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        Div: A
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In [1]: import pandas as pd
        import numpy as np

In [18]: df=pd.read_csv('/home/student/Downloads/emails.csv')

In [19]: df.head()
```

Out[19]:

| | Email No. | the | to | ect | and | for | of | a | you | hou | ... | connevey | jay | valued | lay | infrastructure | military | allowing | ff | dry | Prediction |
|---|-----------|-----|----|-----|-----|-----|----|-----|-----|-----|-----|----------|-----|--------|-----|----------------|----------|----------|----|-----|------------|
| 0 | Email 1 | 0 | 0 | 1 | 0 | 0 | 0 | 2 | 0 | 0 | ... | 0 | 0 | 0 | 0 | | 0 | 0 | 0 | 0 | 0 |
| 1 | Email 2 | 8 | 13 | 24 | 6 | 6 | 2 | 102 | 1 | 27 | ... | 0 | 0 | 0 | 0 | | 0 | 0 | 0 | 1 | 0 |
| 2 | Email 3 | 0 | 0 | 1 | 0 | 0 | 0 | 8 | 0 | 0 | ... | 0 | 0 | 0 | 0 | | 0 | 0 | 0 | 0 | 0 |
| 3 | Email 4 | 0 | 5 | 22 | 0 | 5 | 1 | 51 | 2 | 10 | ... | 0 | 0 | 0 | 0 | | 0 | 0 | 0 | 0 | 0 |
| 4 | Email 5 | 7 | 6 | 17 | 1 | 5 | 2 | 57 | 0 | 9 | ... | 0 | 0 | 0 | 0 | | 0 | 0 | 0 | 1 | 0 |

5 rows x 3002 columns

```
In [20]: df.columns

Out[20]: 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)

In [21]: df.info

Out[21]: <bound method DataFrame.info of
0      Email 1      0      0      1      0      0      0      2      0      0      ...      0      0      0      0      0      0      0      0      0      0
1      Email 2      8     13     24      6      6      2    102      1     27      ...      0      0      0      0      0      0      0      0      0      0
2      Email 3      0      0      1      0      0      0      8      0      0      ...      0      0      0      0      0      0      0      0      0      0
3      Email 4      0      5     22      0      5      1     51      2     10      ...      0      0      0      0      0      0      0      0      0      0
4      Email 5      7      6     17      1      5      2     57      0      9      ...      0      0      0      0      0      0      0      0      0      0
...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...      ...
5167   Email 5168      2      2      2      3      0      0     32      0      0      ...      0      0      0      0      0      0      0      0      0      0
5168   Email 5169     35     27     11      2      6      5    151      4      3      ...      0      0      0      0      0      0      0      0      0      0
5169   Email 5170      0      0      1      1      0      0     11      0      0      ...      0      0      0      0      0      0      0      0      0      0
5170   Email 5171      2      7      1      0      2      1     28      2      0      ...      0      0      0      0      0      0      0      0      0      0
5171   Email 5172     22     24      5      1      6      5    148      8      2      ...      0      0      0      0      0      0      0      0      0      0

      jay  valued  lay  infrastructure  military  allowing  ff  dry  \
0      0      0      0      0      0      0      0      0      0
1      0      0      0      0      0      0      0      1      0
2      0      0      0      0      0      0      0      0      0
3      0      0      0      0      0      0      0      0      0
4      0      0      0      0      0      0      0      1      0
...      ...      ...      ...      ...      ...      ...      ...      ...
5167      0      0      0      0      0      0      0      0      0
5168      0      0      0      0      0      0      0      1      0
5169      0      0      0      0      0      0      0      0      0
5170      0      0      0      0      0      0      0      1      0
5171      0      0      0      0      0      0      0      0      0

      Prediction
0      0
1      0
2      0
3      0
4      0
...      ...
5167      0
5168      0
5169      1
5170      1
5171      0

[5172 rows x 3002 columns]>

In [22]: df.isnull().sum()
```

```
Out[22]: Email No.      0
         the           0
         to           0
         ect          0
         and          0

         military     0
         allowing     0
         ff           0
         dry          0
         Prediction   0
         Length: 3002, dtype: int64
```

