# JOHNNY & JUGNU

DATA SCIENCE PROJECT

Waleed, Mustafa, Shanza, Amna, Hafsa.



#### <u>Unveiling JJ's Success: Local Flavors, Global Impact</u>

- Overview: JJ's Menu Diversity & Consumer Appeal
- Dataset: October & November 2022 Insights

- Goal: Improving Operational Efficiency
- Opportunities Identified:

Sales Boost & Brand Development

- Recommendations: Service Enhancement, Product Alignment with Trends
- **Economic Contribution:** Fostering Local Brands for Growth
- Vision: Strengthening the Economy via Local Brand Support



# THE GODIE

### RESEARCH QUESTIONS

- 1) What is the most likely delivery/queuing time for a given order, and how can predictive modeling provide more accurate delivery and queuing time?
- 2) To what extent does the mode of payment influence the value of customer orders, and is there a correlation between specific payment methods and order values? Additionally, should JJ consider offering discounts for particular payment modes to enhance revenue generation strategically?
- a) How do promotional campaigns on Instagram affect customer behaviour, and are their specific promotions that lead to a significant increase in orders or revenue?
- 3) What are the most common reasons for customer complaints or service issues and what measures can be implemented to address these concerns and enhance customer retention?

### DATA CLEANING:

#### **Ensuring Data Accuracy:**

Chronologically Ordered the data then combined the data frames into one combined data frame and separate ones for each branch

#### Adding Our Own Columns:

Total Order Time, Processed\_Items, Hour were some of the columns we added

#### **Making Data Consistent:**

We had separate entires titled "generic\_creditcard" and "Credit Card" in "Payment method" and we made that consistent.

#### Daily Sales Totals Extraction:

Creation of 'Date' Column for Sales Analysis

Data Types of Columns:	
Restaurant name	string
Order ID	string
Platform Key	string
Restaurant ID	string
Order status	string
Created at	datetime64[ns]
Accepted at	string
Cancelled at	string
Cancellation reason	string
Cancellation owner	string
Rejected at	float64
Rejection reason	float64
Dispatched at	string
Delivered at	string
Address	float64
Transport type	string
Payment type	string
Payment method	string
Food value	int64
Packaging charges	int64
Discount	int64
Voucher	int64
Minimum order value fee	int64
Delivery Fee	int64
Items	string
Preorder	string
Billable Status	string

- Four JJ Branches in Lahore
- Elements Covered: Restaurant Info, Order Details, Location, Payments, Financials, Order Contents

#### **Cleaning Highlights:**

- Data Unification: Combining Diverse Branch Data for Comprehensive Analysis
- **Temporal Analysis:** Enabling Sales Pattern Examination via Timestamp Conversion
- **Data Transformation:** Conversion of 'Created at' Timestamps to Datetime Format

