

### **COFFEE SALES PREDICTION USING MACHINE LEARNING**

PRESENTED BY

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### PROBLEM STATEMENT

UNDERSTANDING COFFEE SALES TRENDS IS CRUCIAL FOR OPTIMIZING STOCK REFILLING, PRICING STRATEGIES, AND CUSTOMER SATISFACTION IN VENDING MACHINES. THIS PROJECT AIMS TO PREDICT COFFEE SALES USING **HISTORICAL VENDING MACHINE DATA** TO IMPROVE BUSINESS OPERATIONS.

## PROPOSED SOLUTION

- DATA COLLECTION: TRANSACTION DETAILS FROM VENDING MACHINE SALES.
- **DATA PREPROCESSING:** HANDLING MISSING VALUES, FEATURE ENGINEERING (MONTH, DAY, HOUR).
- MACHINE LEARNING MODEL: LINEAR REGRESSION TO PREDICT COFFEE SALES.
- **DEPLOYMENT STRATEGY:** USING PREDICTIVE INSIGHTS FOR INVENTORY OPTIMIZATION.

## SYSTEM APPROACH

- LIBRARIES USED: PANDAS, NUMPY, MATPLOTLIB, SEABORN, SKLEARN.
- FEATURE ENGINEERING: EXTRACTING
   MONTH, DAY, HOUR, CASH\_TYPE,
   COFFEE\_NAME.
- **MODEL TRAINING:** SPLITTING DATA (80% TRAINING, 20% TESTING).

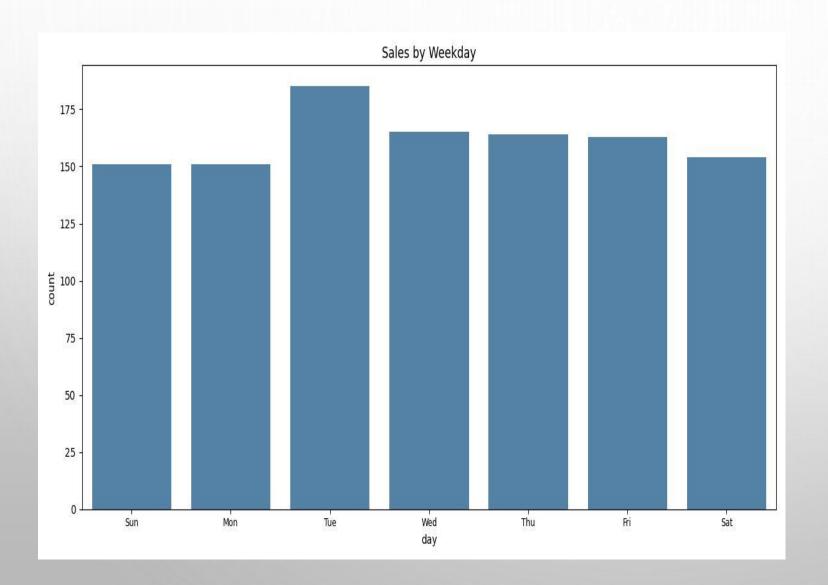
```
coffee_sales.py > ...
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     import seaborn as sns
     import datetime as dt
     import warnings
     import os
     from sklearn.model selection import train test split
     from sklearn.linear model import LinearRegression
      from sklearn.metrics import mean squared error, r2 score
     warnings.filterwarnings('ignore')
     #for dirname, , filenames in os.walk('/kaggle/input'):
        # print(os.path.join(dirname, filename))
     # Load data
     coffee data = pd.read csv('E:\\AICETE & Edunet\\Coffee Sales project details\\coffee sales.csv')
     # Data cleaning
     coffee data['date'] = pd.to datetime(coffee data['date'])
     coffee data['datetime'] = pd.to datetime(coffee data['datetime'])
     coffee data['month'] = coffee data['date'].dt.strftime('%Y-%m')
     coffee data['day'] = coffee data['date'].dt.strftime('%w')
     coffee_data['hour'] = coffee data['datetime'].dt.strftime('%H')
```

