

### Survey on Location Based Recommendation System Using POI

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**Abstract** – POI based recommendation system is a web based application, That facilitates, analysis of an individual. The point of interest (POI) is geographical can also be temporal. Such a system would prove valuable tool in the inventory of a corporation or individual. The services provided By this system could be optimized with the consideration of user interest drift. Development of mobile devices with parallel advancement in GPS (Global Position System) and web technology unable the user to share data information based on its relevance to application user. This information is gathered using data mining technique. It is important to foe users

Key Words: Data Mining, POI recommendation, User Interest Drift, Collective Inference, User Modelling.

#### **1.INTRODUCTION**

In modern world with the rapid development of mobile devices web technologies and point of interest is becoming very popular now days locations revolutionized communication in social temporal with their friends they share their locations and photos uploading via internet for POI.

To make use of the rich information on social relationships Users recommend the new locations to explore them. These data is then received and represents it in different formats like HTML, JavaScript and CSS.

It has been traditionally used to communicate with the web services. It also provides the messaging services. On the internet where the number of choices is overwhelming, there is need to filter, prioritize and efficiently deliver the relevant information in order to alleviate the problem of information overload, which create a potential problem too many users.

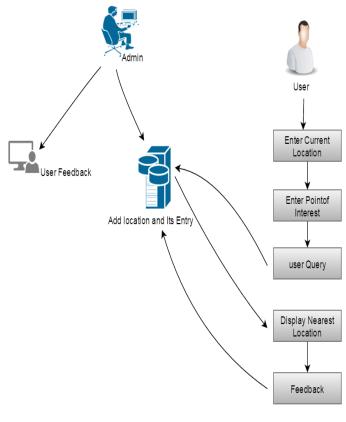
Recommender system solves this problem by searching through the large volume of dynamically generated information to provide users with personalized content which are in user's point of interest. So it is easy for users to Drift across the region.

#### 1. Data Mining

Data mining is the process of discovering actionable information from large sets of data. Data mining is typically uses the mathematical analysis to derive the patterns that exist in data.

In data mining these patterns cannot be discovered by traditional data exploration because the relationships are too complex or maybe there is too large amount of data.

#### **1.2 System Architecture**



**Fig. System Architecture** 



#### 2. Advantages

We proposed a socialspatial collective inferreing framework to enhance the interface of users' interests in each region, out-of-town regions, by exploiting the social and spatial correlation information.

We exploit the temporal patterns of the public visiting POIs to improve the process of topic discovery.

#### 3. Sequence diagram

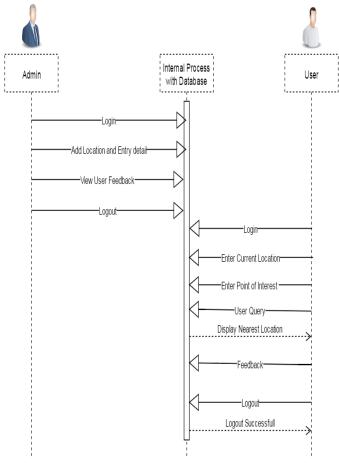


Fig. Sequence Diagram

#### 4. Literature survey

#### 1. Latent dirichlet allocation.

#### AUTHORS: D. M. Blei, A. Y. Ng

We describe latent Dirichlet allocation (LDA), a generative probabilistic model for collections of discrete data such as text corpora. LDA is a three-level hierarchical Bayesian model, in which each item of a collection is modeled as a finite mixture over an underlying set of topics. Each topic is, in turn, modeled as an infinite mixture over an underlying set of topic probabilities. In the context of text modeling, the topic probabilities provide an explicit representation of a document. We present efficient approximate inference techniques based on variational methods and an EM algorithm for empirical Bayes parameter estimation. We report results in document modeling, text classification, and collaborative filtering, comparing to a mixture of unigrams model and the probabilistic LSI model.

## 2. Fused matrix factorization with geographical and social influence in location-based social networks.

#### Author: C. Cheng, H. Yang, I. King,

Recently, location-based social networks (LBSNs), such as Gowalla, Foursquare, Facebook, and Brightkite, etc., have attracted millions of users to share their social friendship and their locations via check-ins. The available check-in information makes it possible to mine users' preference on locations and to provide favorite recommendations. Personalized Point-of-interest (POI) recommendation is a significant task in LBSNs since it can help targeted users explore their surroundings as well as help third-party developers to provide personalized services. To solve this task, matrix factorization is a promising tool due to its success in recommender systems. However, previously proposed matrix factorization (MF) methods do not explore geographical influence, e.g., multi-center check-in property, which yields suboptimal solutions for the recommendation. In this paper, to the best of our knowledge, we are the first to fuse MF with geographical and social influence for POI recommendation in LBSNs. We first capture the geographical influence via modeling the probability of a user's check-in on a location as a Multi-center Gaussian Model (MGM). Next, we include social information and fuse the geographical influence into a generalized matrix factorization framework. Our solution to POI recommendation is efficient and scales linearly with the number of observations. Finally, we conduct thorough experiments on a large-scale real-world LBSNs dataset and demonstrate that the fused matrix factorization framework with MGM utilizes the distance information sufficiently and outperforms other state-of-theart methods significantly.

# 3. Where you like to go next: Successive point-of-interest recommendation.

#### Author: C. Cheng, H. Yang, M. R. Lyu,

Personalized point-of-interest (POI) recommendation is a significant task in location-based social networks (LBSNs) as it can help provide better user experience as well as enable third-party services, e.g., launching advertisements. To provide a good recommendation, various research has been conducted in the literature. However, pervious efforts mainly consider the "check-ins" in a whole and omit their temporal relation. They can only recommend POI globally and cannot know where a user would like to go tomorrow or in the next few days. In this paper, we consider the task of successive personalized POI recommendation in LBSNs, which is a much harder task than standard personalized POI recommendation or prediction. To solve this task, we observe two prominent properties in the check-in sequence: personalized Markov chain and region localization. Hence, we propose a novel matrix factorization method, namely FPMC-LR, to embed the personalized Markov chains and the localized regions. Our proposed FPMC-LR not only exploits the personalized Markov chain in the check-in sequence, but also takes into account users' movement constraint, i.e., moving around a localized region. More importantly, utilizing the information of localized regions, we not only reduce the computation cost largely, but also discard the noisy information to boost recommendation. Results on two real-world LBSNs datasets demonstrate the merits of our proposed FPMC-LR.

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#### REFERENCES

- 2. C. Cheng, H. Yang, M. R. Lyu, and I. King. Where you like to go next: Successive point-of-interest recommendation. In *IJCAI*, pages 2605–2611, 2013.
- 3. H. Cheng, J. Ye, and Z. Zhu. What's your next move: User activity prediction in location-based social networks. In *SDM*, pages 171–179, 2013.
- 4. Z. Cheng, J. Caverlee, K. Lee, and D. Z. Sui. Exploring millions of footprints in location sharing services. In *ICWSM*, 2011.
- 5. G. Ference, M. Ye, and W.-C. Lee. Location recommendation for out-of-town users in location-based social networks. In *CIKM*, pages 721–726, 2013.