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```
In [16]:
          import pandas as pd
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
          from sklearn.model_selection import train_test_split
          from sklearn.svm import SVC, LinearSVC
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn import metrics
          from sklearn import preprocessing
          df = pd.read csv("emails.csv")
In [17]:
          df.info()
In [18]:
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5172 entries, 0 to 5171
          Columns: 3002 entries, Email No. to Prediction
          dtypes: int64(3001), object(1)
          memory usage: 118.5+ MB
In [19]:
          df.head()
Out[19]:
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         5 rows × 3002 columns
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          Prediction
                          int64
          Length: 3002, dtype: object
```

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```
df.drop(columns=['Email No.'], inplace=True)
In [21]:
          df.isna().sum()
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          Length: 3001, dtype: int64
          df.describe()
In [23]:
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                   11.745009
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            max
         8 rows × 3001 columns
In [24]:
          X=df.iloc[:, :df.shape[1]-1] #Independent Variables
          y=df.iloc[:, -1] #Dependent Variable
          X.shape, y.shape
          ((5172, 3000), (5172,))
Out[24]:
In [25]:
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.15, random_state
In [26]:
          models = {
          "K-Nearest Neighbors": KNeighborsClassifier(n neighbors=2),
          "Linear SVM":LinearSVC(random_state=8, max_iter=900000),
          "Polynomial SVM":SVC(kernel="poly", degree=2, random_state=8),
          "RBF SVM":SVC(kernel="rbf", random_state=8),
          "Sigmoid SVM":SVC(kernel="sigmoid", random_state=8)
```

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 \t: {metrics.accuracy_score(y_test, y_pred

In []:

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C:\ProgramData\Anaconda3\lib\site-packages\sklearn\neighbors_classification.py:228:
FutureWarning: Unlike other reduction functions (e.g. `skew`, `kurtosis`), the defaul
t behavior of `mode` typically preserves the axis it acts along. In SciPy 1.11.0, thi
s behavior will change: the default value of `keepdims` will become False, the `axis`
over which the statistic is taken will be eliminated, and the value None will no long
er be accepted. Set `keepdims` to True or False to avoid this warning.
 mode, _ = stats.mode(_y[neigh_ind, k], axis=1)

Accuracy for K-Nearest Neighbors model : 0.8672680412371134