```
In [1]: import pandas as pd
        from sklearn.preprocessing import StandardScaler
        from sklearn.decomposition import PCA
        from sklearn.model selection import train test split
        from sklearn.naive bayes import GaussianNB
        from sklearn.metrics import accuracy score
        from sklearn.metrics import confusion matrix, classification report
        import seaborn as sns
        import matplotlib.pyplot as plt
        Importing the Dataset
In [2]: file = 'spambase.data'
        data = pd.read csv(file, header=None)
        print(data.head())
            0
                                                         8
                                                               9
                                                                          48 \
                            3
                                  4
                                        5
                                              6
                                                    7
               0.64
                     0.64
                           0.0 0.32
                                     0.00
                                           0.00
                                                 0.00
                                                       0.00
                                                             0.00
                                                                        0.00
               0.28
                     0.50
                           0.0
                                0.14
                                      0.28
                                           0.21
                                                 0.07
                                                       0.00
                                                             0.94
                                                                        0.00
         0.06 0.00 0.71 0.0 1.23 0.19 0.19 0.12 0.64
                                                                        0.01
                                                             0.25
                           0.0 0.63 0.00 0.31 0.63 0.31
                                                                        0.00
               0.00
                    0.00
                                                             0.63
         0.00 0.00 0.00 0.0 0.63 0.00 0.31 0.63 0.31 0.63 ...
                                                                        0.00
                               52
                 50
                        51
                                      53
                                             54
                                                  55
                                                        56 57
         0.000 0.0 0.778 0.000
                                   0.000
                                         3.756
                                                 61
                                                      278
                                                           1
         0.132 0.0 0.372 0.180
                                         5.114 101
                                   0.048
                                                     1028
         0.143 0.0 0.276
                            0.184 0.010
                                         9.821 485
                                                     2259
                                                            1
         0.137 0.0 0.137 0.000 0.000 3.537
                                                 40
                                                      191
                                                            1
                                                      191
                                                            1
         0.135 0.0 0.135 0.000 0.000 3.537
       [5 rows x 58 columns]
In [3]: data.columns = [
            'word freq make',
            'word freq address',
            'word freq all',
            'word freq 3d',
            'word freq our',
            'word freq over',
            'word freq remove',
            'word freq internet',
            'word freq order',
            'word freq mail',
```

```
'word_freq_receive',
'word_freq_will',
'word_freq_people',
'word freq report',
'word_freq_addresses',
'word freq free',
'word freq business',
'word freq email',
'word_freq_you',
'word freq credit',
'word freq your',
'word freq font',
'word freq 000',
'word freq money',
'word freq hp',
'word freq hpl',
'word_freq_george',
'word_freq_650',
'word freq lab',
'word freq labs',
'word_freq_telnet',
'word_freq_857',
'word freq data',
'word freq 415',
'word freq 85',
'word_freq_technology',
'word_freq_1999',
'word_freq_parts',
'word_freq_pm',
'word_freq_direct',
'word_freq_cs',
'word freq meeting',
'word_freq_original',
'word_freq_project',
'word_freq_re',
'word_freq_edu',
'word_freq_table',
'word_freq_conference',
'char_freq_;',
'char_freq_(',
'char_freq_[',
'char_freq_!',
'char_freq_$',
'char_freq_#',
'capital_run_length_average',
```

```
'capital_run_length_longest',
             'capital_run_length_total',
             'spam'
In [4]: data.head()
Out[4]:
            word freq make word freq address word freq all word freq 3d word freq our word freq over word freq remove word freq internet w
                      0.00
                                                       0.64
                                                                      0.0
                                                                                   0.32
                                                                                                   0.00
                                                                                                                     0.00
                                                                                                                                        0.00
         0
                                         0.64
         1
                       0.21
                                         0.28
                                                       0.50
                                                                      0.0
                                                                                   0.14
                                                                                                   0.28
                                                                                                                     0.21
                                                                                                                                        0.07
         2
                      0.06
                                         0.00
                                                       0.71
                                                                      0.0
                                                                                   1.23
                                                                                                   0.19
                                                                                                                     0.19
                                                                                                                                        0.12
                                                       0.00
         3
                       0.00
                                         0.00
                                                                      0.0
                                                                                   0.63
                                                                                                   0.00
                                                                                                                     0.31
                                                                                                                                        0.63
                      0.00
                                         0.00
                                                       0.00
                                                                      0.0
                                                                                   0.63
                                                                                                   0.00
                                                                                                                     0.31
                                                                                                                                        0.63
         4
        5 rows × 58 columns
        Pre-process the dataset
In [5]: duplicated rows = data.duplicated()
        sum duplicate = sum(duplicated rows)
        print(f"Number of duplicate rows: {sum duplicate}")
       Number of duplicate rows: 391
In [6]: data = data.drop duplicates()
In [7]: data.loc[(data['word_freq_george'] > 0) | (data['word_freq_650'] > 0), 'spam'] = 0
In [8]: columns_to_display = ['word_freq_george', 'word_freq_650', 'spam']
        df_subset = data.loc[:, columns_to_display]
        df subset
```

