cultural differences in the use of public transportation between France and Canada.

Experiments are carried out on two real datasets. The first data set was collected on the public transport network of Rennes Metropole (France) and provided by the STAR network (Service de Transport en Commun de l'Agglomération Rennaise). The network is composed of one metro line (ligne a) and around 135 bus lines. 43 municipalities containing more than 400,000 inhabitants are serviced by this network. The second dataset was provided by the Société de Transport de l'Outaouais (STO) based in Gatineau, Canada. As a medium-sized public transit authority, the STO operates 310 buses and services 291,000 inhabitants. The STO has been operating its smart card system since 2001.

The obtained results exhibits different cluster profiles that can be used to differentiate between passengers with regular travel hours and those with diffuse travel hours. The approach is also able to detect subtle trends such as the presence of nightlife activity. The results provided by our method could help the transport operators to have a better knowledge of their passengers' behaviour patterns and adapt their offer to suite the different demands they portray (e.g. specific card type for passengers' activity peak than classic aggregate time clustering is provided.

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