PHASE-5 STOCK PRICE PREDICTION

```
In [21]: import seaborn as sb
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   In [31]: features = ['Open', 'High', 'Low', 'Close', 'Volume']
             plt.subplots(figsize=(10,8))
             for i, col in enumerate(features):
                 plt.subplot(2,3,i+1)
                 sb.distplot(stock[col])
                 plt.show()
                 0.08
                 0.07
                 0.06
                 0.05
               Density
60.0
                 0.03
                 0.02 -
                 0.01 -
                 0.00
                                               150
                                 50
                                       100
                                    Onen
```

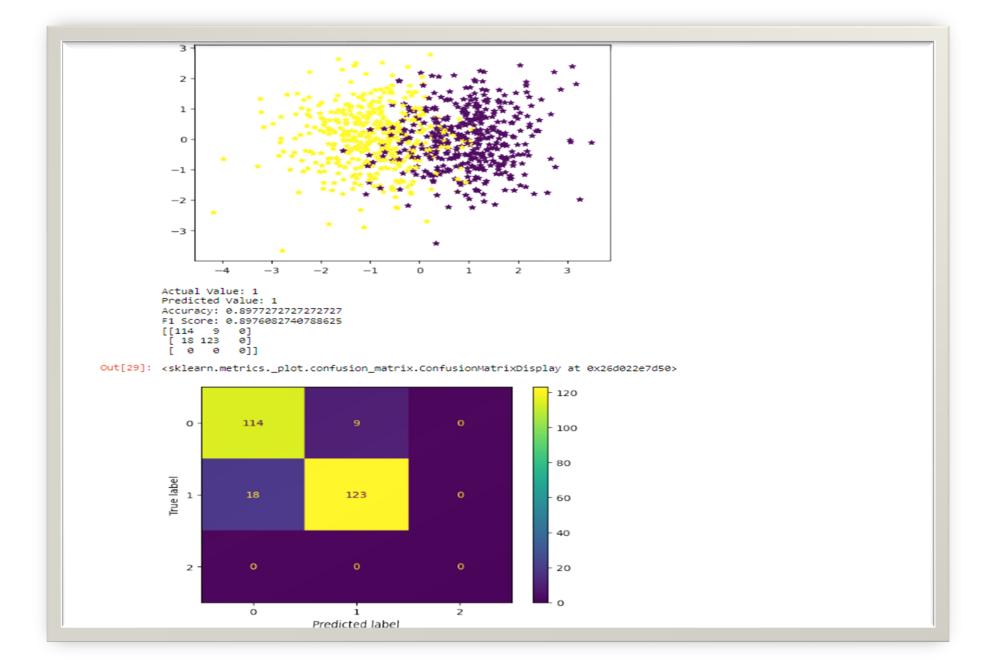


In [26]: plt.figure(figsize=(10, 10)) sb.heatmap(stock.corr() > 0.9, annot=True, cbar=False) plt.show() C:\Users\CSE LAB\AppData\Local\Temp\ipykernel_5832\4185554148.py:2: FutureWarning: The default value of numeric_only in DataFra me.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric _only to silence this warning. sb.heatmap(stock.corr() > 0.9, annot=True, cbar=False) 1 1 1 1 0 1 1 1 1 0 1 1 1 1 0 1 1 1 1 1 0 1 1 1 1 0 Open High Low Close Adj Close Volume

```
In [28]: from sklearn.model_selection import train_test_split
         from sklearn.linear_model import LogisticRegression
         from sklearn.metrics import accuracy score, classification report, confusion matrix
         import numpy as np
         np.random.seed(0)
         data = {
             'Exam1': np.random.rand(100) * 100,
             'Exam2': np.random.rand(100) * 100,
             'Admitted': np.random.randint(2, size=100)
         df = pd.DataFrame(data)
         print(df)
         X = df[['Exam1', 'Exam2']]
         y = df['Admitted']
         print(X)
         print(y)
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
         model = LogisticRegression()
         model.fit(X train, y train)
         y pred = model.predict(X test)
         print("----")
         print(y_pred)
         accuracy = accuracy_score(y_test, y_pred)
         print(f'Accuracy: {accuracy:.2f}')
         print(classification_report(y_test, y_pred))
         print(confusion_matrix(y_test, y_pred))
```

```
Exam2 Admitted
       Exam1
0 54.881350 67.781654
1 71.518937 27.000797
                              1
2 60.276338 73.519402
3 54.488318 96.218855
4 42.365480 24.875314
        . . .
                            . . .
95 18.319136 49.045881
96 58.651293 22.741463
                            1
97 2.010755 25.435648
98 82.894003 5.802916
99 0.469548 43.441663
[100 rows x 3 columns]
       Exam1
                 Exam2
0 54.881350 67.781654
1 71.518937 27.000797
2 60.276338 73.519402
3 54.488318 96.218855
4 42.365480 24.875314
95 18.319136 49.045881
96 58.651293 22.741463
97 2.010755 25.435648
98 82.894003 5.802916
99 0.469548 43.441663
[100 rows x 2 columns]
0
     0
     1
3
95
96
     1
97
     8
98
     1
Name: Admitted, Length: 100, dtype: int32
[10100101100011010001]
Accuracy: 0.45
            precision recall f1-score support
                                   0.35
          8
                 0.27
                          0.50
                                               6
                 0.67
                          0.43
                                   0.52
                                              14
  accuracy
                                   0.45
                                              20
  macro avg
                 0.47
                          0.46
                                   0.44
                                              20
weighted avg
                 0.55
                                   0.47
                                              20
                          0.45
[[3 3]
[8 6]]
```

```
In [29]: from sklearn.datasets import make classification
         value1, y = make_classification(n_features=6,n_classes=2,n_samples=800,n_informative=2,random_state=66,n_clusters_per_class=1)
         import matplotlib.pyplot as plt
         plt.scatter(value1[:, 0], value1[:, 1], c=y, marker="*")
         plt.show()
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(value1, y, test_size=0.33, random_state=125)
         from sklearn.naive_bayes import GaussianNB
         model = GaussianNB()
         model.fit(X_train, y_train)
         predicted = model.predict([X_test[6]])
         print("Actual Value:", y_test[6])
         print("Predicted Value:", predicted[0])
         from sklearn.metrics import (accuracy_score,confusion_matrix,ConfusionMatrixDisplay,f1_score,)
         y_pred = model.predict(X_test)
         accuracy = accuracy_score(y_pred, y_test)
         f1 = f1_score(y_pred, y_test, average="weighted")
         print("Accuracy:", accuray)
         print("F1 Score:", f1)
         labels = [0,1,2]
         cm = confusion_matrix(y_test, y_pred, labels=labels)
         print(cm)
         disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=labels)
         disp.plot()
```



CONCLUSION

 To preprocess the dataset for stock price prediction, you can consider steps like removing duplicates, handling missing values, scaling the data, and splitting it into training and testing sets. **Team Member**

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