Classify the email using the binary classification method. Email Spam detection has two states: a) Normal State – Not Spam, b) Abnormal State – Spam. Use K-Nearest Neighbors and Support Vector Machine for classification. Analyze their performance. Dataset link: The emails.csv dataset on the Kaggle https://www.kaggle.com/datasets/balaka18/email-spam-classification-dataset-csv

```
import pandas as pd
import numpy as np
from sklearn.model selection import train test split
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
df = pd.read csv('emails.csv')
df
                    the
        Email No.
                          to
                               ect
                                     and
                                           for
                                                 of
                                                           you
connevey
          Email 1
                       0
                            0
                                                        2
                                  1
                                                  0
0
1
          Email 2
                       8
                          13
                                24
                                       6
                                             6
                                                  2
                                                     102
                                                              1
                                                                  27
0
2
          Email 3
                            0
                                 1
                                       0
                                                  0
                                                        8
0
3
                            5
                                22
          Email 4
                       0
                                       0
                                                  1
                                                       51
                                                              2
                                                                  10
0
4
          Email 5
                            6
                                17
                                                       57
0
. . .
5167
       Email 5168
                            2
                                 2
                                                       32
                                                  0
       Email 5169
5168
                      35
                          27
                                11
                                       2
                                                  5
                                                     151
                                                                   3
5169
       Email 5170
                            0
                                                       11
5170
       Email 5171
                       2
                                                       28
                                                  1
                                                              2
5171
      Email 5172
                      22
                          24
                                  5
                                                     148
                                                             8
                                                  5
                                              military
                                                          allowing
                            infrastructure
       jay
            valued
                      lay
                                                                      ff
                                                                          dry
                                                                                1
0
         0
                        0
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                  0
                                                                             0
1
         0
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2
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3
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4
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5167
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5168
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5169
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5170
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5171
        0
      Prediction
0
1
                0
2
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3
                0
4
                0
5167
                0
5168
                0
                1
5169
5170
                1
5171
                0
[5172 rows x 3002 columns]
df.shape
(5172, 3002)
df.isnull().any()
Email No.
               False
the
               False
to
               False
               False
ect
and
               False
military
               False
allowing
               False
ff
               False
dry
               False
               False
Prediction
Length: 3002, dtype: bool
df.drop(columns='Email No.', inplace=True)
df
      the to ect and for of
                                   a you
                                              hou in ... connevey
jay ∖
0
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             0
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5168	35	27	11	2	6	5	151	4	3	23			0	
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5171	22	24	5	1	6	5	148	8	2	23			0	
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5168		0	0			0		0		0	1	0		
0 5169		0	0			0		0		0	0	0		
1		U	U			U		U		U	U	U		
5170		0	0			0		0		0	1	0		
1														
5171		0	0			0		0		0	0	0		
0														
[5172	rows	x 3	3001 c	olumns]									
df.co	lumns													
<pre>Index(['the', 'to', 'ect', 'and', 'for', 'of', 'a', 'you', 'hou', 'in',</pre>														
<pre> 'connevey', 'jay', 'valued', 'lay', 'infrastructure',</pre>														
1			ey',	'jay',	'va	Lued	', 'la	ay', '	infr	astru	cture	٠,		
'military',														

```
'allowing', 'ff', 'dry', 'Prediction'],
      dtype='object', length=3001)
df.Prediction.unique()
array([0, 1], dtype=int64)
df['Prediction'] = df['Prediction'].replace({0:'Not spam', 1:'Spam'})
df
      the to ect and for of a you how in ... connevey
jay
0
        0
             0
                  1
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            13
                 24
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3
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             6
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5167
                  2
                        3
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5168
       35
            27
                 11
                        2
                             6
                                  5
                                     151
                                                     23
                                                                       0
                                                                       0
5169
                        1
                             0
                                  0
                                      11
                                             0
                                                  0
                                                      1
5170
        2
                             2
                                  1
                                      28
                                             2
                                                  0
                                                                       0
                        0
                                                      8
5171
       22 24
                 5
                        1
                          6
                                  5
                                     148
                                            8
                                                  2 23
                                                                       0
      valued lay infrastructure military allowing ff dry
Prediction
                                   0
                                              0
                                                         0
                                                             0
                                                                  0
                                                                        Not
spam
                 0
                                   0
                                                         0
                                                                  0
            0
                                                             1
                                                                        Not
spam
            0
                 0
                                                             0
                                                                  0
                                                                        Not
2
spam
3
            0
                 0
                                                         0
                                                             0
                                                                  0
                                                                        Not
spam
                 0
                                                                        Not
                                                         0
                                                             1
spam
. . .
. . .
```

5167	0	0	0	0	0	0	0	Not
spam	_	_		_		_	_	
5168	0	0	0	0	0	1	0	Not
spam	_	_	_	_	_	_	_	
5169	0	0	0	0	0	0	0	
Spam								
5170	0	0	0	0	0	1	0	
Spam								
5171	0	0	0	0	0	0	0	Not
spam								
[5172 rows	s x 3	001 colum	ins]					

KNN

```
X = df.drop(columns='Prediction',axis = 1)
Y = df['Prediction']
X.columns
Index(['the', 'to', 'ect', 'and', 'for', 'of', 'a', 'you', 'hou',
'in',
        'enhancements', 'connevey', 'jay', 'valued', 'lay',
'infrastructure',
       'military', 'allowing', 'ff', 'dry'],
      dtype='object', length=3000)
Y.head()
     Not spam
0
1
     Not spam
2
     Not spam
3
     Not spam
4
     Not spam
Name: Prediction, dtype: object
x_train, x_test, y_train, y_test = train_test_split(X, Y,
test size=0.2, random state=1)
KN = KNeighborsClassifier
knn = KN(n neighbors=7)
knn.fit(x_train, y_train)
y \text{ pred} = \overline{knn.predict}(x \text{ test})
print("Prediction: \n")
print(y pred)
```

```
Prediction:
['Not spam' 'Spam' 'Not spam' ... 'Not spam' 'Not spam' 'Not spam']
# Accuracy

M = metrics.accuracy_score(y_test,y_pred)
print("KNN accuracy: ", M)

KNN accuracy: 0.8714975845410629

C = metrics.confusion_matrix(y_test,y_pred)
print("Confusion matrix: ", C)

Confusion matrix: [[635 84]
  [ 49 267]]
```

SVM Classifier