

CAMPUS CAFETERIA INSIGHTS SALES ANALYSIS

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ABSTRACT

The Campus Cafeteria Insights Sales Analysis system is a data-driven application designed to analyze cafeteria sales transactions and generate useful insights about food-item demand, customer preferences, revenue, and sales performance. Campus cafeterias handle a large number of transactions every day, creating valuable data that can be used to understand purchasing patterns. Analyzing this information manually can be time-consuming and may not provide a complete view of cafeteria performance.

The proposed system collects and processes sales information such as food item, category, quantity sold, price, transaction date, transaction time, payment method, and total sales amount. Data preprocessing techniques are applied to clean the dataset, remove duplicate records, handle missing values, and standardize inconsistent information. The processed data is then used for exploratory and comparative analysis. The system analyzes important sales indicators such as total revenue, total orders, quantity sold, average transaction value, and best-selling food items. It can also analyze sales according to food categories, weekdays, months, time periods, and payment methods. Interactive dashboards, charts, graphs, and KPI cards provide a simple way to understand cafeteria sales performance.

The system can identify peak sales periods and frequently purchased food items. Analysis of sales by time and category can help cafeteria administrators understand demand patterns and plan inventory accordingly. Historical sales trends can also provide useful information for menu planning and resource management.

Overall, the Campus Cafeteria Insights Sales Analysis system provides a centralized platform for converting raw cafeteria transaction data into meaningful insights. It reduces manual reporting and supports data-driven operational decisions. Future enhancements can include demand forecasting, inventory prediction, personalized food recommendations, waste analysis, real-time sales monitoring, and intelligent menu optimization.

I. INTRODUCTION

Campus cafeterias provide food and beverage services to students, faculty, staff, and visitors. Every day, cafeterias generate a significant number of transactions involving different food items, prices, quantities, and payment methods. This transaction data contains useful information that can help cafeteria management understand customer preferences and sales performance.

Traditional cafeteria management often focuses on recording transactions, managing inventory, and collecting payments. Although these activities are important, basic transaction systems may provide limited analytical capabilities. Management may need to manually prepare spreadsheets and reports to determine which food items are popular or which periods generate higher sales.

Sales patterns can change depending on weekdays, academic schedules, examination periods, holidays, and time of day. Certain food items may have higher demand

during breakfast, lunch, or evening hours. Understanding these patterns can help cafeteria operators plan food preparation and inventory more efficiently.

Data analytics provides techniques for analyzing large volumes of transaction data and identifying meaningful relationships. Descriptive analytics can calculate revenue, order volume, average transaction value, and item-wise sales. Visualization techniques can present these metrics through charts and dashboards, making complex sales information easier to understand.

The proposed Campus Cafeteria Insights Sales Analysis system combines data preprocessing, exploratory data analysis, sales analysis, visualization, and reporting. It provides insights into food-item performance, revenue trends, customer purchasing patterns, and peak sales periods. The system can also serve as a foundation for future predictive analytics and intelligent cafeteria-management solutions.

II. LITERATURE SURVEY

Sales analytics has become an important application of data analysis in the food-service and retail industries. Transaction datasets typically contain information about products, quantities, prices, dates, times, and payment methods. Organizations analyze this information to understand product performance, customer demand, and revenue patterns.

Descriptive analytics is commonly used to calculate important sales metrics. Measures such as total revenue, total orders, quantity sold, average transaction value, and product-level sales provide a basic understanding of business performance. These metrics can be grouped by product category, date, time, or location for deeper analysis. Market-basket and purchasing-pattern analysis are also used to understand relationships between products. In a cafeteria environment, transaction data can help identify food items that are frequently purchased together. Such information can support menu planning and promotional strategies while maintaining appropriate operational oversight.

Time-based sales analysis is another important area of research. Sales can vary according to hour, weekday, month, academic calendar, or special events. Identifying peak periods can help food-service operators plan staffing, preparation schedules, and inventory levels more effectively.

Data visualization is widely used for communicating sales insights. Dashboards can display KPIs, sales trends, product rankings, category distributions, and time-based comparisons. Interactive filtering allows users to focus on specific dates, food categories, or items and obtain more detailed information.

Recent sales analytics systems increasingly incorporate Machine Learning for demand forecasting, inventory prediction, recommendation systems, and anomaly detection. Such approaches can extend descriptive analysis into predictive decision support. The proposed Campus Cafeteria Insights Sales Analysis system follows this direction by combining transaction-data processing, sales analytics, visualization, and a foundation for future predictive capabilities.