Out[8]:		word_freq_george	word_freq_650	spam
	0	0.0	0.0	1
	1	0.0	0.0	1
	2	0.0	0.0	1
	3	0.0	0.0	1
	4	0.0	0.0	1
	•••		•••	•••
	4596	0.0	0.0	0
	4597	0.0	0.0	0
	4598	0.0	0.0	0
	4599	0.0	0.0	0
	4600	0.0	0.0	0

4210 rows × 3 columns

```
In [9]: data.to_csv('spambase.csv', index=False)
    new_data = pd.read_csv('spambase.csv', header=0)
    new_data.head()
```

Out[9]:		word_freq_make	word_freq_address	word_freq_all	word_freq_3d	word_freq_our	word_freq_over	word_freq_remove	word_freq_internet	W
	0	0.00	0.64	0.64	0.0	0.32	0.00	0.00	0.00	
	1	0.21	0.28	0.50	0.0	0.14	0.28	0.21	0.07	
	2	0.06	0.00	0.71	0.0	1.23	0.19	0.19	0.12	
	3	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	
	4	0.00	0.00	0.00	0.0	0.63	0.00	0.31	0.63	

5 rows × 58 columns

```
In [10]: X = \text{new data.iloc}[:, 0:57]
         y = new data.iloc[:, 57]
In [11]: scaler = StandardScaler()
         X scaled = scaler.fit transform(X)
         X scaled
Out[11]: array([[-0.34792164, 1.16102457, 0.67588944, ..., -0.04911671,
                  0.04439849, -0.02130997],
                [0.3521501, 0.3684328, 0.40439129, ..., -0.00814327,
                  0.24484101, 1.1911417 ],
                [-0.14790114, -0.24802746, 0.81163851, ..., 0.13387587,
                  2.16908914, 3.18117903],
                [0.65218084, -0.24802746, 0.01653679, ..., -0.12008102,
                 -0.23120996, -0.27996632],
                [2.85240628, -0.24802746, -0.56524496, ..., -0.12783519,
                 -0.23622103, -0.34463041],
                [-0.34792164, -0.24802746, 0.69528216, ..., -0.12472749,
                 -0.23622103, -0.40606129]])
In [12]: pca = PCA(0.95)
         X pca = pca.fit transform(X scaled)
         X.shape, X pca.shape
Out[12]: ((4210, 57), (4210, 49))
In [13]: X train, X test, y train, y test = train test split(X pca, y, test size=0.2, random state=42)
In [14]: | classifier = GaussianNB()
         classifier.fit(X_train, y_train)
Out[14]:
         ▼ GaussianNB
         GaussianNB()
In [15]: y_pred = classifier.predict(X_test)
```

Model results and Confusion Matrix

```
In [16]: accuracy = accuracy_score(y_test, y_pred)
         print(f"Accuracy: {accuracy:.2f}")
         print("Classification Report:")
         print(classification report(y test, y pred))
        Accuracy: 0.86
        Classification Report:
                                   recall f1-score
                      precision
                                                      support
                   0
                                     0.86
                                               0.88
                           0.90
                                                          490
                           0.82
                                     0.87
                                               0.84
                   1
                                                          352
                                               0.86
                                                          842
            accuracy
                                               0.86
                                                          842
           macro avg
                           0.86
                                     0.87
        weighted avg
                           0.87
                                     0.86
                                               0.87
                                                          842
In [17]: cm = confusion matrix(y test,y pred)
         \mathsf{cm}
Out[17]: array([[421, 69],
                [ 45, 307]])
In [18]: sns.heatmap(cm, annot=True, fmt="d", cmap='Blues')
         plt.xlabel('Predicted')
         plt.ylabel('True')
         plt.title('Confusion Matrix')
         plt.show()
```

