

Customer Purchase Prediction using Decision Tree Classifier

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Objective:

To build a machine learning model that predicts whether a customer will subscribe to a term deposit (purchase a financial product), based on their demographic and behavioral data.

Technologies Used:

Python
Pandas, NumPy – data preprocessing
Scikit-learn – model building and evaluation
Matplotlib, Seaborn – visualization

Data Used:

Source:UCI Machine Learning Repository

Download Link
(Direct):<https://github.com/Prodigy-InfoTech/data-science-datasets/tree/main/Task%203>



```
In [ ]: import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.metrics import accuracy_score, classification_report
        from sklearn.preprocessing import LabelEncoder
```

```
In [ ]: df = pd.read_csv('Customer-Churn-Records.csv')
```

```
In [ ]: df.head(10)
```

```
Out[ ]:
```

| | RowNumber | CustomerId | Surname | CreditScore | Geography | Gender | Age | T |
|---|-----------|------------|----------|-------------|-----------|--------|-----|---|
| 0 | 1 | 15634602 | Hargrave | 619 | France | Female | 42 | |
| 1 | 2 | 15647311 | Hill | 608 | Spain | Female | 41 | |
| 2 | 3 | 15619304 | Onio | 502 | France | Female | 42 | |
| 3 | 4 | 15701354 | Boni | 699 | France | Female | 39 | |
| 4 | 5 | 15737888 | Mitchell | 850 | Spain | Female | 43 | |
| 5 | 6 | 15574012 | Chu | 645 | Spain | Male | 44 | |
| 6 | 7 | 15592531 | Bartlett | 822 | France | Male | 50 | |
| 7 | 8 | 15656148 | Obinna | 376 | Germany | Female | 29 | |
| 8 | 9 | 15792365 | He | 501 | France | Male | 44 | |
| 9 | 10 | 15592389 | H? | 684 | France | Male | 27 | |

```
In [ ]: data = df.drop(['RowNumber', 'Surname', 'Geography', 'Gender', 'Card Type'], axis=1)
        data.head()
```

```
Out[ ]:
```

| | CustomerId | CreditScore | Age | Tenure | Balance | NumOfProducts | HasCrCard |
|---|------------|-------------|-----|--------|-----------|---------------|-----------|
| 0 | 15634602 | 619 | 42 | 2 | 0.00 | 1 | 1 |
| 1 | 15647311 | 608 | 41 | 1 | 83807.86 | 1 | 0 |
| 2 | 15619304 | 502 | 42 | 8 | 159660.80 | 3 | 1 |
| 3 | 15701354 | 699 | 39 | 1 | 0.00 | 2 | 0 |
| 4 | 15737888 | 850 | 43 | 2 | 125510.82 | 1 | 1 |

```
In [ ]: data.duplicated().sum()
```

```
Out[ ]: 0
```

```
In [ ]: data.isnull().sum()
```

```
Out[ ]: CustomerId      0
        CreditScore    0
        Age            0
        Tenure         0
        Balance        0
        NumOfProducts  0
        HasCrCard      0
        IsActiveMember 0
        EstimatedSalary 0
        Exited         0
        Complain       0
        Satisfaction Score 0
        Point Earned   0
        dtype: int64
```

```
In [ ]: data.shape
```

```
Out[ ]: (10000, 13)
```

```
In [ ]: threshold_balance = data["Balance"].mean()
        threshold_active = data["IsActiveMember"].mean()

def create_purchase_label(row):
    if row['IsActiveMember'] > threshold_active and row['Balance'] > threshold_balance:
        return 1
    else:
        return 0
```

```
In [ ]: data['PurchaseLabel'] = data.apply(create_purchase_label, axis=1)
```

```
In [ ]: print(data[['Balance', 'IsActiveMember', 'PurchaseLabel']])
```

| | Balance | IsActiveMember | PurchaseLabel |
|------|-----------|----------------|---------------|
| 0 | 0.00 | 1 | 0 |
| 1 | 83807.86 | 1 | 1 |
| 2 | 159660.80 | 0 | 0 |
| 3 | 0.00 | 0 | 0 |
| 4 | 125510.82 | 1 | 1 |
| ... | ... | ... | ... |
| 9995 | 0.00 | 0 | 0 |
| 9996 | 57369.61 | 1 | 0 |
| 9997 | 0.00 | 1 | 0 |
| 9998 | 75075.31 | 0 | 0 |
| 9999 | 130142.79 | 0 | 0 |

```
[10000 rows x 3 columns]
```

```
In [ ]: (data['PurchaseLabel']).sum()
```

```
Out[ ]: 3016
```

```
In [ ]: y = data['PurchaseLabel']
x = data.drop(['PurchaseLabel'], axis=1)
```

```
In [ ]: x.head()
```

```
Out[ ]:
```

| | CustomerId | CreditScore | Age | Tenure | Balance | NumOfProducts | HasCrCard |
|---|------------|-------------|-----|--------|-----------|---------------|-----------|
| 0 | 15634602 | 619 | 42 | 2 | 0.00 | 1 | 1 |
| 1 | 15647311 | 608 | 41 | 1 | 83807.86 | 1 | 0 |
| 2 | 15619304 | 502 | 42 | 8 | 159660.80 | 3 | 1 |
| 3 | 15701354 | 699 | 39 | 1 | 0.00 | 2 | 0 |
| 4 | 15737888 | 850 | 43 | 2 | 125510.82 | 1 | 1 |

```
In [ ]: y.head()
```

```
Out[ ]:
```

| | |
|---|---|
| 0 | 0 |
| 1 | 1 |
| 2 | 0 |
| 3 | 0 |
| 4 | 1 |

Name: PurchaseLabel, dtype: int64

```
In [ ]: X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.35, random_state=42)
```

```
In [ ]: clf = DecisionTreeClassifier(random_state=42)
```

```
In [ ]: clf.fit(X_train, y_train)
```

```
Out[ ]:
```

▼ DecisionTreeClassifier

DecisionTreeClassifier(random_state=42)

```
In [ ]: y_pred = clf.predict(X_test)
```

```
In [ ]: accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
```

Accuracy: 0.9997142857142857

```
In [ ]: print("Classification Report:\n", classification_report(y_test, y_pred))
```

Classification Report:

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 1.00 | 1.00 | 1.00 | 2426 |
| 1 | 1.00 | 1.00 | 1.00 | 1074 |
| accuracy | | | 1.00 | 3500 |
| macro avg | 1.00 | 1.00 | 1.00 | 3500 |
| weighted avg | 1.00 | 1.00 | 1.00 | 3500 |

```
In [ ]: from sklearn.tree import export_graphviz
import graphviz

dot_data = export_graphviz(clf, out_file=None, feature_names=x.columns, class_
graph = graphviz.Source(dot_data)
graph.render("decision_tree")
```

```
Out[ ]: 'decision_tree.pdf'
```