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BE A Computer

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Classify emails 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

## Importing Libraries

```
In [1]: 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.preprocessing import scale
    from sklearn.model_selection import train_test_split
    from sklearn.neighbors import KNeighborsClassifier
    from sklearn.svm import SVC
    from sklearn import metrics
```

```
In [2]: df = pd.read_csv('./Datasets/emails.csv')
    df.head()
```

```
Out[2]:
            Email
                   the to ect and for of a you hou ... connevey jay valued
              No.
            Email
                         0
                              1
                                   0
                                        0
                                           0
                                                 2
                                                      0
                                                           0
                                                                               0
                                                                                        0
                1
            Email
                     8 13
                             24
                                            2 102
                                                      1
                                                          27
                                                                               0
                                                                                        0
                                   6
                                        6
            Email
         2
                         0
                              1
                                   0
                                        0
                                           0
                                                 8
                                                      0
                                                           0
                                                                               0
                                                                                        0
                                                                          0
            Email
                         5
                             22
                                                                                        0
         3
                                   0
                                        5
                                            1
                                                51
                                                      2
                                                          10
                                                                               0
            Email
                         6
                             17
                                   1
                                        5
                                           2
                                                57
                                                      0
                                                           9 ...
                                                                               0
                                                                                        0
                5
        5 \text{ rows} \times 3002 \text{ columns}
In [3]: df.columns
Out[3]: Index(['Email No.', 'the', 'to', 'ect', 'and', 'for', 'of', 'a', 'you', 'ho
         u',
                 'connevey', 'jay', 'valued', 'lay', 'infrastructure', 'military',
                 'allowing', 'ff', 'dry', 'Prediction'],
               dtype='object', length=3002)
In [4]: df.isnull().sum()
Out[4]: Email No.
                        0
         the
                        0
         to
                        0
                        0
         ect
         and
                        0
         military
                        0
         allowing
                        0
         ff
                        0
         dry
                        0
                        0
         Prediction
         Length: 3002, dtype: int64
In [5]: df.dropna(inplace=True)
        df.drop(['Email No.'], axis=1, inplace=True)
        x = df.drop(['Prediction'], axis = 1)
        y = df['Prediction']
In [6]: 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, rar
In [7]: knn = KNeighborsClassifier(n neighbors=7)
        knn.fit(x train, y train)
        y pred = knn.predict(x test)
```

```
In [8]: model = SVC(C = 1)
         # fit
         model.fit(x_train, y_train)
         # predict
         y pred = model.predict(x test)
In [9]: metrics.confusion_matrix(y_true=y_test, y_pred=y_pred)
Out[9]: array([[1091,
                          6],
                [ 90, 365]])
In [10]: print('Predictions:\n', y pred)
         print('\nKNN accuracy = \n', metrics.accuracy_score(y_test,y_pred))
         print('\nConfusion matrix:\n', metrics.confusion_matrix(y_test,y_pred))
         print('\nSVM accuracy = \n', metrics.accuracy score(y test,y pred))
        Predictions:
         [0 0 1 ... 0 0 1]
        KNN accuracy =
         0.9381443298969072
        Confusion matrix:
         [[1091
                  6]
         [ 90 365]]
        SVM accuracy =
```

0.9381443298969072