#### **Importance**:

- Cross-Location Analysis Preparation
- Understanding Sales Patterns over Time

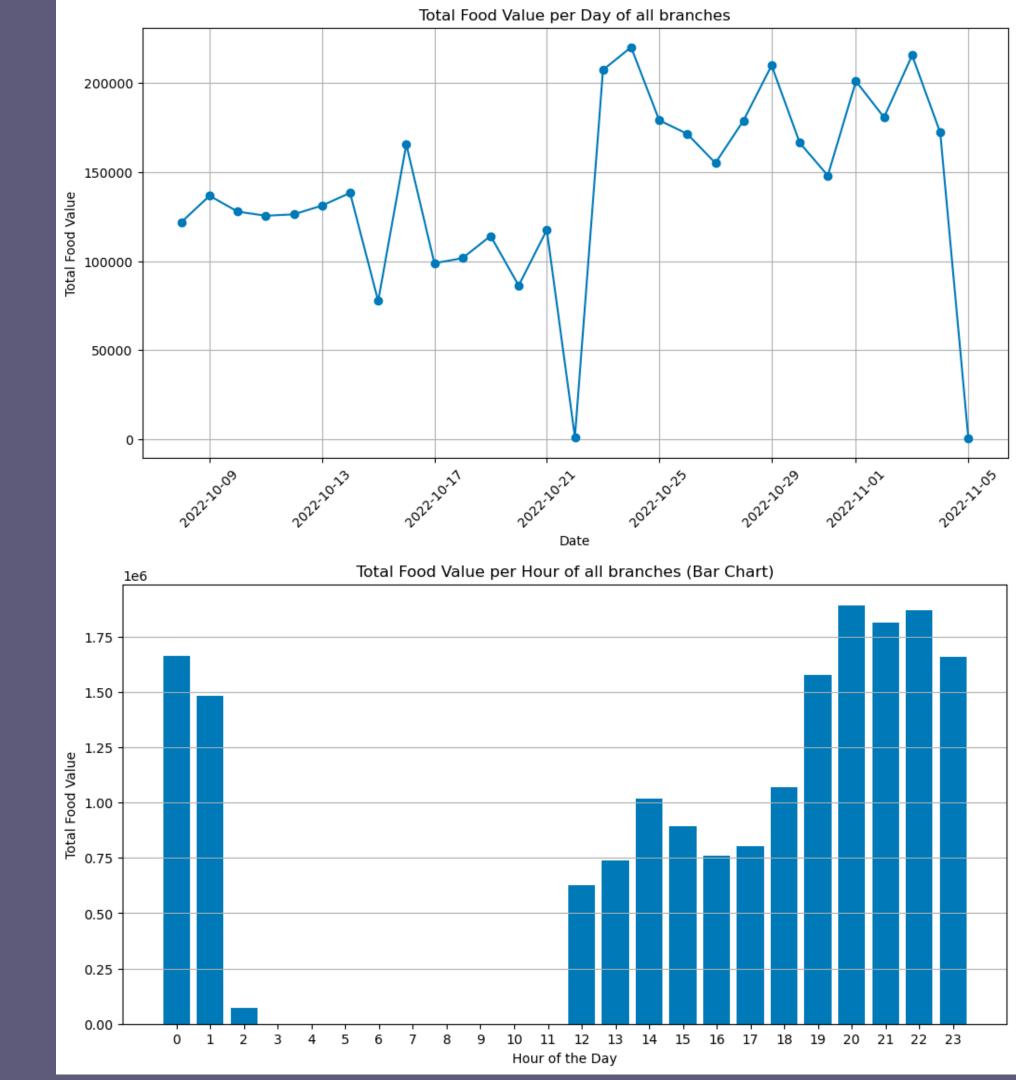
# DATA SET OVERVIEW



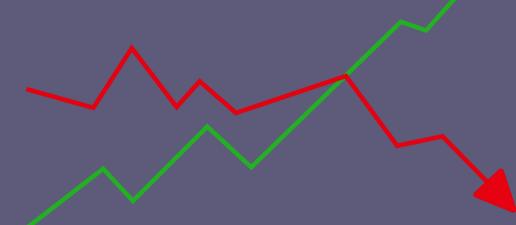
```
Restaurant name Order to Platform Key Restaurant to
Johnny & Jugnu - Johan Town p128-88ff
    ny & Jughu - Johan Town p120-0153
 Delivered 1021-10-08 19:09 1022-10-08 19:09
 Delivered 2022-10-13 16:31 2022-10-13 16:31
 Delivered 2022-10-19 00:59 2022-10-19 01:00
 Delivered 1822-18-18 14:59 1822-18-18 15:88
 peliwered 2022-10-21 15:51 2022-10-21 15:51
                                              Payment method
                                                         600
                                                         600
rood value Packaging charges Discount Voucher Minimum order value fee
     3768
      678
       120 1 SWOL MEAF [1 CHIPOTLE, 1 ATOMIC, 1 NOT REQUI...
       120 1 CHICKEN FILLET [1 GARLIC, 1 GREEK, 1 NOT REQ...
       120 1 NUGS MRAP [1 ATOMIC, 1 GREEK, 1 NOT REQUIRED...
       120 1 TÜRTILLA MRAP [1 GARLIC, 1 GREEK, 1 NOT REQU....
```

#### **Key Findings:**

- A. Exploratory Data Analysis
  - 1. Order Volume Trends
  - 2. Branch Performance
  - 3. Payment methods



### TOTAL FOOD VALUE ANALYSIS BY DAY OF THE WEEK



**All Branches:** 

Mean: PKR 2,562,310

**Std Dev: PKR 221,703** 

Min: PKR 2,257,570 | Max: PKR 2,926,830

25th: PKR 2,437,510 | 50th: PKR 2,511,550 | 75th: PKR

2,682,600

**Phase 4 Branch:** 

Mean: PKR 822,015 | Std Dev: PKR 105,010

Min: PKR 621,680 | Max: PKR 940,330

Phase 6 Branch:

