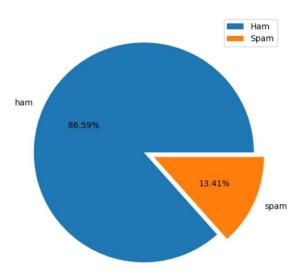
	Class	Text	numClass	Count
0	ham	Go until jurong point, crazy Available only	0	111
1	ham	Ok lar Joking wif υ oni	0	29
3	ham	U dun say so early hor U c already then say	0	49
4	ham	Nah I don't think he goes to usf, he lives aro	0	61
6	ham	Even my brother is not like to speak with me	0	77

print(ham)



	Class	Text	 numClass	Count
2	spam	Free entry in 2 a wkly comp to win FA Cup fina	1	155
5	spam	FreeMsg Hey there darling it's been 3 week's n	1	148
8	spam	WINNER!! As a valued network customer you have	1	158
9	spam	Had your mobile 11 months or more? U R entitle	1	154
11	spam	SIX chances to win CASH! From 100 to 20,000 po	1	136

print(spam)

vectorizer.fit_transform(data_read.Text)

```
(5570, 4161) 1
                     (0, 8030) 1
(5570, 903) 1
                     (0, 4350) 1
(5570, 1546) 1
                     (0, 5920) 1
(5571, 7756) 1
                     (0, 2327) 1
(5571, 5244) 1
                     (0, 1303) 1
(5571, 4225) 2
                     (0, 5537) 1
(5571, 7885) 1
                     (0, 4087) 1
(5571, 6505) 1
                     (0, 1751) 1
                     (0, 3634) 1
```

data read.numClass

5567	1	0	0	
5568	0	1	0	
5569	0	2	1	
5570	0	3	0	
5571	0	4	0	
	·			

Adaboost:

Accuracy in %: 98.08612440191388

F1 Score:

0.9130434782608695

KNN1:

Accuracy in %: 94.67703349282297

KNN3:

Accuracy in %: 92.16507177033493

F1 Score:

0.7588075880758809

F1 Score:

0.5969230769230769

KNN5:

Accuracy in %: 91.02870813397129

KNN7:

Accuracy in %: 90.19138755980862

F1 Score:

0.5098039215686275

F1 Score:

0.4383561643835616

SVM:

Accuracy in %: 98.20574162679426

F1 Score:

0.9308755760368664

tfidf_transformer = TfidfTransformer().fit(x)
dummy_transformed = tfidf_transformer.transform(x)
print(dummy_transformed)

```
      (5570, 1546)
      0.3402048888248921

      (5570, 1438)
      0.1429585509124154

      (5570, 1084)
      0.11225268140936365

      (5570, 903)
      0.3247623397615813

      (5571, 7885)
      0.42752913176432156

      (5571, 7756)
      0.14849350328973984

      (5571, 6505)
      0.5565029307246045

      (5571, 5244)
      0.39009002726386227

      (5571, 4225)
      0.5773238083586979
```

```
(0, 8489) 0.22080132794235655
(0, 8267) 0.18238655630689804
(0, 8030) 0.22998520738984352
(0, 7645) 0.15566431601878158
(0, 5920) 0.2553151503985779
(0, 5537) 0.15618023117358304
(0, 4476) 0.2757654045621182
(0, 4350) 0.3264252905795869
```

Now, lets check IDF for *you*, the most frequently repeated word in the message against *hey*, a least repeated word

```
you: 2.2548286210328206
hey: 4.907189916274442
```

As you can see, words with lower frequency are weighed higher than words with higher frequency in the dataset.

```
Multi-NB:

Accuracy in %:
98.74401913875597

F1 Score:
0.952808988764045
```

```
DecisionTreeClassifier:

Accuracy in %:
96.88995215311004

F1 Score:
0.8864628820960699
```

```
regular_MultinomialNB:

Accuracy in %:
97.54784688995215

F1 Score:
0.9154639175257732
```

```
Top 10 Spam words are :
call 346
free
       217
txt
       156
ur
       144
       144
mobile
       123
text
       121
       114
stop
claim
       113
reply
        104
```

```
Top 10 Ham words are :
u
        974
gt
        318
1t
        316
get
        301
        246
go
ok
       246
got
       242
ur
        237
know
        234
like
        231
```

```
Testing specific messages:

SMS1 = '[URGENT!] Your Mobile No 398174814449 was awarded a vacation'

SMS2 = 'Hello my friend, how are you?'

SMS1 is spam .. SMS2 is ham
```

```
please write a new sentence using words from the top spam words or regular words:

stop free

SMS1 is spam .. SMS2 is ham .. new sentence is spam
```