```
In [23]: df.head(10)
```

```
Out[23]:
```

| | Email No. | the | to | ect | and | for | of | a | you | hou | ... | connevey | jay | valued | lay | infrastructure | military | allowing | ff | dry | Prediction |
|---|-----------|-----|----|-----|-----|-----|----|-----|-----|-----|-----|----------|-----|--------|-----|----------------|----------|----------|----|-----|------------|
| 0 | Email 1 | 0 | 0 | 1 | 0 | 0 | 0 | 2 | 0 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | Email 2 | 8 | 13 | 24 | 6 | 6 | 2 | 102 | 1 | 27 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| 2 | Email 3 | 0 | 0 | 1 | 0 | 0 | 0 | 8 | 0 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 3 | Email 4 | 0 | 5 | 22 | 0 | 5 | 1 | 51 | 2 | 10 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 4 | Email 5 | 7 | 6 | 17 | 1 | 5 | 2 | 57 | 0 | 9 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| 5 | Email 6 | 4 | 5 | 1 | 4 | 2 | 3 | 45 | 1 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 |
| 6 | Email 7 | 5 | 3 | 1 | 3 | 2 | 1 | 37 | 0 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 7 | Email 8 | 0 | 2 | 2 | 3 | 1 | 2 | 21 | 6 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 1 |
| 8 | Email 9 | 2 | 2 | 3 | 0 | 0 | 1 | 18 | 0 | 0 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 9 | Email 10 | 4 | 4 | 35 | 0 | 1 | 0 | 49 | 1 | 16 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

10 rows × 3002 columns

```
In [24]: df.drop(columns=['Email No.'],inplace=True)
         df.tail()
```

```
Out[24]:
```

| | the | to | ect | and | for | of | a | you | hou | in | ... | connevey | jay | valued | lay | infrastructure | military | allowing | ff | dry | Prediction |
|------|-----|----|-----|-----|-----|----|-----|-----|-----|----|-----|----------|-----|--------|-----|----------------|----------|----------|----|-----|------------|
| 5167 | 2 | 2 | 2 | 3 | 0 | 0 | 32 | 0 | 0 | 5 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 5168 | 35 | 27 | 11 | 2 | 6 | 5 | 151 | 4 | 3 | 23 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| 5169 | 0 | 0 | 1 | 1 | 0 | 0 | 11 | 0 | 0 | 1 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 |
| 5170 | 2 | 7 | 1 | 0 | 2 | 1 | 28 | 2 | 0 | 8 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 1 |
| 5171 | 22 | 24 | 5 | 1 | 6 | 5 | 148 | 8 | 2 | 23 | ... | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |

5 rows × 3001 columns

```
In [25]: df.isnull().any().value_counts()
```

```
Out[25]: False      3001
         dtype: int64
```

```
In [26]: #Separate Features And Labels
```

```
x=df.iloc[:, :df.shape[1]-1]
y=df.iloc[:,-1]
x.shape,y.shape
```

```
Out[26]: ((5172, 3000), (5172,))
```

```
In [27]: #Train
```

```
In [28]: from sklearn.model_selection import train_test_split
```

```
In [29]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.15)
```

```
In [30]: #Ml Models
```

```
In [31]: from sklearn.linear_model import LogisticRegression
         from sklearn.svm import SVC,LinearSVC
         from sklearn.neural_network import MLPClassifier
```

```
In [32]: models={"Logistic Regression":LogisticRegression(solver='lbfgs',max_iter=2000),
               "Linear SVM":LinearSVC(max_iter=3000),
               "Polynomial SVM":SVC(kernel='poly',degree = 2),
               "RBF SVM":SVC(kernel='rbf'),
               "Sigmoid SVM":SVC(kernel='sigmoid'),
               "Multi-layer Perception Classification":MLPClassifier(hidden_layer_sizes=[20,20])
               }
```

```
In [33]: #Predict Accuracy
```

```
In [34]: from sklearn.metrics import accuracy_score
```

```
In [35]: for model_name,model in models.items():
           y_pred=model.fit(x_train,y_train).predict(x_test)
           print(f"Accuracy for {model_name} model is : {accuracy_score(y_test,y_pred)}")
```

```
Accuracy for Logistic Regression model is : 0.9652061855670103
```

```
/home/student/anaconda3/lib/python3.10/site-packages/sklearn/svm/_base.py:1244: ConvergenceWarning: Liblinear failed to converge, increase the number of iterations.
```

```
    warnings.warn(
```

```
Accuracy for Linear SVM model is : 0.9458762886597938
```

```
Accuracy for Polynomial SVM model is : 0.7667525773195877
```

```
Accuracy for RBF SVM model is : 0.8195876288659794
```

```
Accuracy for Sigmoid SVM model is : 0.6494845360824743
```

```
Accuracy for Multi-layer Perception Classification model is : 0.9742268041237113
```

```
In [ ]:
```