9/22/24, 9:38 PM Loan Data - Colab

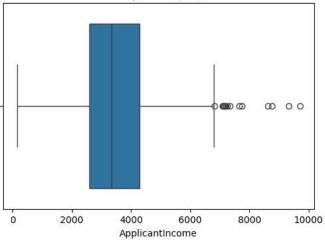
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import OneHotEncoder
from sklearn.impute import SimpleImputer
from \ sklearn.linear\_model \ import \ LinearRegression
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis as LDA
from sklearn.neighbors import KNeighborsClassifier
from sklearn.tree import DecisionTreeClassifier
from \ sklearn.metrics \ import \ accuracy\_score, \ confusion\_matrix, \ classification\_report
from google.colab import files
uploaded = files.upload()
    Choose Files No file chosen
                                         Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
data= pd.read_excel('Loan.xlsx')
data.head()
₹
         Gender Married Dependents
                                       Education Self_Employed ApplicantIncome CoapplicantIncome LoanAmount Loan_Amount_Term Credit_H
      0
                                                                              4583
                                                                                                 1508.0
           Male
                      Yes
                                    1
                                         Graduate
                                                              No
                                                                                                                128
                                                                                                                                  360.0
      1
           Male
                                    0
                                         Graduate
                                                              Yes
                                                                              3000
                                                                                                    0.0
                                                                                                                 66
                                                                                                                                  360.0
                      Yes
                                              Not
                                                                                                                                  360.0
                                                                              2583
                                                                                                 2358 0
      2
                                    0
                                                                                                                120
           Male
                      Yes
                                                              No
                                         Graduate
                                                                                                                                  360.0
      3
           Male
                                    0
                                         Graduate
                                                                              6000
                                                                                                    0.0
                                                                                                                141
                      No
                                                              No
categorical_cols = data.select_dtypes(include=['object']).columns
numerical_cols = data.select_dtypes(include=['int64', 'float64']).columns
imputer_num=SimpleImputer(strategy='mean')
data[numerical_cols]=imputer_num.fit_transform(data[numerical_cols])
imputer_cat=SimpleImputer(strategy='most_frequent')
data[categorical_cols]=imputer_cat.fit_transform(data[categorical_cols])
data_encoded=pd.get_dummies(data, columns=categorical_cols, drop_first=True)
data_encoded.head()
<del>_____</del>
         ApplicantIncome CoapplicantIncome LoanAmount Loan_Amount_Term Credit_History Gender_Male Married_Yes Dependents_1 Depende
                   4583.0
                                       1508.0
                                                     128.0
                                                                        360.0
                                                                                           1.0
                                                                                                       True
                                                                                                                     True
                                                                                                                                    True
      1
                   3000.0
                                          0.0
                                                      66.0
                                                                        360.0
                                                                                           1.0
                                                                                                       True
                                                                                                                     True
                                                                                                                                   False
      2
                   2583.0
                                       2358.0
                                                     120.0
                                                                        360.0
                                                                                           1.0
                                                                                                       True
                                                                                                                     True
                                                                                                                                   False
      3
                   6000.0
                                          0.0
                                                     141.0
                                                                        360.0
                                                                                           1.0
                                                                                                       True
                                                                                                                    False
                                                                                                                                   False
      4
                   2333.0
                                       1516.0
                                                                        360.0
                                                      95.0
                                                                                           1.0
                                                                                                       True
                                                                                                                     True
                                                                                                                                   False
```

for columns in numerical_cols:
 plt.figure(figsize=(6,4))
 sns.boxplot(data=data_encoded, x=columns)
 plt.title('Boxplot for {col}')
 plt.show()

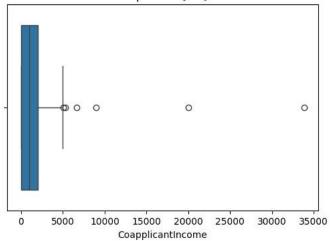
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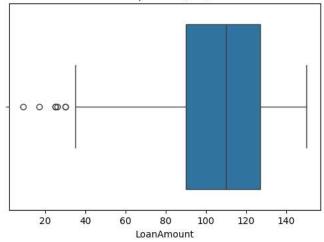




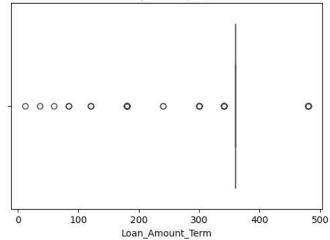
Boxplot for {col}



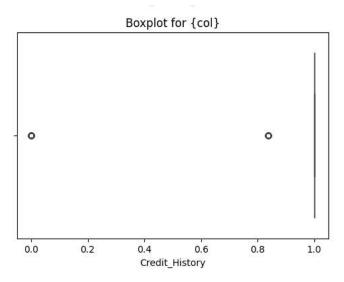
Boxplot for {col}



Boxplot for {col}



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```
q1= data_encoded[numerical_cols].quantile(0.25)
q3= data_encoded[numerical_cols].quantile(0.75)
iqr= q3-q1

outliers= ((data_encoded[numerical_cols]<(q1-1.5*iqr))|(data_encoded[numerical_cols]>(q3+1.5*iqr))).any(axis=1)

outliers.sum()

149

X = data_encoded.drop(columns=['Loan_Status_Y'])
y = data_encoded['Loan_Status_Y']

x_train, x_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=42)

x_train.shape, x_test.shape
((266, 14), (115, 14))

Logistic Regression

from sklearn.linear_model import LogisticRegression
log_reg = LogisticRegression(max_iter=1000)
log_reg.fit(x_train, y_train)
```

