KENTUCKY Recommendation System with Minimized Transaction Data Dr. Yangyang Tao, Dr. Junxiu Zhou, Leah Strecker, Matt Longwell, Olivia Ryan, and Kristoffe Wilson School of Computing and Analytics | Northern Kentucky University

Abstract

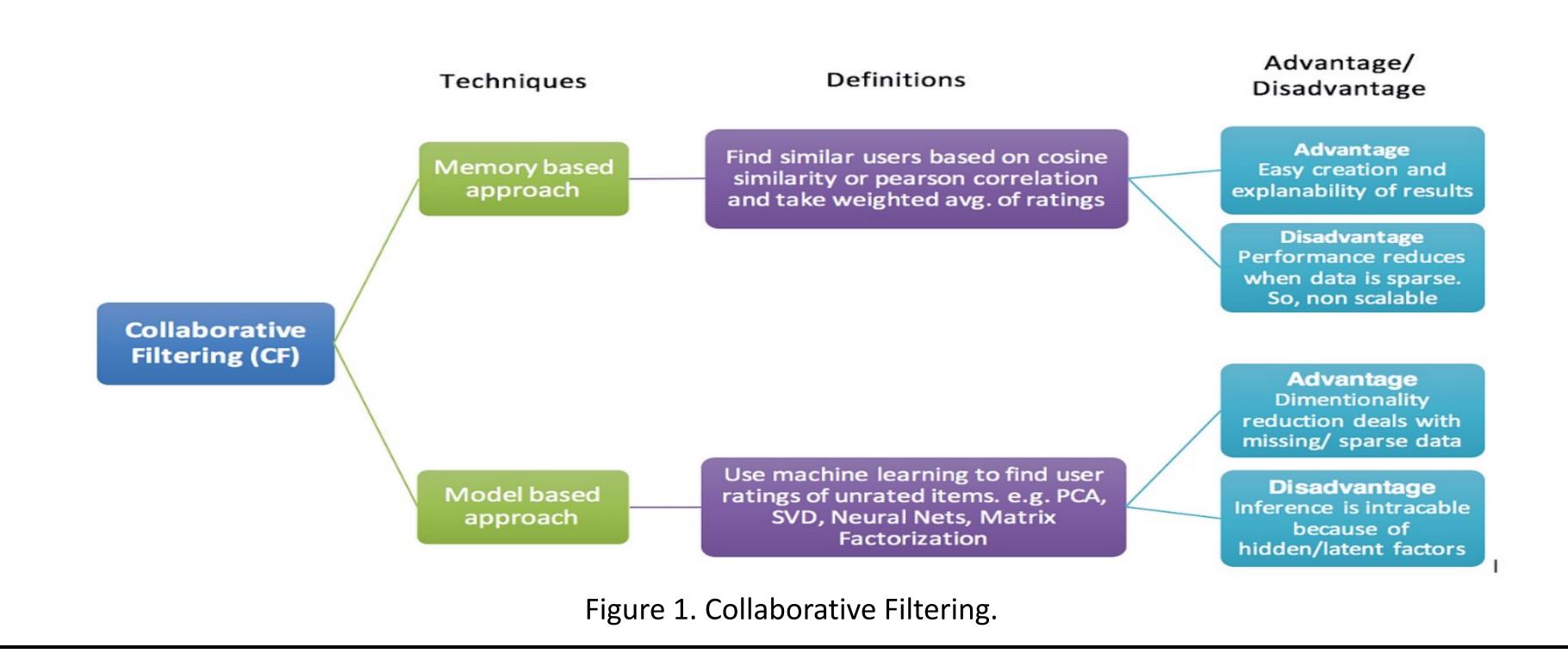
This work is a course research project that with Fifth Third Bank. The collaborates motivation of the project is that through collaboration reputable with external organizations, it has become apparent that these organizations, while eager to foster with educational partnerships institutions, in sharing encounter substantial limitations particularly comprehensive datasets, concerning customer personal information. The purpose of this project is to use the user transaction dataset which contains the daily activities address transaction the to development of an AI-based recommendation system. The issue here is. In this unique setting, the AI model is built on only essential purchase data. The study focuses on leveraging minimal the transaction enhance data to recommendation system's performance.

