Upload the Dataset

```
from google.colab import files
uploaded = files.upload()
```



Choose Files card_transdata.csv

• card_transdata.csv(text/csv) - 76277977 bytes, last modified: 5/9/2025 - 100% done Saving card_transdata.csv to card_transdata.csv

Load the Dataset python

```
import pandas as pd
```

```
df = pd.read_csv("/content/card_transdata.csv")
df.head()
```

₹	(distance_from_home	distance_from_last_transaction	ratio_to_median_purchase_price	repeat_retailer	used_chip
	0	57.877857	0.311140	1.945940	1.0	1.0
	1	10.829943	0.175592	1.294219	1.0	0.0
	2	5.091079	0.805153	0.427715	1.0	0.0
	3	2.247564	5.600044	0.362663	1.0	1.0
	4	44.190936	0.566486	2.222767	1.0	1.0

Data Exploration

```
df.info()
df.describe()
df.shape
df.columns
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000000 entries, 0 to 999999
Data columns (total 8 columns):
 # Column
                                    Non-Null Count
                                                     Dtype
--- -----
                                    -----
 0 distance_from_home
                                    1000000 non-null float64
    distance_from_last_transaction 1000000 non-null float64
    ratio_to_median_purchase_price 1000000 non-null float64
    repeat_retailer
                                   1000000 non-null float64
                                    1000000 non-null float64
 4
    used chip
                                    1000000 non-null float64
    used_pin_number
                                    1000000 non-null float64
    online_order
     fraud
                                    1000000 non-null float64
dtypes: float64(8)
memory usage: 61.0 MB
Index(['distance_from_home', 'distance_from_last_transaction',
       'ratio_to_median_purchase_price', 'repeat_retailer', 'used_chip',
       'used_pin_number', 'online_order', 'fraud'],
      dtype='object')
```

Check for Missing Values and Duplicates

```
5/10/25, 9:59 AM
                                                                google colab.ipynb - Colab
    # Missing values
    df.isnull().sum()
    # Duplicates
    df.duplicated().sum()
     → np.int64(0)
    Visualize a Few Features
    print(df.columns.tolist())
     🚁 ['distance_from_home', 'distance_from_last_transaction', 'ratio_to_median_purchase_price', 'repeat_retailer', 'use
    ['distance_from_home', 'distance_from_last_transaction', 'ratio_to_median_purchase_price', 'repeat_retailer', 'used_chir

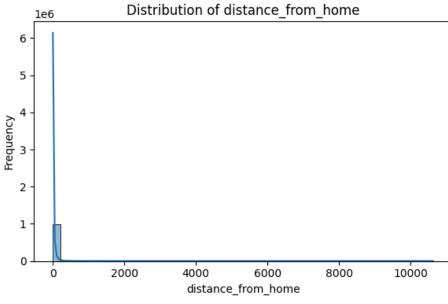
    ['distance_from_home',
           'distance_from_last_transaction',
           'ratio_to_median_purchase_price',
           'repeat_retailer',
           'used_chip',
           'used_pin_number',
           'online_order',
           'amount',
           'lat',
           'long',
           'is_fraud']
    df.head()
```

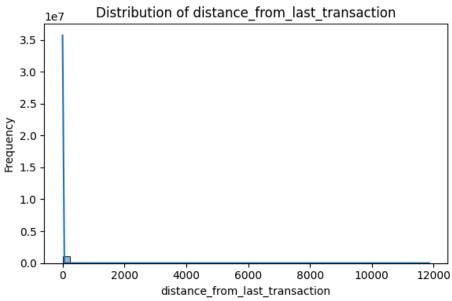
₹		distance_from_home	distance_from_last_transaction	ratio_to_median_purchase_price	repeat_retailer	used_chip
	0	57.877857	0.311140	1.945940	1.0	1.0
	1	10.829943	0.175592	1.294219	1.0	0.0
	2	5.091079	0.805153	0.427715	1.0	0.0
	3	2.247564	5.600044	0.362663	1.0	1.0
	4	44.190936	0.566486	2.222767	1.0	1.0