y_pred = log_reg.predict(x_test)

v LogisticRegression
LogisticRegression(max_iter=1000)

print("Logistic Regression Accuracy:",accuracy_score(y_test, y_pred))
print(confusion_matrix(y_test, y_pred))
print(classification_report(y_test, y_pred))

Logistic Regression Accuracy: 0.8173913043478261 [[14 21] [0 80]]

	precision	recall	f1-score	support
False	1.00	0.40	0.57	35
True	0.79	1.00	0.88	80
accuracy			0.82	115
macro avg	0.90	0.70	0.73	115
weighted avg	0.86	0.82	0.79	115

Linear Discriminant Analysis (LDA)

lda=LDA()

```
lda.fit(x_train, y_train)
     ▼ LinearDiscriminantAnalysis
     LinearDiscriminantAnalysis()
y_pred_lda=lda.predict(x_test)
print("LDA Accuracy:",accuracy_score(y_test, y_pred_lda))
print(confusion_matrix(y_test, y_pred_lda))
print(classification_report(y_test, y_pred_lda))
→ LDA Accuracy: 0.808695652173913
     [[14 21]
      [ 1 79]]
                   precision
                                recall f1-score
                                                   support
            False
                        0.93
                                  0.40
                                            0.56
                                                         35
                        0.79
                                  0.99
                                                         80
             True
                                            0.88
                                            0.81
                                                       115
         accuracy
                        0.86
                                  0.69
                                            0.72
        macro avg
                                                        115
     weighted avg
                        0.83
                                  0.81
                                            0.78
                                                       115
Start coding or generate with AI.
K-Nearest Neighbours KNN
knn=KNeighborsClassifier(n\_neighbors=5)
knn.fit(x_train, y_train)
    KNeighborsClassifier
     KNeighborsClassifier()
y_pred_knn=knn.predict(x_test)
print("KNN Accuracy:",accuracy_score(y_test, y_pred_knn))
print(confusion_matrix(y_test, y_pred_knn))
print(classification_report(y_test, y_pred_knn))
    KNN Accuracy: 0.6869565217391305
     [[10 25]
      [11 69]]
                   precision
                                recall f1-score
                                                   support
                                  0.29
                        0.48
                                            0.36
                                                         35
            False
                        0.73
                                            0.79
                                                         80
             True
                                  0.86
         accuracy
                                            0.69
                                                        115
        macro avg
                        0.61
                                  0.57
                                            0.58
                                                        115
     weighted avg
                        0.66
                                  0.69
                                            0.66
                                                        115
Decision Tree Classifier
tree=DecisionTreeClassifier()
tree.fit(x_train, y_train)
     ▼ DecisionTreeClassifier
     DecisionTreeClassifier()
y_pred_tree=tree.predict(x_test)
print("Decision Tree Accuracy:",accuracy_score(y_test, y_pred_tree))
print(confusion_matrix(y_test, y_pred_tree))
print(classification_report(y_test, y_pred_tree))
    Decision Tree Accuracy: 0.7565217391304347
     [[19 16]
      [12 68]]
                                recall f1-score
                   precision
                                                   support
                        0.61
                                  0.54
                                            0.58
            False
                                                         35
             True
                        0.81
                                  0.85
                                            0.83
                                                         80
```