III. SYSTEM ANALYSIS

The Campus Cafeteria Insights Sales Analysis system is designed to process cafeteria transaction data and generate meaningful sales insights. The system accepts information such as food item, category, quantity, unit price, transaction date, transaction time, payment method, and total amount. The input dataset is first validated to ensure that required fields and supported formats are available. A preprocessing module removes duplicate transactions and handles missing, invalid, or

inconsistent values. Date and time fields are standardized to support daily, weekly, monthly, and time-period analysis. The analytics engine calculates total revenue, total orders, quantity sold, average transaction value, and item-wise sales. Comparative analysis is performed across food categories, payment methods, weekdays, and time periods. The system identifies top-selling and low-selling food items based on historical transaction data. Trend analysis helps identify periods with high and low sales activity. Visualization components present results using charts, graphs, tables, and KPI cards. Users can apply filters to examine specific food categories, dates, items, or transaction periods. The system can generate reports that summarize important cafeteria sales information for authorized administrators.

Existing System

Existing campus cafeteria systems generally focus on order processing, billing, payment collection, and basic transaction recording. These systems are useful for completing daily sales operations and maintaining transaction records. However, basic systems may provide limited analytical capabilities for understanding overall sales performance. Cafeteria administrators may need spreadsheets or separate tools to analyze historical transaction data. Calculating total revenue, average transaction value, and item-wise sales manually can consume significant time. Identifying peak sales hours and high-demand food items may also require additional processing. Traditional systems may provide basic reports but often have limited interactive dashboards and visualization capabilities. Comparing sales across categories, weekdays, and months can become difficult when the dataset grows. Missing or inconsistent transaction data can affect manually generated reports. Sales trends may not be immediately visible when information is presented only as raw transaction tables. Inventory and menu decisions may therefore rely heavily on manual observation. An integrated analytical platform can provide a more efficient approach by combining transaction processing with automated sales analysis and visualization.

Disadvantages of Existing System

- Limited sales analytics capabilities.
- Manual preparation of sales reports.
- Difficult to analyze large transaction datasets.
- Limited visualization and dashboard support.
- Difficult to identify peak sales periods.
- Best-selling items may require manual calculation.
- Limited category-wise sales comparison.
- Historical trends can be difficult to analyze.
- Data cleaning may require manual effort.
- Limited customer purchasing insights.
- Difficult to monitor performance in real time.
- Limited support for predictive sales analysis.

Proposed System

The proposed Campus Cafeteria Insights Sales Analysis system provides a centralized platform for analyzing cafeteria transactions and generating useful business insights. Users can upload historical sales datasets or connect the application to an authorized transaction source. The system validates the data and performs preprocessing to remove duplicates and handle missing, invalid, or inconsistent records. Date and time fields are transformed into useful features such as weekday, month, hour, and sales period. The analytics module calculates KPIs including total revenue, total orders,

quantity sold, average transaction value, and category-wise performance. Users can compare sales across different food categories, individual items, payment methods, and time periods. The system identifies top-selling and low-selling food items using historical transaction data. Time-based analysis highlights peak sales hours and periods with higher demand. Interactive dashboards present the results using KPI cards, charts, graphs, tables, and filters. The reporting module can generate summaries for cafeteria administrators and management teams. Historical sales data can also be used later to build demand forecasting and inventory-prediction models. Overall, the proposed system combines cafeteria transaction analysis and business intelligence in a unified platform.

Advantages of Proposed System

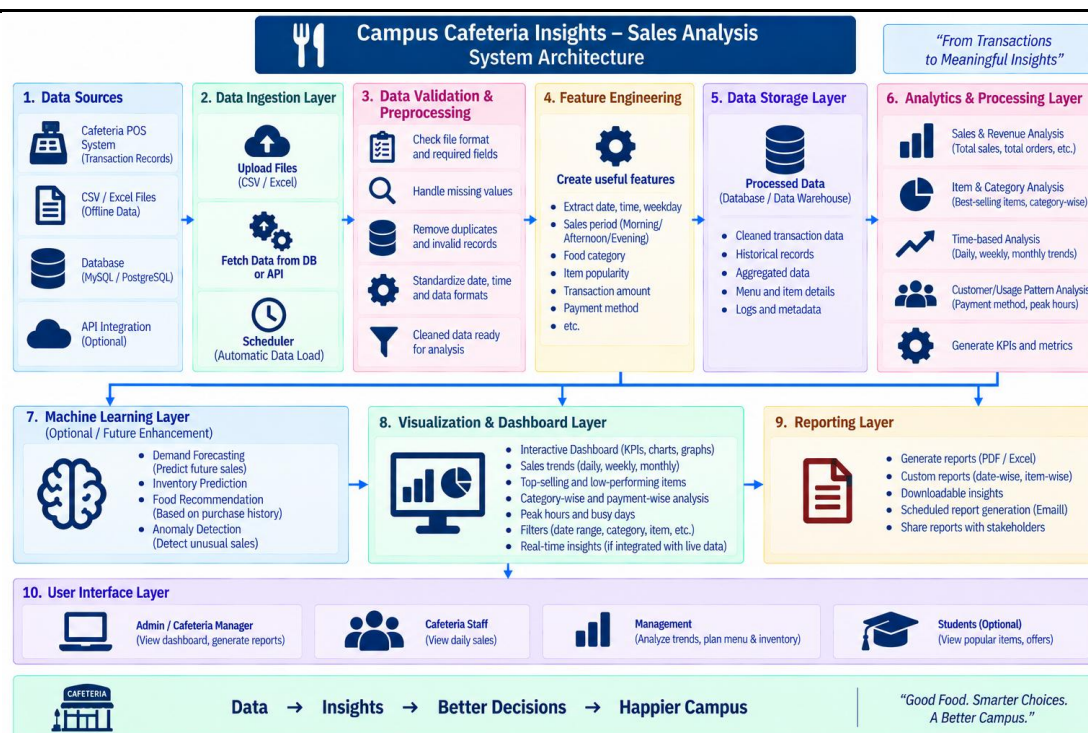
- Provides centralized cafeteria sales analysis.
- Automates sales-data processing.
- Calculates important sales KPIs automatically.
- Identifies best-selling food items.
- Identifies low-performing food items.
- Analyzes sales by food category.
- Identifies peak sales periods.
- Provides interactive dashboards.
- Supports time-based sales analysis.
- Reduces manual report preparation.
- Supports customized filtering and reporting.
- Provides a foundation for future demand forecasting.

