```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
df = pd.read_csv("/content/emails.csv")
df.head(2)
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                                                                                                                        2.024362
      mean
       std
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                             9.534576
                                         14.101142
                                                       6.045970
                                                                    4.680522
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      max
     8 rows × 3001 columns
     1
```

Creating the NB Model

```
X = df.iloc[:,1:3001]
X
```

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2	0	0	1	0	0	0	8	0	0	4		0	0	0	0	0	0	0
3	0	5	22	0	5	1	51	2	10	1		0	0	0	0	0	0	0
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Naive Bayes

```
mnb = MultinomialNB(alpha=1.9)  # alpha by default is 1. alpha must always be > 0.
# alpha is the '1' in the formula for Laplace Smoothing (P(words))
mnb.fit(train_x,train_y)
y_pred1 = mnb.predict(test_x)
print("Accuracy Score for Naive Bayes : ", accuracy_score(y_pred1,test_y))
```

Accuracy Score for Naive Bayes : 0.9365815931941222

Support Vector Machines

Support Vector Machine is the most sought after algorithm for classic classification problems. SVMs work on the algorithm of Maximal Margin, i.e, to find the maximum margin or threshold between the support vectors of the two classes (in binary classification). The most effective Support vector machines are the soft maximal margin classifier, that allows one misclassification, i.e, the model starts with low bias(slightly poor performance) to ensure low variance later.

Let us see the model performance.

```
svc = SVC(C=1.0,kernel='rbf',gamma='auto')
svc.fit(train_x,train_y)
y_pred2 = svc.predict(test_x)
print("Accuracy Score for SVC : ", accuracy_score(y_pred2,test_y))

Accuracy Score for SVC : 0.888631090487239
```

As expected, SVM's performance is slightly poorer than Multinomia Naive Bayes

Random Forests (Bagging)

Ensemble methods turn any feeble model into a highly powerful one. Let us see if ensemble model can perform better than Naive Bayes

```
rfc = RandomForestClassifier(n_estimators=100,criterion='gini')
# n_estimators = No. of trees in the forest
# criterion = basis of making the decision tree split, either on gini impurity('gini'), or on infromation gain('entropy')
rfc.fit(train_x,train_y)
y_pred3 = rfc.predict(test_x)
print("Accuracy Score of Random Forest Classifier : ", accuracy_score(y_pred3,test_y))
```

Accuracy Score of Random Forest Classifier: 0.9682907965970611

Ending notes:

This was a purely comparative study to check the workability of the dataset that I created, and to check how conventional models perform on my dataset. In my next kernel, I will show the code behind the extraction of this dataset from the raw text files.

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