Big Data - Case Study: Coffee sales Analysis

1. Introduction

In today's data-driven world, understanding customer preferences and sales trends is essential for business growth. This project focuses on Coffee Sales Analysis using MySQL, Hive, and Sqoop.

The objective is to analyze coffee sales performance, identify top-performing products, and gain insights into consumer buying patterns.

MySQL serves as the foundation for structured data management and querying. Hive, a Hadoop-based tool, enables large-scale data analysis, while Sqoop bridges data transfer between MySQL and Hadoop for scalable computation.

Through this case study, we aim to demonstrate the power of integrated data technologies to perform efficient, insightful, and data-driven decision-making in the retail coffee business.

2. Description of the Dataset

Dataset Name: Coffee_Sales.csv

This dataset records sales transactions for different coffee products across multiple dates and customer types.

Attributes:

- 1. **transaction_id** Unique identifier for each transaction.
- 2. date Date of the coffee sale.
- 3. **time** Time of the transaction.
- 4. day_of_week Day name (e.g., Monday, Tuesday).
- 5. **customer_type** Indicates whether the customer is a member or non-member.
- 6. **coffee_name** Type of coffee sold (e.g., Latte, Espresso, Cappuccino).
- 7. **size** Coffee size (Small, Medium, Large).
- 8. **quantity** Number of cups sold.
- 9. **unit price** Price per cup.
- 10. **money** Total amount of the sale (quantity × unit price).
- 11. payment_mode Payment method used (e.g., Cash, Card, UPI).
- 12. **branch** Store location or branch where the sale took place.

This structured dataset enables a detailed exploration of customer behavior, sales performance by coffee type, and revenue analysis across time periods.

3. Project Scope

The scope of this project is to conduct a comprehensive analysis of coffee sales data to uncover business insights.

Using MySQL, Hive, and Sqoop, we will integrate and process the dataset to understand:

- Daily and weekly sales trends
- Top-selling coffee types and branches
- Customer purchase behavior (members vs. non-members)
- Revenue contribution by size and payment mode

By combining SQL and big data tools, the project aims to build a robust analytical framework to help coffee businesses improve marketing, stock management, and overall sales strategy.

4. Goals

- 1. Sales Trend Analysis: Examine daily, weekly, and monthly sales patterns.
- 2. **Top Product Identification:** Identify best-selling coffee products and their revenue contribution.
- 3. **Customer Insights:** Compare member vs. non-member buying behaviors.
- 4. **Branch Performance:** Evaluate revenue performance across different branches.
- 5. Payment Method Analysis: Understand customer preferences in payment modes.
- 6. Integration: Implement data transfer between MySQL and Hadoop using Sqoop.
- 7. **Scalable Analytics:** Use Hive for efficient querying on large-scale datasets.
- 8. **Visualization:** Create visual dashboards for better insight communication.

5. Tools and Working Environment

1. MySQL

Description:

MySQL is an open-source relational database management system used to store and manage structured data efficiently.

• Working Environment:

In this project, MySQL was used to import the Coffee Sales.csv dataset, create

tables, and perform SQL queries for preliminary data analysis — such as total sales, popular coffee types, and weekday vs weekend trends.

2. Python (for Visualization)

• Description:

Python, with libraries such as **Matplotlib** and **Seaborn**, was used to visualize key insights from the dataset.

• Working Environment:

Visualization scripts were executed in Jupyter Notebook to create bar charts and graphs for:

- Top 3 best-selling coffees
- Most profitable month
- Peak sales hours
- o Day-wise and branch-wise revenue comparison

Performing Analysis on MySQL

Then Verify-

```
ysql> SELECT * FROM Coffe sales LIMIT 10;
transaction_id | date | time
                                                      | month | coffee_name | size | quanti
                                   day
y | unit_price | money | payment_method |
          10 | 0000-00-00 | 00:00:39 | Latte
                                                      | Morning | Fri
                                                                             Mar
        3.00 | 1.00 | 15:50.5
           12 | 0000-00-00 | 00:00:39 | Hot Chocolate
                                                      | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 19:22.5
          12 | 0000-00-00 | 00:00:39 | Hot Chocolate
                                                      | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 20:18.1
          13 | 0000-00-00 | 00:00:29 | Americano
                                                      | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 46:33.0
          13 | 0000-00-00 | 00:00:39 | Latte
                                                      | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 48:14.6
           15 | 0000-00-00 | 00:00:34 | Americano with Milk | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 39:47.7
           16 | 0000-00-00 | 00:00:39 | Hot Chocolate
                                                    | Afternoon | Fri
                                                                             Mar
        3.00 | 1.00 | 19:02.8
          18 | 0000-00-00 | 00:00:34 | Americano with Milk | Night | Fri
                                                                             Mar
        3.00 | 1.00 | 39:03.6
          19 | 0000-00-00 | 00:00:39 | Cocoa
                                                      Night
                                                               Fri
                                                                             Mar
        3.00 | 1.00 | 22:01.8
          19 | 0000-00-00 | 00:00:34 | Americano with Milk | Night
                                                               Fri
                                                                              Mar
        3.00 | 1.00 | 23:15.9
```