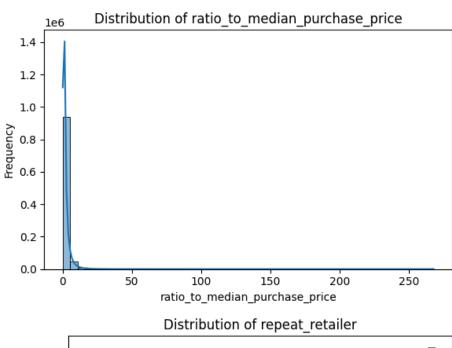
```
import seaborn as sns
import matplotlib.pyplot as plt
# Automatically select numeric columns
numeric_cols = df.select_dtypes(include=['float64', 'int64']).columns
# Histogram for each numeric column
for col in numeric_cols:
    plt.figure(figsize=(6, 4))
    sns.histplot(df[col], bins=50, kde=True)
   plt.title(f"Distribution of {col}")
   plt.xlabel(col)
   plt.ylabel("Frequency")
   plt.tight_layout()
   plt.show()
# Boxplot of all numeric features
plt.figure(figsize=(10, 6))
sns.boxplot(data=df[numeric cols])
```

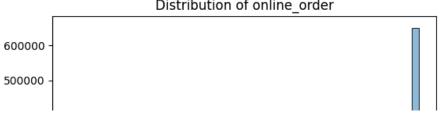
plt.title("Boxplot of Numeric Features")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()

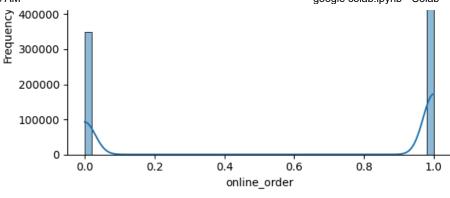


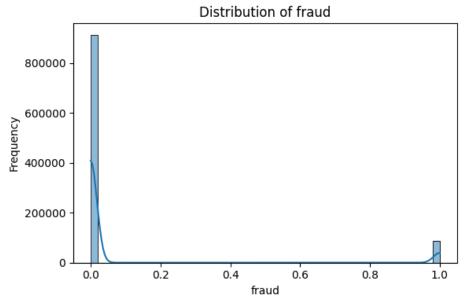


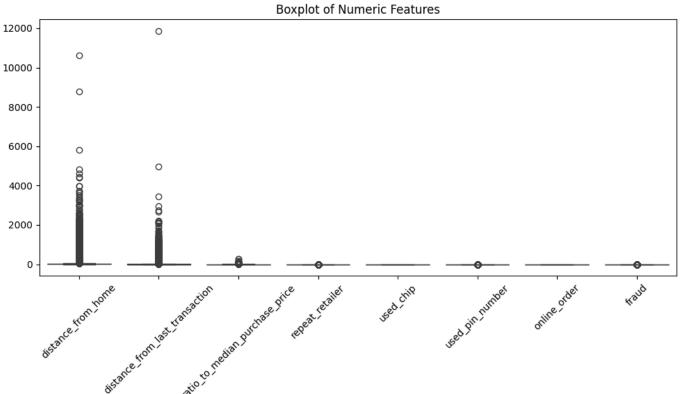












Identify Target and Features python

```
# Try to automatically detect the fraud column
target_candidates = [col for col in df.columns if 'fraud' in col.lower()]
if target_candidates:
    target = target_candidates[0] # Use the first match
    print(f"Detected target column: {target}")
    X = df.drop(columns=[target])
    y = df[target]
else:
    raise ValueError("No column containing 'fraud' found in the dataset.")
→ Detected target column: fraud
Convert Categorical Columns to Numerical
# Assuming 'category_column' as a placeholder for any categorical columns:
categorical_cols = X.select_dtypes(include='object').columns
X[categorical_cols] = X[categorical_cols].astype('category').apply(lambda x: x.cat.codes)
One-Hot Encoding
X = pd.get_dummies(X, columns=categorical_cols, drop_first=True)
Feature Scaling
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)
Train-Test Split
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)
Model Building
from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier(random_state=42)
model.fit(X_train, y_train)
→
            {\tt RandomForestClassifier}
     RandomForestClassifier(random_state=42)
```

