FAKE BILLS PREDICTION

In [41]: **import** numpy **as** np **import** pandas **as** pd **import** seaborn **as** sns

import matplotlib.pyplot as plt from sklearn.model_selection import

train_test_split **from** sklearn.linear_model **import** LogisticRegression

from sklearn.metrics **import** mean_absolute_error,mean_squared_error, r2_score **from** sklearn.preprocessing **import** StandardScaler

from sklearn.metrics import accuracy_score,classification_report,confusion_mat

In [13]:
df = pd.read_excel("fake_bills_converted.xlsx")

In [14]: df

Out[14]: is_genuine diagonal height_left height_right margin_low margin_up le

39 11	2.8	4.52	104.95	104.86	171.81	1	0
	2.99	3.77	103.66	103.36	171.46	1	. 1
	2.94	4.40	103.50	104.48	172.69	1	2
	3.01	3.62	103.94	103.91	171.36	1	3
11							
11	3.48	4.04	103.46	104.28	171.73	1	4
11	3.09	4.42	104.17	104.38	171.75	О	1495
11	3.37	5.27	104.44	104.63	172.19	0	1496
11	3.36	5.51	104.12	104.01	171.80	0	1497
11	3.46	5.17	104.06	104.28	172.06	0	1498
11	3.37	4.63	103.82	104.15	171.47	0	1499

1500 rows × 7 columns

In [15]: df.isnull().sum()

margin_low 37 margin_up
0 length 0 dtype: int64

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In [16]:
             df["margin_low"]=df["margin_low"].fillna(df["margin_low"].mean())
In [17]:
             df["is genuine"].unique()
Out[17]: array([1, 0], dtype=int64)
In [18]:
             df.isnull().sum()
Out[18]: is_genuine
                     0 diagonal
             height_left 0 height_right 0
             margin_low 0 margin_up
             length
                       0 dtype: int64
In [19]:
             x=df.drop(columns=["is_genuine"])
            y=df["is_genuine"]
In [20]:
Out[20]:
                                   diagonal height_left height_right margin_low margin_up length
                  0
                                         104.86
                         171.81
                                                           104.95
                                                                              4.52
                                                                                                2.89 112.83
                  1
                                        103.36
                                                                                             2.99
                                                                                                     113.09
                         171.46
                                                          103.66
                                                                              3.77
                  2
                         172.69
                                        104.48
                                                          103.50
                                                                              4.40
                                                                                             2.94
                                                                                                     113.16
                  3
                         171.36
                                        103.91
                                                          103.94
                                                                              3.62
                                                                                             3.01
                                                                                                     113.51
                  4
                         171.73
                                        104.28
                                                          103.46
                                                                              4.04
                                                                                             3.48
                                                                                                     112.54
             1495
                         171.75
                                        104.38
                                                          104.17
                                                                              4.42
                                                                                             3.09
                                                                                                     111.28
             1496
                                                                                                     110.97
                         172.19
                                        104.63
                                                          104.44
                                                                              5.27
                                                                                             3.37
             1497
                         171.80
                                        104.01
                                                          104.12
                                                                                                     111.95
                                                                              5.51
                                                                                             3.36
             1498
                         172.06
                                        104.28
                                                                                             3.46
                                                                                                     112.25
                                                          104.06
                                                                              5.17
             1499
                         171.47
                                        104.15
                                                          103.82
                                                                              4.63
                                                                                             3.37
                                                                                                     112.07
            1500 rows × 6 columns
In [21]:
Out[21]: 0
             11
                   1
             2
                       1
             3
                       1
             4
                       1
```

```
1496
            1497
                       0
             1498
                       0
             1499
                       0
             Name: is genuine, Length: 1500, dtype: int64
In [22]:
             x_train,x_test,y_train,y_test,=train_test_split(x,y,train_size=0.8,random_stat
In [23]:
             s=StandardScaler()
In [24]:
             x_train=s.fit_transform(x_train) x_test=s.fit_transform(x_test)
In [25]:
             model=LogisticRegression()
In [26]:
             model.fit(x train,y train)
Out[26]: 2 LogisticRegression
            LogisticRegression()
In [27]:
            y_pred=model.predict(x_test)
In [28]:
            y pred
 0, 1,
                      1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
                      1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1,
                      1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1,
                      0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 0,
                      1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1,
                      0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1,
                      1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1,
                      0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
                      1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1,
                      0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1,
                      1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1], dtype=int64)
In [29]:
             np.array(y test)
 0, 1,
                      0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
                      1, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1,
                      1, 1, 0, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 1,
                      1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1,
```

1495

0

```
0, 1, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
                        1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1,
                        0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1,
                        1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1], dtype=int64)
In [30]:
              accuracy_score(y_test,y_pred)
In [31]:
              accuracy
Out[31]: 0.99
In [32]:
              print(classification_report(y_test,y_pred))
                                                      precision recall f1-score support
                                                  0
                                                              1.00
                                                                      0.97
                                                                              0.99
                                                                                       110
                                                  1
                                                              0.98
                                                                      1.00
                                                                              0.99
                                                                                       190
                  accuracy
                                          0.99
                                                   300 macro avg
                                                                      0.99
                                                                              0.99
                                                                                      0.99
           300 weighted avg
                                 0.99
                                         0.99
                                                 0.99
                                                          300
In [40]:
              cm = confusion_matrix(y_test, y_pred)
              # Plot heatmap
              sns.heatmap(cm, annot=True, fmt="d", cmap="Blues", xticklabels=["Pred 0", "Pre plt.ylabel("Actual")
              plt.xlabel("Predicted") plt.title("Confusion Matrix")
              plt.show()
```