## **ALGORITHM & DEPLOYMENT**

- ALGORITHM: LINEAR REGRESSION TRAINED ON EXTRACTED FEATURES.
- MODEL TRAINING: USED HISTORICAL SALES PATTERNS TO LEARN DEMAND TRENDS.
- PREDICTION PROCESS: ESTIMATED
   FUTURE COFFEE SALES BASED ON
   EXTRACTED FEATURES.

```
coffee sales.py > ...
      # Prepare dataset for ML
      features = ['month', 'day', 'hour', 'cash type', 'coffee name']
      # One-hot encode categorical features
      X = pd.get dummies(coffee data[features], drop first=True)
      # Target variable
      y = coffee data['money']
122
      # Split data
      X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
      # Initialize and train model
     model = LinearRegression()
      model.fit(X train, y train)
      # Predict
      y pred = model.predict(X test)
      # Evaluation
      mse = mean squared error(y test, y pred)
      r2 = r2 score(y test, y pred)
      print(f"Model Evaluation:\nMean Squared Error: {mse:.2f}\nR2 Score: {r2:.2f}")
      # Coefficients for interpretation
      coefficients = pd.DataFrame({'Feature': X.columns, 'Coefficient': model.coef })
      print("\nModel Coefficients:")
      print(coefficients.sort values(by='Coefficient', ascending=False))
      # Save model coefficients
      coefficients.to csv('model coefficients.csv', index=False)
```

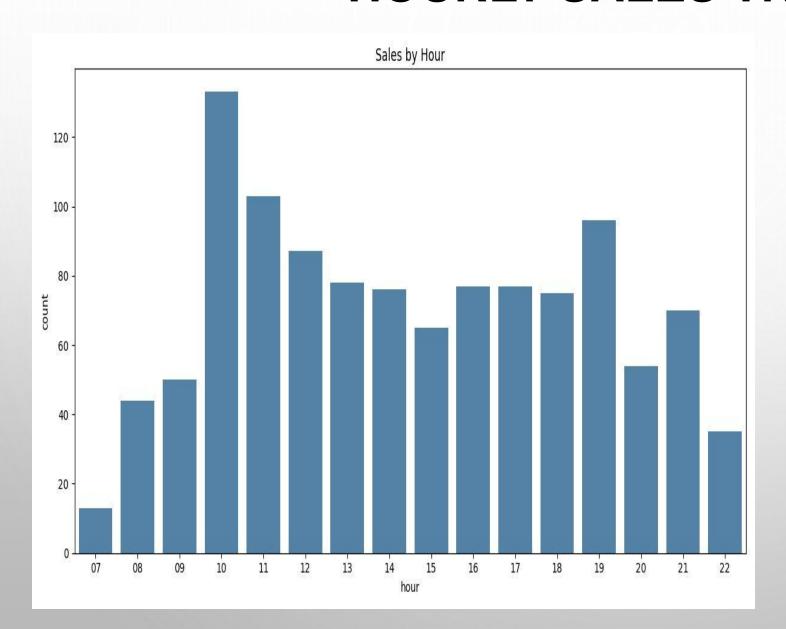
### SALES BY WEEKDAY



#### **KEY INSIGHTS:**

- TUESDAY: HIGHEST SALES.
- MONDAY: LOWEST DEMAND.
- BUSINESS IMPACT: ADJUST
   RESTOCKING SCHEDULE
   BASED ON WEEKDAY TRENDS.

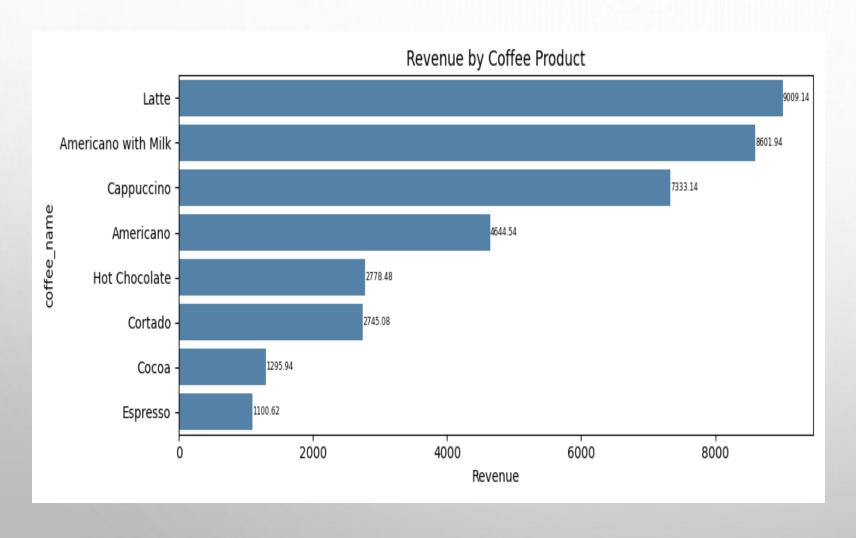
### **HOURLY SALES TRENDS**



#### **KEY FINDINGS:**

- PEAK HOUR: 10:00 AM (HIGHEST DEMAND).
- EVENING PEAKS: 16:00, 17:00, 19:00 PM.
- BUSINESS IMPACT: OPTIMIZE
   STOCK AVAILABILITY DURING PEAK
   SALES TIMES.