Total Sales Revenue-

```
mysql> SELECT SUM(money) AS Total_Sales
-> FROM Coffe_sales;
+-----+
| Total_Sales |
+-----+
| 55257.00 |
+-----+
1 row in set (0.00 sec)
```

Most Popular Coffee-

Average Sale Value-

```
mysql> SELECT AVG(money) AS Average_Sale
    -> FROM Coffe_sales;
+-----+
| Average_Sale |
+-----+
| 15.578517 |
+-----+
1 row in set (0.00 sec)
```

Weekend vs Weekday Sales-

Most Profitable Month-

Most Popular Cup Size-

```
mysql> SELECT size, SUM(quantity) AS Cups_Sold
   -> FROM Coffe_sales
    -> GROUP BY size
    -> ORDER BY Cups_Sold DESC;
 size | Cups_Sold |
      1889
1578
1469
1344
1100
 Mar
 0ct
 Feb
  Sep
  Aug
           1058
 Nov
            1028
 Dec
             930
925
  Jun
 May
  Jul
              896
              770
  Jan
 Apr
              654
12 rows in set (0.01 sec)
```

Peak Hour of Sales-

```
mysql> SELECT HOUR(time) AS Hour, SUM(money) AS Total_Sales
    -> FROM Coffe_sales
    -> GROUP BY HOUR(time)
    -> ORDER BY Total_Sales DESC;
+----+
| Hour | Total_Sales |
+----+
| 0 | 55257.00 |
+----+
1 row in set (0.04 sec)
```

Average Quantity Sold per Transaction-

Day with Highest Total Sales-

Data Visulization

Step 1: Import Libraries

```
[1]: import pandas as pd import matplotlib.pyplot as plt import seaborn as sns

[2]: sns.set(style="whitegrid")

*[3]: i Load the dataset df = pd.read_csv("Coffe_sales.csv")

[4]: df['Date'] = pd.to_datetime(df['Date'], errors='coerce')

print(" Data Loaded Successfully")
 print(df.head())

Data Loaded Successfully Hours cash_type money coffee_name 0 10 card 38.7 Latte Morning Fri Mar
1 12 card 38.7 Hot Chocolate Afternoon Fri Mar
2 12 card 38.7 Hot Chocolate Afternoon Fri Mar
3 13 card 28.9 Americano Afternoon Fri Mar
4 13 card 38.7 Latte Afternoon Fri Mar

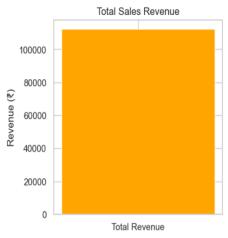
Weekdaysort Monthsort Date Time.1
0 5 3 2024-01-03 15:50.5
1 5 3 2024-01-03 19:22.5
2 5 3 2024-01-03 20:18.1
3 5 3 2024-01-03 48:14.6
```

Total Sales Revenue

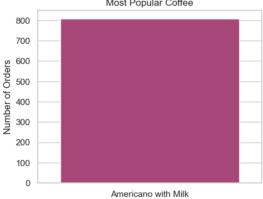
```
# Total Sales Revenue
total_sales = df['money'].sum()
print("  Total Sales Revenue:", round(total_sales, 2))

plt.figure(figsize=(4,4))
plt.bar(['Total Revenue'], [total_sales], color='orange')
plt.title('Total Sales Revenue')
plt.ylabel('Revenue (₹)')
plt.show()
```

