

Retail Data Analytics - Recommendation System for Products in Retail Business

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Abstract: *The field of retail data analytics is transforming the retail industry by employing sophisticated tools and methodologies to derive important insights from large datasets. Retailers may effectively evaluate consumer behavior, optimize inventory management, and improve overall operational efficiency by utilizing advanced analytics platforms like Power BI in conjunction with data manipulation tools like Python. Businesses may stay flexible in the cutthroat retail industry by using these insights to inform choices, customize marketing plans, and enhance consumer experiences. Thus, in the fast-paced world of retail, retail data analytics is essential for promoting data-driven decision-making and innovation.*

Keywords: Data Analytics, Dataset, PowerBI, Python

I. INTRODUCTION

Retail data analytics is a critical component of contemporary business strategy that entails using cutting-edge technologies and techniques to mine large datasets in the retail industry for insights that can be use. Python is essential in this ever-changing environment because of its robust libraries, such Random and Pandas. The process of generating and manipulating simulated data for thorough analysis involves the production of organized data frames. The smooth conversion of these data frames into CSV files makes it easier to integrate them with Power BI for reliable display. Businesses can successfully address goals, make educated decisions, and draw strategic solutions from the abundance of data generated within the retail arena by employing this integrated strategy. Moreover, the Pandas package for Python facilitates effective data organization and manipulation, guaranteeing the production of insightful data frames that faithfully capture the complexities of retail operations. Making use of the Random library improves the simulation element and offers a realistic framework for analysis. Businesses may quickly adjust to the constantly changing retail scene and obtain better insights by fusing the strengths of Power BI with Python libraries. This gives them a competitive advantage. Essentially, in the extremely dynamic retail industry, this integrated strategy equips decision-makers to handle difficult challenges, maximize operational efficiency, and remain at the forefront of innovation.

II. LITERATURE SURVEY

2.1 Retail Data Analytics

In recent years, the retail industry has witnessed a significant transformation driven by the proliferation of digital technologies and the abundance of data generated through various channels. Retailers are increasingly turning to data analytics to gain actionable insights into customer behavior, optimize operations, and enhance overall performance. This literature review examines key research and trends in retail data analytics, focusing on its applications, methodologies, and challenges.

Retail data analytics encompasses a wide range of applications, including customer segmentation, demand forecasting, pricing optimization, inventory management, personalized marketing, and store layout optimization. Various studies have demonstrated the effectiveness of data-driven approaches in improving decision-making processes and driving competitive advantage in the retail sector (Chen et al., 2012; Verhoef et al., 2015).

Despite its immense potential, retail data analytics faces several challenges and limitations. Data quality and integration issues, privacy concerns, and the complexity of omni-channel data are significant hurdles that retailers must address (Verhoef et al., 2017). Moreover, the scarcity of skilled data analysts and the need for organizational change pose

additional barriers to the successful implementation of data analytics initiatives in retail environments (Bose and Sugumaran, 2003).

Looking ahead, the future of retail data analytics lies in the integration of emerging technologies such as artificial intelligence, Internet of Things (IoT), and big data analytics. Real-time analytics and predictive modeling will become increasingly important for retailers seeking to adapt to dynamic market conditions and deliver personalized customer experiences (Huang and Benyoucef, 2013).

III. METHODOLOGY

Analysis of the data that comes from a shop or store which is also called as retail data. To provide useful insights and analysis for business development

Software requirements:

- a) Python
- b) Big Query
- c) Excel
- d) PowerBI
- e) SQL
- f) Advance SQL (Windowing Function)