Mean: PKR 456,313 | Std Dev: PKR 90,932

Min: PKR 324,990 | Max: PKR 568,430

**Bahria Town Branch:** 

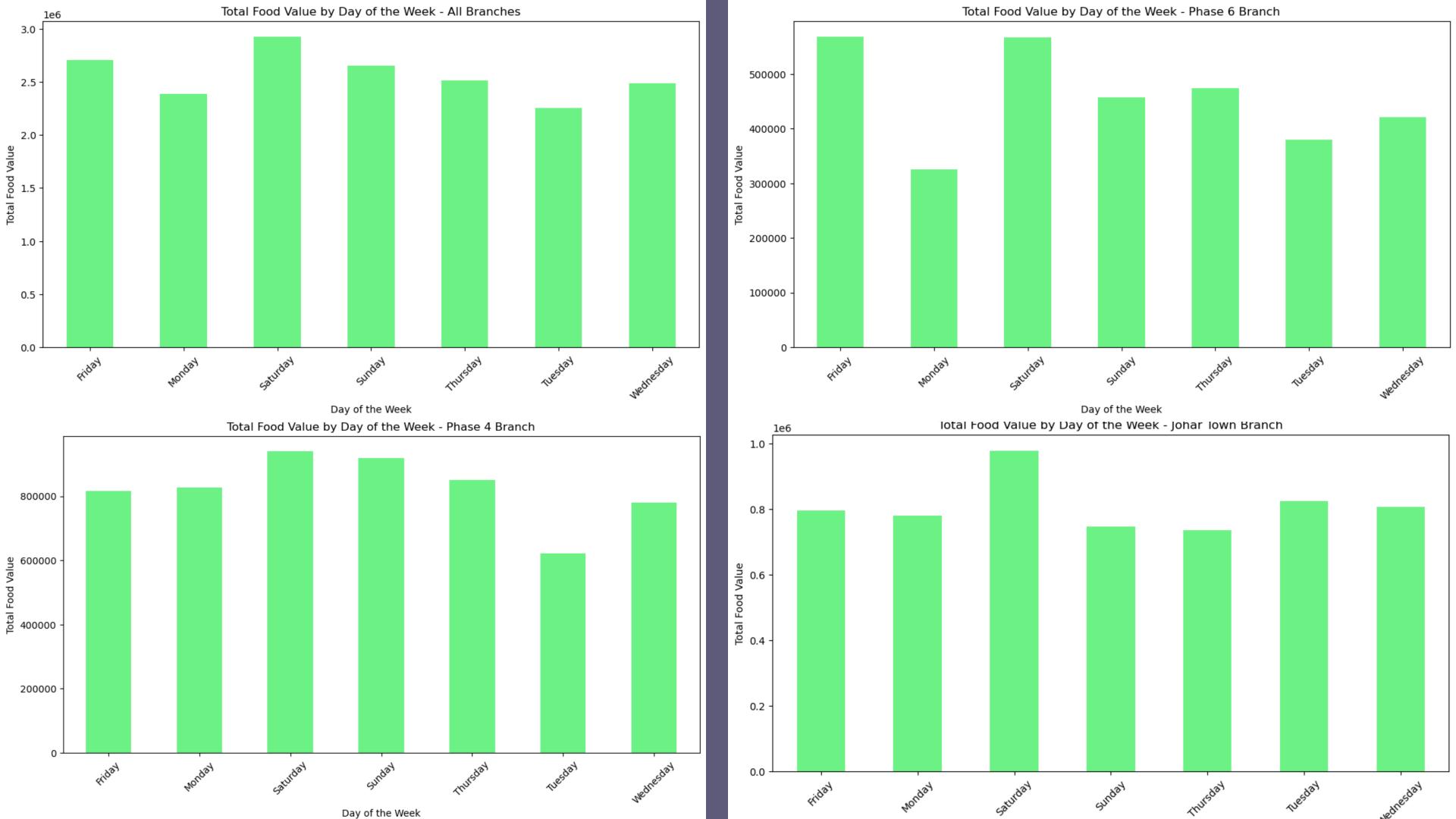
Mean: PKR 474,566 | Std Dev: PKR 41,139

Min: PKR 432,600 | Max: PKR 532,500

**Johar Town Branch:** 

Mean: PKR 809,416 | Std Dev: PKR 80,687

Min: PKR 735,010 | Max: PKR 977,880



# ORDER TIME ANALYSIS

ORDERS ARE SLOWER TILL 2AM, QUICKER THROUGH THE DAY, SLOWER AGAIN AT NIGHT TIME. (NO ORDERS AFTER 2AM TILL 12PM)