### REVENUE BY COFFEE PRODUCT



#### TOP SELLING COFFEES:

- LATTE (₹9009.14), AMERICANO
   WITH MILK (₹8601.94),
   CAPPUCCINO (₹7333.14).
- **LEAST SELLING:** COCOA, ESPRESSO.
- BUSINESS IMPACT: OPTIMIZE INVENTORY BASED ON HIGHEST REVENUE.

### MACHINE LEARNING MODEL EVALUATION

```
PS E:\AICETE & Edunet> & C:/Users/hp/AppData/Local/Programs/Python/Python313/python.exe "e:/AICETE & Edunet/coffee sales.py"
Model Evaluation:
Mean Squared Error: 0.79
R2 Score: 0.96
Model Coefficients:
                            Feature Coefficient
28
                  coffee name Cocoa
                                        9.827475
27
             coffee name Cappuccino
                                        9.812987
32
                  coffee name Latte
                                        9.781796
31
          coffee name Hot Chocolate
                                        9.753614
26
    coffee name Americano with Milk
                                        4.935424
25
                     cash type cash
                                        1.323496
18
                            hour 16
                                        0.570958
24
                            hour 22
                                        0.455914
20
                            hour 18
                                        0.347354
12
                            hour 10
                                        0.346046
14
                            hour 12
                                        0.290758
                            hour 13
                                        0.288361
8
                              day 5
                                        0.274694
                              day 3
                                        0.199089
21
                            hour 19
                                        0.167856
19
                            hour 17
                                        0.164874
17
                            hour 15
                                        0.158001
16
                            hour 14
                                        0.138231
                              day 2
                                        0.095135
                              day 4
                                        0.089492
                            hour 21
                                        0.074238
                            hour 11
                                        0.062855
                              day 6
                                        0.053696
22
                            hour 20
                                        0.053190
4
                              day 1
                                        0.045242
10
                            hour 08
                                        -0.028731
29
                coffee name Cortado
                                        -0.090195
                            hour 09
                                       -0.146980
                      month 2024-04
                                        -0.317911
                      month 2024-06
                                        -0.893711
                      month 2024-05
                                        -0.924665
30
               coffee name Espresso
                                        -4.719472
                      month 2024-07
                                       -5.275066
PS E:\AICETE & Edunet>
```

#### METRICS:

- MEAN SQUARED ERROR (MSE): 0.79 (LOWER IS BETTER).
- > R<sup>2</sup> SCORE: **0.96** (HIGH ACCURACY).

#### TOP FEATURES IMPACTING SALES:

- COFFEE TYPES WITH THE HIGHEST DEMAND: COCOA, CAPPUCCINO, LATTE.
- > PEAK HOURS INFLUENCING DEMAND.
- CASH TRANSACTIONS SLIGHTLY AFFECTING SALES.

### **KEY FINDINGS & BUSINESS APPLICATIONS**

- OPTIMIZING STOCK FOR HIGH-DEMAND COFFEES.
- ADJUSTING RESTOCKING SCHEDULES BASED ON WEEKDAY DEMAND.
- LEVERAGING MACHINE LEARNING FOR INVENTORY PREDICTIONS.

## CONCLUSION

#### . PROJECT SUMMARY:

- > MACHINE LEARNING SUCCESSFULLY PREDICTS COFFEE SALES.
- > HELPS OPTIMIZE VENDING MACHINE INVENTORY.

#### . FUTURE SCOPE:

- > ADVANCED FORECASTING MODELS (NEURAL NETWORKS).
- > CUSTOMER SEGMENTATION FOR PERSONALIZED RECOMMENDATIONS.

## REFERENCES

- RESEARCH PAPERS/ARTICLES ON MACHINE LEARNING & SALES PREDICTION.
- DATA SOURCE: GITHUB



# Thank you