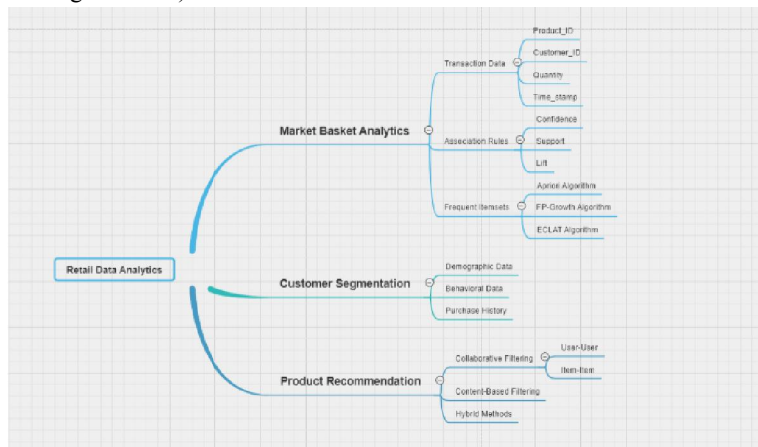


Fig. 1 : Architecture Diagram

1. Member Sale Analytics using Pareto Chart

Algorithms run through the data from the previous six months, finding the significant members (Top N) and using Pareto analysis to classify the members. Using Power BI's Pareto Analysis, you can visually examine and rank the elements that contribute to a specific outcome by applying the Pareto Principle, commonly referred to as the 80/20 rule. According to the theory, about 20% of the causes account for 80% of the consequences in many circumstances. This idea is frequently applied in the context of Power BI to assist businesses concentrate their efforts where they will have the greatest impact by highlighting the most important performance-influencing aspects.

According to the 80/20 rule, which was first proposed by Italian economist Vilfredo Pareto, a small percentage (20%) of the causes or inputs account for a substantial proportion (80%) of the outcomes or effects. This could entail determining and ranking the important variables in Power BI, like products

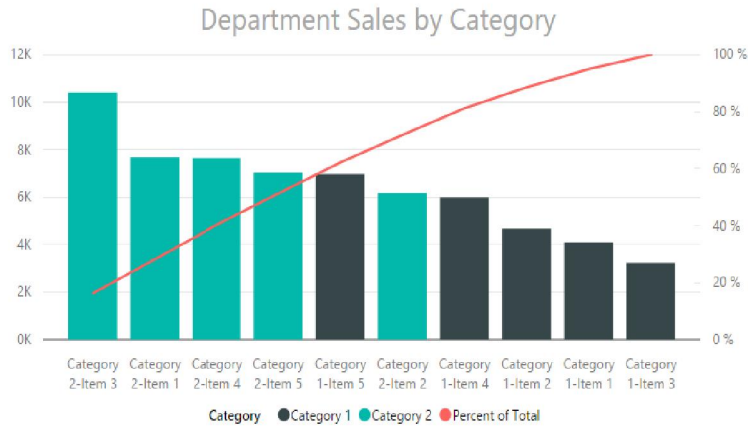


Fig.2: Pareto analysis

2. Member IPI Algorithm

An essential part of retail data analytics is the Inter-Purchase Interval (IPI) algorithm, which offers insightful information about consumer behavior and purchasing trends. The goal of this algorithm is to determine how long it takes for a consumer to make a successive purchase. The IPI algorithm determines the time interval between each member's subsequent purchases by using historical transaction data, giving a clear picture of their shopping cadence. This data is essential for seeing patterns, projecting the possibility of future purchases, and customizing marketing plans. While longer intervals could encourage focused re-engagement attempts, shorter intervals might suggest greater consumer engagement or loyalty.

3. Member Reward Score

Every month, the Member Reward Score Pipeline performs a thorough examination of rolling 12-month data to ascertain the reward scores for every member-item pair. The dynamic pipeline is based on an intricate computation procedure that takes into account the transactional data from the previous year, guaranteeing a refined comprehension of the way members interact with particular things. Using this rolling data strategy, the pipeline adjusts to changing member-item interaction patterns and trends. The computed reward scores are an important statistic that allows for customized reward classification and reflects the frequency of member-item transactions. In the constantly evolving field of retail analytics, this proactive and data-driven strategy enables businesses to customize rewards based on specific member behaviors, promoting loyalty and optimizing the efficacy of reward programs.

4. Item Seasonality Algorithm

Comprehending the seasonality of items is crucial for effective inventory management, marketing tactics, and comprehensive business planning. When it comes to ice cream, businesses may maximize supply chain management, promotional efforts, and stock levels by taking into account the anticipated spike in demand that occurs throughout the summer. Retailers can also target clients who have already expressed interest in seasonal items by personalizing their marketing strategies.

