Assignment 2

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
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings('ignore')
from sklearn.model selection import train test split
from sklearn.svm import SVC
from sklearn import metrics
df=pd.read csv('emails.csv')
df.head()
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```

[5 rows x 3002 columns]

```
df.columns
Index(['Email No.', 'the', 'to', 'ect', 'and', 'for', 'of', 'a',
'you', 'hou',
       'connevey', 'jay', 'valued', 'lay', 'infrastructure',
'military',
       'allowing', 'ff', 'dry', 'Prediction'],
      dtype='object', length=3002)
df.isnull().sum()
Email No.
the
              0
              0
to
ect
              0
and
              0
military
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allowing
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ff
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dry
Prediction
              0
Length: 3002, dtype: int64
df.dropna(inplace = True)
df.drop(['Email No.'],axis=1,inplace=True)
X = df.drop(['Prediction'],axis = 1)
y = df['Prediction']
from sklearn.preprocessing import scale
X = scale(X)
# split into train and test
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.3, random state = 42)
##KNN classifier
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n neighbors=7)
knn.fit(X train, y train)
y pred = \overline{k}nn.predict(X_test)
print("Prediction",y pred)
Prediction [0 0 1 ... 1 1 1]
print("KNN accuracy = ",metrics.accuracy_score(y_test,y_pred))
KNN \ accuracy = 0.8009020618556701
```

```
print("Confusion matrix", metrics.confusion_matrix(y_test,y_pred))
Confusion matrix [[804 293]
[ 16 439]]
SVM classifier
\# cost C = 1
model = SVC(C = 1)
# fit
model.fit(X_train, y_train)
# predict
y_pred = model.predict(X_test)
metrics.confusion_matrix(y_true=y_test, y_pred=y_pred)
array([[1091,
                 6],
              365]], dtype=int64)
       [ 90,
print("SVM accuracy = ",metrics.accuracy_score(y_test,y_pred))
SVM \ accuracy = 0.9381443298969072
```