Datasets

The dataset we used is a transaction dataset from Fifth Third bank. The dataset captures a snapshot of user transactions, offering a view of consumer behaviors and nuanced spending habits. It provides valuable insights transactions, into diverse array the OŤ encompassing various merchant categories and spanning different locations. Each transaction is characterized by details such as the transaction amount, merchant information, and geographic coordinates, shedding light on the patterns of financial activity in different cities and countries. The dataset's temporal information, including transaction and posting dates, enables a temporal analysis of consumer spending trends. This dataset serves as a rich resource for researchers and analysts interested in delving into the dynamics of user-initiated payments, with potential applications in financial analytics, behavior studies, and the consumer development of intelligent recommendation systems.

Project Description

- Data Preprocessing
 - Within this stage, our primary emphasis will be directed towards the not only the thorough organization and structuring of the data but also, techniques applied to the bank transaction dataset.
- Model Design
 - We would like to implement collaborative filtering [1] (Figure 1) and Extrapolative Collaborative Filtering [2] is an advanced extension of traditional collaborative filtering methods in recommendation systems. While conventional collaborative filtering primarily relies on user-item further by incorporating additional contextual information to enhance the accuracy and relevance of recommendations.
- Evaluation and Optimization
 - Assessing the performance of the recommendation system using relevant metrics. Several metrics are commonly used to evaluate recommendation system performance.
 - Fine-tuning models and algorithms for improved accuracy and efficiency. Fine-tuning machine learning involves optimizing the parameters and configurations to enhance the system's performance.