Slowest: Phase 4 branch
Overall Average Time: 36.4 minutes
Peak order times extend into later
evening, around 9 PM

Fastest: Phase 6 branch
Overall Average Time: 29.4 minutes
Consistent trend with a slight increase
during the evening

### PAYMENT METHODS

CASH ON DELIVERY

02 CREDIT CARD

03 BALANCE

03 invoice

COD 9768 Orders

60.5% of Orders

BALANCE 158 Orders

0.97% of Orders

&

**VALUES** 

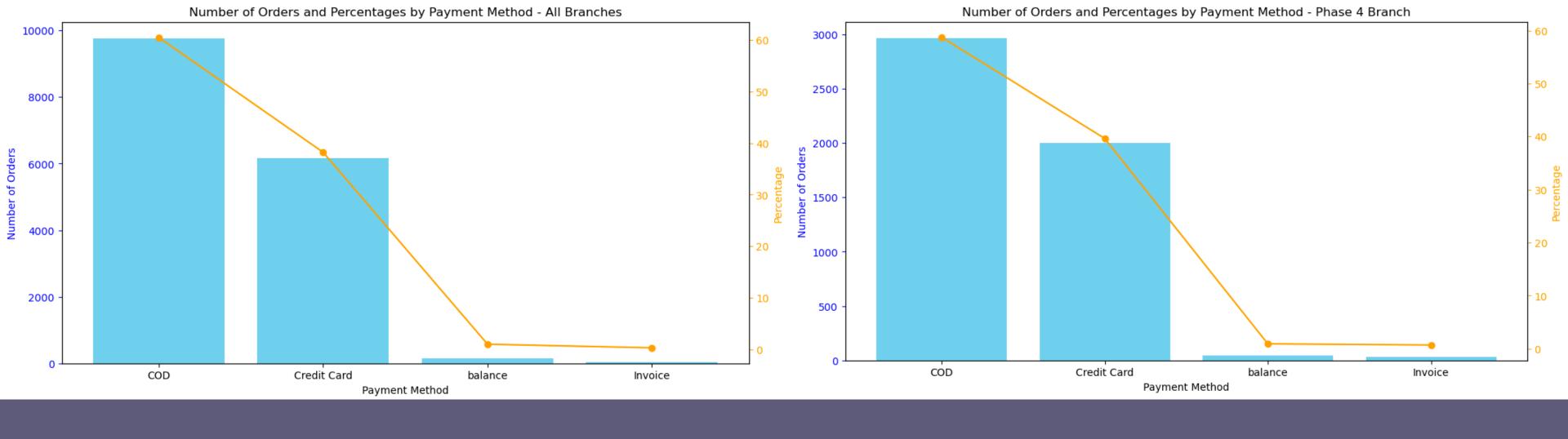
PERCENTAGES

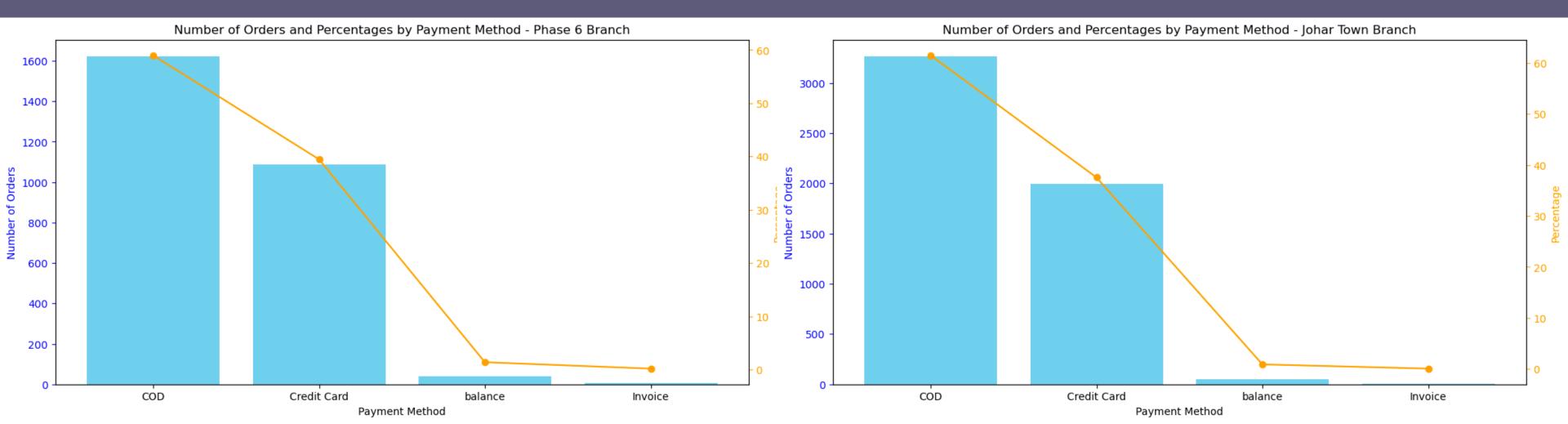
CREDIT CARD 6170 Orders

38.2% of Orders

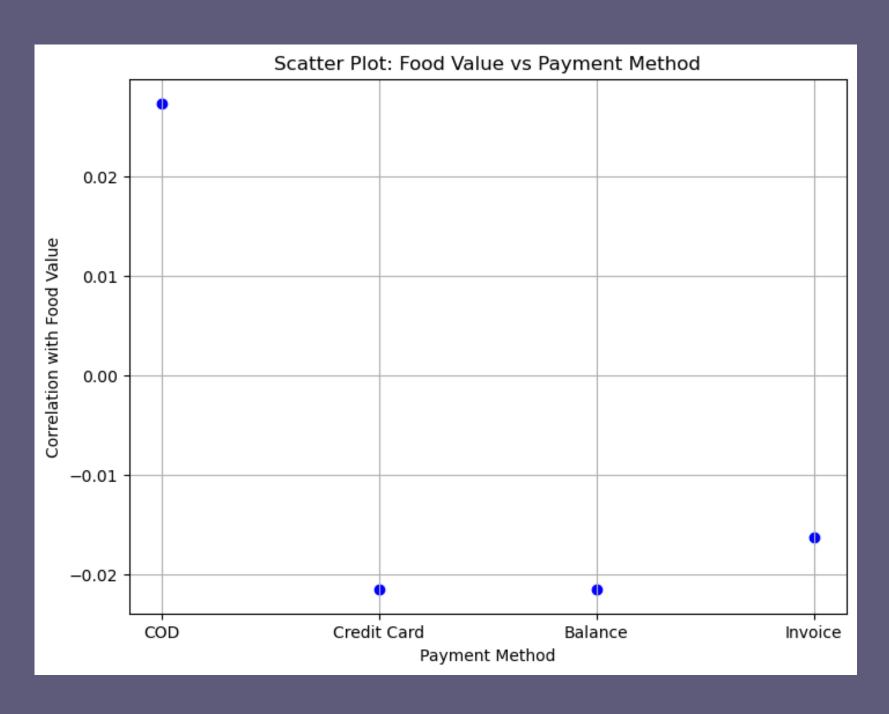
INVOICE 45 Orders

0.27% of Orders





### PAYMENT METHODS VS FOOD VALUE



- COD (Cash on Delivery):
  - Positive correlation (0.0274): Indicates a slight positive relationship with food value.
  - On average, higher food values are associated with COD payments.
- Credit Card:
  - Negative correlation (-0.0214): Weak negative relationship with food value.
  - Suggests a minor tendency for lower food values with Credit Card payments.