Item seasonality analysis provides valuable insights for strategic decision-making in the larger context of retail data analytics. These insights enable firms to better align their operations with customer preferences and market dynamics. Retailers can obtain actionable insights to improve overall business performance and customer satisfaction by utilizing SQL and Power BI.

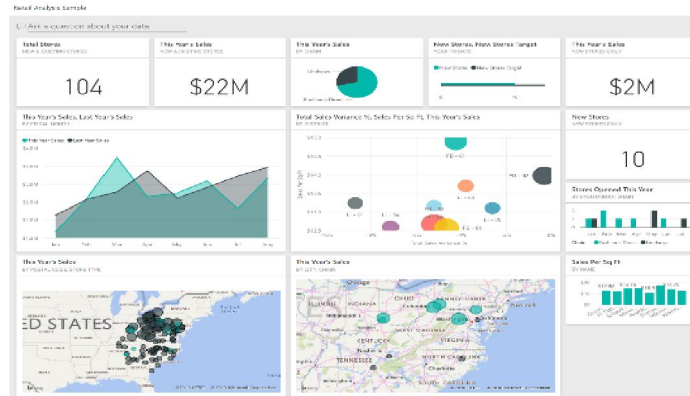


Fig. 4: Item Seasonality Algorithm

5. Affinity Score

Computed by collaborative filtering, association rule mining, and other techniques, affinity scores are essential for evaluating correlations between items in retail data analytics projects based on customer purchase behavior. While association rule mining uses metrics like support, confidence, and lift to identify significant item correlations, collaborative filtering evaluates similarities between people or objects. Personalized recommendations are driven by affinity scores, which point users toward products based on past behavior or attributes that they are likely to find interesting.

IV. RESULT

Our project's output is that a retailer may now use data analytics to understand the value and quantity of products sold in an average order by studying total retail data. Determine which goods sell the best, the worst, and all in between. Determine who your most valuable clients are. A business can maximize profit, perform more effectively, optimize performance, or make more strategically sound judgments with the aid of data analytics

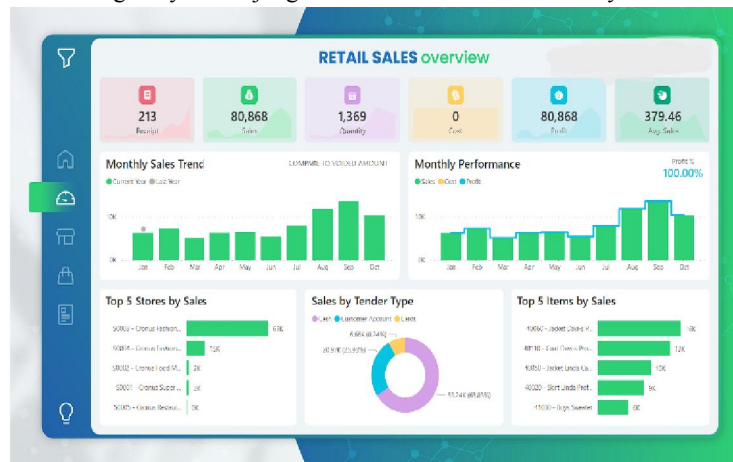


Fig.5: Final output

- A. Benefit
 1. Optimize pricing.
 2. Enhance marketing targeting
 3. A better supply chain
 4. Engines that recommend
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B. Drawback

1. Issues with privacy
2. Safety

V. CONCLUSION

The quality of data available from online purchases, social media talks, and, more recently, location-specific smart phone interactions, has grown into a new entity for digital-based transactions, positioning retailing as the platform for additional data-driven disruption. Organizations that use big data management benefit from increased performance, greater risk management, and the capacity to uncover insights that might otherwise go undiscovered. Retailers can gain a great deal of insight into how their customers use their goods and services, how their operations and supply chain are operating, how to manage their workforce, and how to identify critical risks by adopting a structured, analytics-driven approach. With this knowledge, they can take appropriate action. By declaring on-the-spot discounts on the sales floor based on their past and present shopping habits, they are able to enhance the allure and modify their prices thanks to this research of the move. The retailer can use this data, which is frequently gathered in-store via interactive mobile devices, to better understand the demands of their customers and make more informed judgments about where to put products in the shop. Consequently, retail analysis has a generally favorable impact on the retail industry

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