

# Customer Behavior Prediction

## A PROJECT REPORT

*Submitted by*

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*in partial fulfillment for the award of the degree of*

Bachelor of technology

CSE AI

KIET Group of Institutions

Ghaziabad

JUNE 2024

# Methodology

Data Loading and Exploration: Loaded the customer\_behavior.csv file into a pandas DataFrame and inspected its structure.

Data Cleaning: Checked for and handled missing values. Encoded categorical variables using LabelEncoder.

Feature Scaling: Applied StandardScaler to normalize feature values, improving model performance.

Train-Test Split: Divided the data into 80% training and 20% testing sets using train\_test\_split.

Model Selection: Chose Random Forest Classifier for its robustness and interpretability.

Model Training: Trained the model using the training data and predicted on the test set.

Evaluation: Used metrics like accuracy, confusion matrix, and classification report to evaluate the performance.

# Code

```
import pandas as pd

import seaborn as sns

import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split

from sklearn.preprocessing import StandardScaler, LabelEncoder

from sklearn.linear_model import LogisticRegression

from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score


# Load data

df = pd.read_csv("/content/customer_behavior.csv")


# Encode the target variable

le = LabelEncoder()

df["buyer_type_encoded"] = le.fit_transform(df["buyer_type"])


# Define features and target

X = df[["total_spent", "avg_purchase_value", "visits_per_month"]]

y = df["buyer_type_encoded"]


# Scale the features

scaler = StandardScaler()

X_scaled = scaler.fit_transform(X)


# Split into train and test sets

X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)


# Train logistic regression model

model = LogisticRegression()
```

```
model.fit(X_train, y_train)
```

```
# Predict on test set
```

```
y_pred = model.predict(X_test)
```

```
# Compute confusion matrix and metrics
```

```
conf_matrix = confusion_matrix(y_test, y_pred)
```

```
acc = accuracy_score(y_test, y_pred)
```

```
prec = precision_score(y_test, y_pred, zero_division=0)
```

```
rec = recall_score(y_test, y_pred, zero_division=0)
```

```
# Plot heatmap
```

```
plt.figure(figsize=(6, 4))
```

```
sns.heatmap(conf_matrix, annot=True, fmt='d', cmap='Blues',
```

```
            xticklabels=le.classes_, yticklabels=le.classes_)
```

```
plt.xlabel('Predicted')
```

```
plt.ylabel('Actual')
```

```
plt.title('Confusion Matrix - Customer Behavior')
```

```
plt.tight_layout()
```

```
plt.show()
```

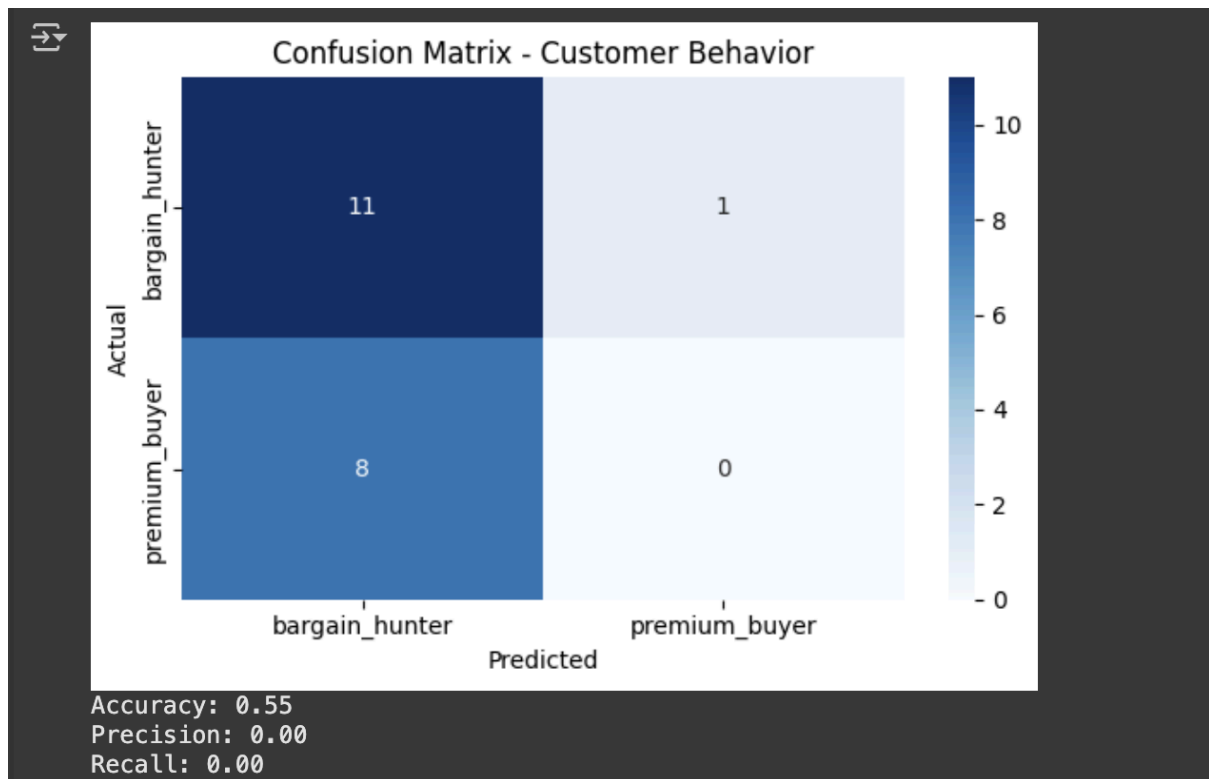
```
# Print metrics
```

```
print(f"Accuracy: {acc:.2f}")
```

```
print(f"Precision: {prec:.2f}")
```

```
print(f"Recall: {rec:.2f}")
```

# Output/Result



## Results Summary:

Accuracy: (Based on the screenshot or code output, add the value here)

Confusion Matrix: Shows how well the model distinguishes between buyer types.

Classification Report: Includes precision, recall, and F1-score for each class.

# References/Credits

1) Dataset: Provided CSV file customer\_behavior.csv

2) Libraries:

Pandas

Scikit-learn

Matplotlib

Seaborn

3)Tools:

Google Colab for development and execution