IV. METHODOLOGY

The methodology begins with collecting historical cafeteria transaction data from an authorized dataset or sales system. The collected data is validated to check its format, required fields, data types, and completeness. Data preprocessing is performed to remove duplicate transactions and handle missing, invalid, or inconsistent values. Transaction dates and times are standardized to support different types of time-based analysis. Feature engineering is used to create attributes such as weekday, month, hour, sales period, and food category. Exploratory Data Analysis is performed to understand the distribution of sales and identify important patterns. Item-wise and category-wise analysis is conducted to determine popular food products and overall category performance. Revenue analysis is performed to calculate total sales, average transaction value, and revenue trends. Time-based analysis identifies peak hours, high-demand days, and seasonal variations in cafeteria sales. Visualization techniques present results using bar charts, line graphs, pie charts, tables, and interactive dashboards. Users can apply filters to examine specific dates, food categories, payment methods, or individual items. The complete methodology combines data collection, validation, preprocessing, feature engineering, exploratory analysis, sales analysis, visualization, filtering, and reporting.

SYSTEM ARCHITECTURE

The system architecture consists of several interconnected layers that process cafeteria transaction data and deliver analytical insights.



The Data Source Layer provides historical sales records from authorized cafeteria systems, CSV files, Excel files, databases, or APIs. The Data Input Layer receives transaction information from the selected source. The Data Validation Layer checks file format, required attributes, data types, and basic data quality. The Data Preprocessing Layer handles duplicate records, missing values, inconsistent formats, and invalid transactions. The Feature Engineering Layer creates useful attributes such as sales date, weekday, month, hour, sales period, and food category. The Data Storage Layer securely stores cleaned transaction records and historical sales information. The Analytics Processing Layer calculates revenue, order count, quantity sold, average transaction value, and other KPIs. The Sales Analysis Layer examines item-wise, category-wise, payment-wise, and time-based sales patterns. The Trend Analysis Layer identifies peak periods, high-demand items, and changes in sales over time. The Visualization Layer presents results through dashboards, charts, graphs, tables, and KPI cards. The Reporting Layer generates analytical summaries and reports for authorized cafeteria administrators. An optional Machine Learning Layer can later provide demand forecasting, inventory prediction, food recommendations, and other predictive insights.

V. RESULT AND OUTPUT



VI. CONCLUSION

The Campus Cafeteria Insights Sales Analysis system provides a data-driven solution for analyzing cafeteria sales transactions and generating meaningful business insights. The system combines data preprocessing, sales analysis, visualization, and reporting to reduce the effort required for manual data analysis.

The system analyzes important factors such as food items, categories, quantity sold, prices, transaction dates, transaction times, and payment methods. This helps identify best-selling items, low-performing products, popular food categories, and overall revenue patterns.

The dashboard provides important metrics such as total sales, total orders, average order value, and customer-related information. Visualizations such as sales trends, category-wise analysis, time-of-day analysis, weekday analysis, and payment-method distribution make the data easier to understand.

The proposed system can help cafeteria administrators monitor sales performance, identify peak sales periods, understand customer purchasing patterns, and make better decisions related to menu planning and inventory management. Historical transaction data can also provide a foundation for future predictive analysis.

In the future, the system can be enhanced with Machine Learning-based demand forecasting, inventory prediction, food recommendations, waste analysis, real-time sales monitoring, anomaly detection, and automated alerts. These enhancements can make the platform more intelligent, scalable, and useful for efficient campus cafeteria management.

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