Evaluation

```
from sklearn.metrics import classification_report, confusion_matrix
y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))
⋽₹
                   precision
                                recall f1-score
                                                    support
              0.0
                        1.00
                                  1.00
                                            1.00
                                                     182557
                        1.00
                                  1.00
              1.0
                                            1.00
                                                     17443
         accuracy
                                            1.00
                                                     200000
                                                     200000
        macro avg
                        1.00
                                  1.00
                                            1.00
     weighted avg
                        1.00
                                  1.00
                                            1.00
                                                     200000
     [[182557
            2 17441]]
Make Predictions from New Input
sample_input = X_test[0].reshape(1, -1)
model.predict(sample input)
\rightarrow array([0.])
Convert to DataFrame and Encode
import pandas as pd
# Load the dataset (replace with the correct path to your dataset)
df = pd.read_csv('/content/card_transdata.csv')
# Step 1: Print the column names to ensure 'Class' or the correct target column exists
print("Columns in dataset:", df.columns)
# Step 2: Clean up any spaces from the column names
df.columns = df.columns.str.strip()
# Step 3: Print the first few rows to visually check the data
print("First few rows:")
print(df.head())
# Step 4: Check for missing values (if any) and handle them (optional)
print("Missing values:")
print(df.isnull().sum())
# Step 5: Clean missing values or fill them with zero
df.fillna(0, inplace=True)
# Step 6: Check the data types to make sure all columns are as expected
print("Data types:")
print(df.dtypes)
# Step 7: If 'Class' is the target column (update accordingly if needed)
# For this example, I'm assuming 'Class' is the target column
X = df.drop('fraud', axis=1) # Drop the target column for features
y = df['fraud'] # Set the target column
# Print the features and target to confirm
print("Features (X):")
print(X.head())
```

```
print("Target (y):")
print(y.head())
                              0.427715
                                                     1.0
                                                                0.0
₹
     3
                              0.362663
                                                     1.0
                                                                1.0
     4
                              2.222767
                                                     1.0
                                                                1.0
        used_pin_number online_order fraud
                                          0.0
     0
                    0.0
                                  0.0
     1
                                          0.0
                    0.0
                                  0.0
     2
                    0.0
                                  1.0
                                          0.0
     3
                    0.0
                                  1.0
                                          0.0
     4
                    0.0
                                  1.0
                                         0.0
     Missing values:
     distance_from_home
                                        0
     distance_from_last_transaction
                                        0
     ratio_to_median_purchase_price
                                        0
     repeat_retailer
     used_chip
                                        0
     used_pin_number
                                        0
     online_order
                                        0
                                        0
     fraud
     dtype: int64
     Data types:
     distance_from_home
                                        float64
     distance_from_last_transaction
                                        float64
     ratio_to_median_purchase_price
                                        float64
     repeat retailer
                                        float64
                                        float64
     used_chip
                                        float64
     used_pin_number
     online_order
                                        float64
                                        float64
     fraud
     dtype: object
     Features (X):
        distance_from_home distance_from_last_transaction \
     0
                 57.877857
                                                   0.311140
                 10.829943
                                                   0.175592
     1
     2
                  5.091079
                                                   0.805153
     3
                  2.247564
                                                   5.600044
     4
                 44.190936
                                                   0.566486
        ratio_to_median_purchase_price repeat_retailer used_chip
                              1.945940
                                                     1.0
                                                                1.0
     1
                              1.294219
                                                     1.0
                                                                0.0
     2
                              0.427715
                                                     1.0
                                                                0.0
     3
                              0.362663
                                                     1.0
                                                                1.0
     4
                              2.222767
                                                     1.0
                                                                1.0
        used_pin_number online_order
     0
                    0.0
     1
                    0.0
                                  0.0
     2
                    0.0
                                  1.0
     3
                    0.0
                                  1.0
     4
                    0.0
                                  1.0
     Target (y):
     0
          0.0
     1
          0.0
     2
          0.0
     3
          0.0
          0.0
     Name: fraud, dtype: float64
Predict the Final Grade
```

```
def predict_fraud(V1, V2, V3, V28):
    # Create input DataFrame
    new_data = pd.DataFrame({
        'V1': [V1],
```