- Balance:
  - Negative correlation (-0.0214): Similar weak negative relationship as Credit Card.
  - Implies a slight inclination for lower food values with Balance payments.
- Invoice:
  - Negative correlation (-0.0162): Weakest negative relationship among all methods.
  - Indicates a minimal tendency for lower food values with Invoice payments.

## CAUSAL INFERENCES FROM RESTAURANT DATA



#### **EFFECT OF FOOD VALUE ON TOTAL ORDER TIME:**

- METHODOLOGY: REGRESSION ANALYSIS CONTROLLING FOR HOUR OF THE DAY AND DAY OF THE WEEK.
- FINDINGS:
  - SIGNIFICANT POSITIVE CORRELATION BETWEEN FOOD VALUE AND TOTAL ORDER TIME.
  - HIGHER FOOD VALUE IMPLIES LONGER PREPARATION AND DELIVERY TIMES.
- BUSINESS IMPLICATION:
  - OPTIMIZE ORDER PROCESSING AND DELIVERY LOGISTICS BASED ON FOOD VALUE.
  - ADJUST MENU PRICING AND COMPOSITION FOR BALANCED DELIVERY TIMES.
  - IMPLEMENT DYNAMIC DELIVERY TIME ESTIMATES FOR IMPROVED CUSTOMER SATISFACTION.