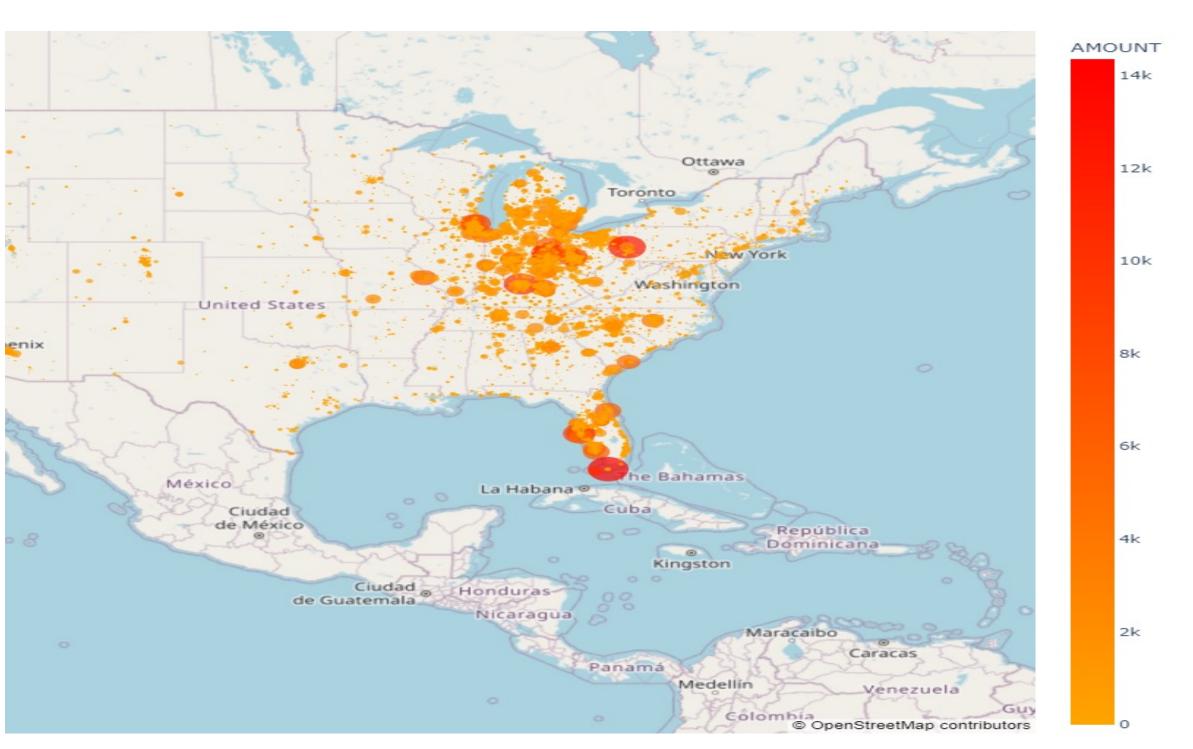


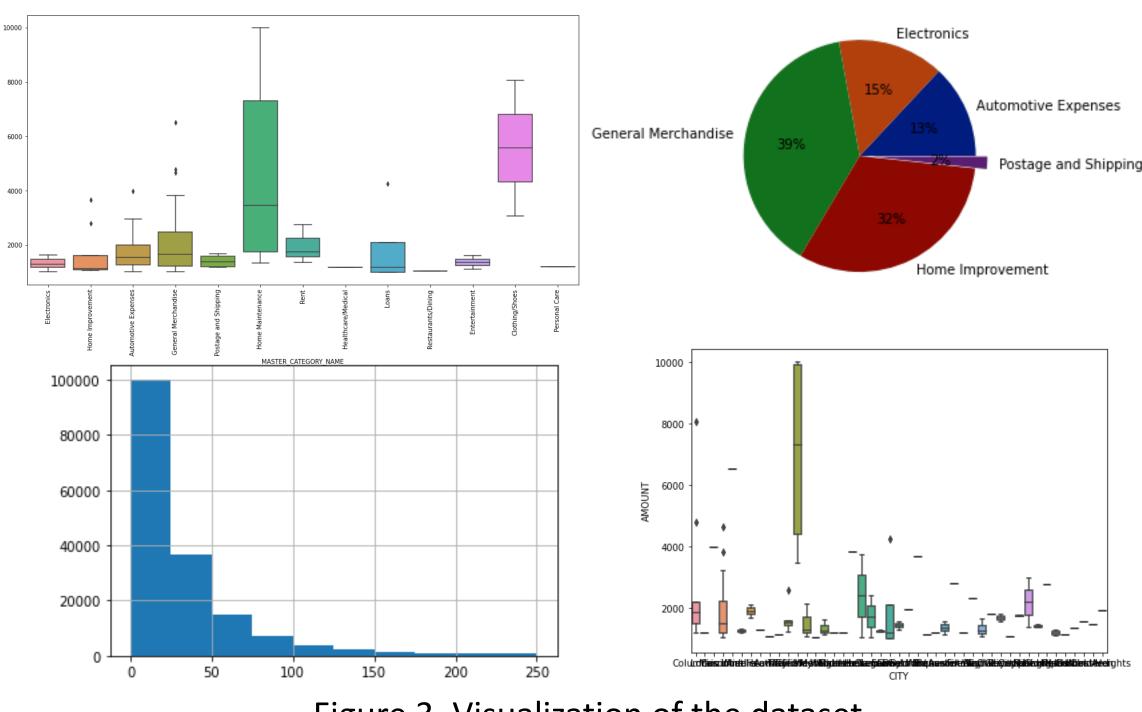
Results

So far, we have conducted a comprehensive preliminary analysis of the dataset, unearthing valuable insights into user transactions. The initial findings are particularly encouraging, revealing promising trends and patterns that have been visually elucidated in Figure 2 and Figure 3, enhancing our understanding of the intricate dynamics within the dataset.

meticulous preparation of the dataset. This intricate process encompasses significantly, the implementation of robust anonymization and minimization

extrapolative collaborative filtering models for recommendation generation. interactions and similarities, extrapolative collaborative filtering goes a step





Conclusion&Future Directions

So far, we have understood the dataset through various data exploration approaches. The next step is using a data transformation strategy such as PCA to hide the sensitive information and suitable models for developing the recommendation system which includes topics like segmentation, product recommendation, user fraud detection, visualization, (visualize the dataset with provided geographic info) and predict user spending amount.

References

456).

Figure 2. Visualization of the dataset.

Figure 3. Visualization of the dataset.

[1] Su, X., & Khoshgoftaar, T. M. (2009). A survey of collaborative filtering techniques. Advances in artificial intelligence, 2009.

[2] Koren, Y. (2009, June). Collaborative filtering with temporal dynamics. In Proceedings of the 15th ACM SIGKDD international conference on Knowledge discovery and data mining (pp. 447-