å Total Sales Revenue: 112245.58



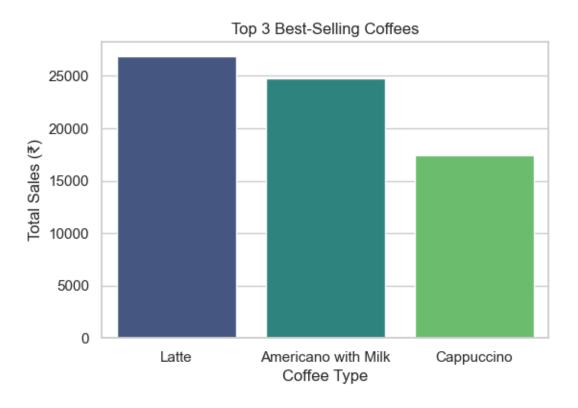
Most Popular Coffee



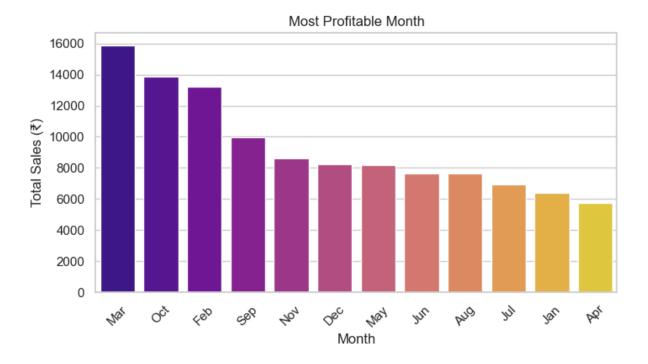
Americano with Milk coffee name

Top 3 Best-Selling Coffee Types

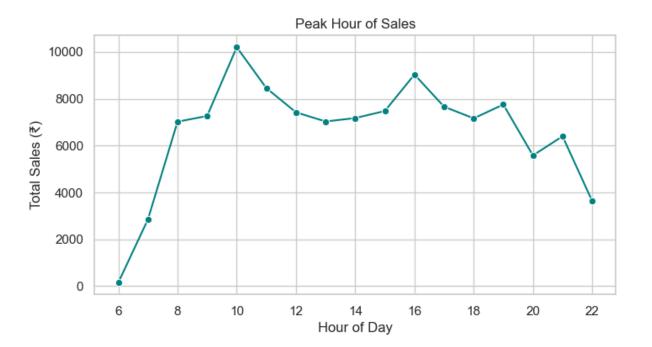
```
[13]: # Top 3 Best-Selling Coffee Types
                                                                                                                              ⑥↑↓去♀▮
      top3 = df.groupby('coffee_name')['money'].sum().sort_values(ascending=False).head(3)
      print("∑ Top 3 Best-Selling Coffees:\n", top3)
      plt.figure(figsize=(6,4))
      sns.barplot(x=top3.index, y=top3.values, palette='viridis')
      plt.title('Top 3 Best-Selling Coffees')
      plt.ylabel('Total Sales (₹)')
      plt.xlabel('Coffee Type')
      plt.show()
      Top 3 Best-Selling Coffees:
      coffee_name
      Latte
                          26875.30
      Americano with Milk 24751.12
      Cappuccino
                          17439.14
      Name: money, dtype: float64
```



```
•[16]: # Most Profitable Month
      month_sales = df.groupby('Month_name')['money'].sum().sort_values(ascending=False)
      plt.figure(figsize=(8,4))
      sns.barplot(x=month_sales.index, y=month_sales.values, palette='plasma')
      plt.title('Most Profitable Month')
      plt.ylabel('Total Sales (₹)')
      plt.xlabel('Month')
      plt.xticks(rotation=45)
      plt.show()
      Monthly Sales:
      Month_name
      Mar 15891.64
      Oct 13891.16
      Feb 13215.48
      Sep 9988.64
      Nov
            8590.54
     Dec
            8237.74
      May 8164.42
      Jun 7617.76
      Aug 7613.84
      Jul
            6915.94
            6398.86
      Jan
            5719.56
      Name: money, dtype: float64
```



```
[18]: hour_sales = df.groupby('Hours')['money'].sum().sort_values(ascending=False)
     plt.figure(figsize=(8,4))
      sns.lineplot(x=hour_sales.index, y=hour_sales.values, marker='o', color='teal')
     plt.title('Peak Hour of Sales')
     plt.xlabel('Hour of Day')
     plt.ylabel('Total Sales (₹)')
     plt.show()
      Hours
      10
          10198.52
      16
            9031.84
      11
            8453.10
            7751.96
      19
            7659.76
      17
      15
            7476.02
      12
            7419.62
            7264.28
      9
      14
            7173.80
      18
            7162.60
            7028.76
      13
            7017.88
      21
            6397.94
            5578.92
      20
      22
            3635.16
      7
            2846.02
      6
            149.40
```



Name: money, dtype: float64

```
daily_sales = df.groupby('Date')['money'].sum().sort_values(ascending=False)
top_day = daily_sales.head(1)
print(" Day with Highest Total Sales:\n", top_day)
    plt.figure(figsize=(10,4))
    sns.lineplot(x=daily_sales.index, y=daily_sales.values, color='red')
    plt.title('Total Sales by Date')
    plt.xlabel('Date')
    plt.ylabel('Total Sales (₹)')
    plt.show()
    print("☑ Visualization Completed Successfully!")
```

Day with Highest Total Sales:

2024-11-10 836.66 Name: money, dtype: float64