#### HOUR OF THE DAY AND AVERAGE FOOD VALUE:

- METHODOLOGY: REGRESSION MODEL CONTROLLING FOR DAY OF THE WEEK AND SPECIAL PROMOTIONS.
- FINDINGS:
  - SIGNIFICANT IMPACT OF CERTAIN HOURS (LATE NIGHT, EARLY MORNING) ON AVERAGE FOOD ORDER VALUE.
  - INDICATES VARIATIONS IN CUSTOMER SPENDING HABITS DURING SPECIFIC TIME PERIODS.
- BUSINESS IMPLICATION:
  - INFORM TARGETED MARKETING AND MENU PLANNING FOR SPECIFIC HOURS.
  - TIME PROMOTIONAL ACTIVITIES TO COINCIDE WITH PERIODS OF HIGHER SPENDING.
  - AID IN INVENTORY MANAGEMENT AND STAFFING FOR DIFFERENT CUSTOMER DEMANDS AT VARIOUS TIMES.

#### Restaurant Data Dynamics

#### 1. Food Value and Delivery Time:

#### Hypotheses:

HO: No correlation between food value and total order time.

H1: Correlation exists between food value and total order time.

#### Analysis and Results:

Pearson correlation analysis: Significant correlation (Correlation Coefficient: 0.1797, p-value: 1.212e-114). Higher food value implies longer delivery times.

#### • Implications:

Emphasizes the role of menu pricing and operational efficiency for larger orders



#### Food Value and Delivery Time:

HO: There is no correlation between the food value of an order and the total order time.

H1: There is a correlation between the food value of an order and the total order time.

#### Peak Hours Effect:

HO: The hour of the day does not affect the average food value of orders.

H1: The hour of the day affects the average food value of orders.

# STATISTICAL TESTING

#### 2. Peak Hours Effect on Average Food Value:

#### Hypotheses:

HO: Hour of the day does not affect average food value.

H1: Hour of the day affects average food value.

#### Analysis and Implications:

**ANOVA test:** Significant impact of hour of the day (F-value: 11.7438, p-value: 1.312e-27).

Customer spending patterns vary at different times, providing opportunities for targeted strategies.

# STATISTICAL TESTING



Pearson Correlation between Food Value and Total Order Time

Correlation Coefficient: 0.17975084965052468

p-value: 1.2126208411216282e-114

Hypothesis 2 is significant: There is a correlation between the food value of an order and the total order time.

ANOVA Test for Effect of Hour of the Day on Average Food Value

sum\_sq df F PR(>F)

C(Hour) 7.707928e+07 14.0 11.743811 1.312644e-2

Residual 7.560094e+09 16126.0 NaN NaN

Hypothesis 3 is significant: The hour of the day affects the average food value of orders.

### ML MODEL: GRADIENT BOOSTING REGRESSOR

#### Objective: predicting total order delivery times

- 1. <u>Dataset preparation and features selection</u> such as food value, discount, and the number of orders in the queue. We employ techniques like imputation to replace missing values.
- 2. Model Selection: Gradient Boosting Regressor
- 3. Loss Function: We designed a custom MSE, which is a specialized version of the Mean Squared Error (MSE) in which predictions within a specified tolerance range are considered accurate.
- 4. Model Performance: 28% Accuracy with an MSE of 7.4.

#### **Low Accuracy:**

The low accuracy of our model could be attributed to several factors:

- complexity of the underlying patterns in the data
- insufficient feature representation
- The chosen features, while insightful, may not comprehensively capture the diverse factors influencing delivery times.
- potential missing features pose a gap in our research, possibly leading to low accuracy.

Our exploration into total order delivery time prediction uncovers not only the strengths of our model but also the inherent challenges in the dataset.

### LIMITATIONS —



### FUTURE DIRECTIONS

- A. REAL-TIME DATA ANALYSIS
- **B. INTEGRATION OF CUSTOMER FEEDBACK DATA**
- C. CONTINUOUS LEVERAGING OF STATISTICAL ANALYSES FOR
- **STRATEGIC PLANNING**