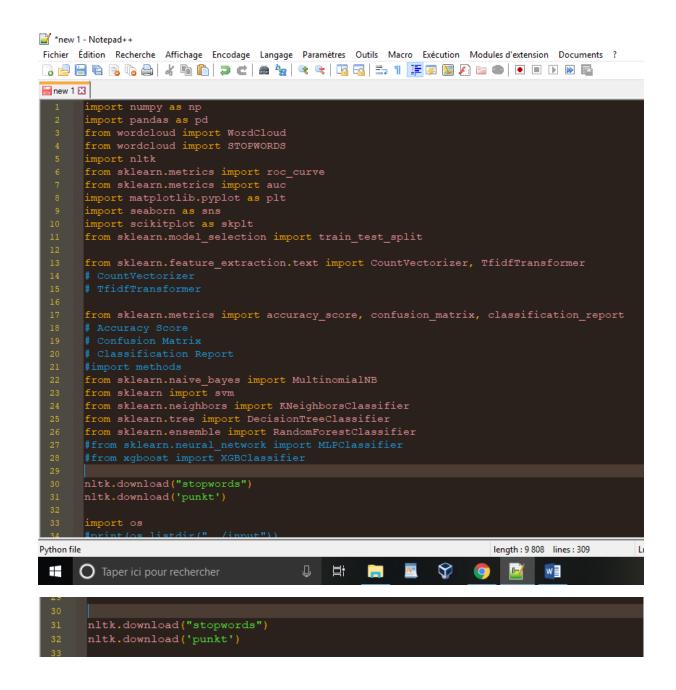
Spam Project of Machine Learning:

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```
import os

print(os.listdir("../input"))

36
```

Read Data:

```
# ###### Read Data:

df = pd.read_csv("/Users/elizabethlorelei/Downloads/spam.csv", encoding = 'latin-1')

42
```

Show Data:

	v1	v2	Unnamed: 2	Unnamed: 3	Unnamed: 4
0	ham	Go until jurong point, crazy Available only	NaN	NaN	NaN
1	ham	Ok lar Joking wif u oni	NaN	NaN	NaN
2	spam	Free entry in 2 a wkly comp to win FA Cup fina	NaN	NaN	NaN
3	ham	U dun say so early hor U c already then say	NaN	NaN	NaN
4	ham	Nah I don't think he goes to usf, he lives aro	NaN	NaN	NaN
5	spam	FreeMsg Hey there darling it's been 3 week's n	NaN	NaN	NaN
6	ham	Even my brother is not like to speak with me	NaN	NaN	NaN
7	ham	As per your request 'Melle Melle (Oru Minnamin	NaN	NaN	NaN
8	spam	WINNER!! As a valued network customer you have	NaN	NaN	NaN
9	spam	Had your mobile 11 months or more? U R entitle	NaN	NaN	NaN
10	ham	I'm gonna be home soon and i don't want to tal	NaN	NaN	NaN

```
df = df.drop(columns=['Unnamed: 2','Unnamed: 3','Unnamed: 4'])
df.columns = ['Label', 'Message']
df.head()
```

	Label	Message
0	ham	Go until jurong point, crazy Available only
1	ham	Ok lar Joking wif u oni
2	spam	Free entry in 2 a wkly comp to win FA Cup fina
3	ham	U dun say so early hor U c already then say
4	ham	Nah I don't think he goes to usf, he lives aro

Splitting the labels and the data separately:

```
0
       ham
1
      ham
2
      spam
3
      ham
4
      ham
5
      spam
6
      ham
7
      ham
8
      spam
9
      spam
      ham
Name: Label, dtype: object
```

Data Visualization:

- To check the most used word in Ham sms and Spam SMS
- To visualize the percentage of Ham and Spam SMS

```
⊟ new 1 🗵
       # Stopwords (Шумовые слова)
       stopwords = STOPWORDS
       stopwords = list(stopwords)
       STOPWORDS = nltk.corpus.stopwords.words('english')
       stopwords = stopwords + STOPWORDS
      ham dataset = df[df.Label == 'ham']
      spam_dataset = df[df.Label == 'spam']
ham_words = ' '
      spam words = ' '
     for words in ham dataset.Message:
118 =
           for word in tokens:
           ham words = ham words + word + " "
     for words in spam dataset.Message:
     spam_words = spam_words + word + " "
     def gen wordcloud(wordcloud):
   plt.figure(figsize = (10,8))
           plt.tight layout(pad=0)
           plt.axis('off')
      print("\n")
       print("\t\t\t\t HAM WORDS")
     ■wordcloud = WordCloud(background_color = 'white', width = 500, height = 500, stopwords = stopwords,
                     max_words = 500, max_font_size = 50, random_state = 42).generate(ham_words)
      gen wordcloud(wordcloud)
       print("\t\t\t\t SPAM WORDS")
     wordcloud = WordCloud(background color = 'white', width = 500, height = 500, stopwords = stopwords,
                           max words = 500, max font size = 50, random state = 42).generate(spam words)
       gen wordcloud(wordcloud)
```

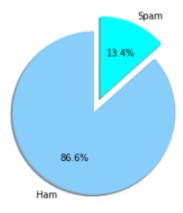
Python file length: 9 802 lines: 308 Ln: 126 C



SPAM WORDS



Plotting ham and spam data % in pie chart:



Splitting the Test and Train Data:

The Trainset consists of 3733 records and 2 features
The Testset consists of 1839 records and 2 features

Extracting N-grams from the Text Data:

```
176
177  # ##### Extracting N-grams from the Text Data:
178
179  countvect = CountVectorizer(ngram_range = (2,2), )
180   x_counts = countvect.fit(train_set.Message)
181
182  # preparing for training set
183   x_train_df = countvect.transform(train_set.Message)
184
185  # preparing for test set
186   x_test_df = countvect.transform(test_set.Message)
187
```

Data Model:

The Algorithms used below are:

- Naive Bayes
- K-Nearest
- Decision Tree
- Support Vector Machine
- Random Forest

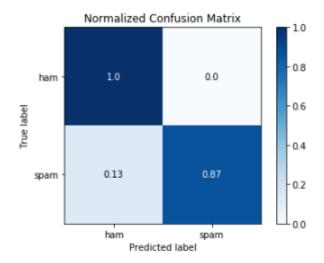
Naive Bayes classifier:

```
###### Naive Bayes classifier :

192
193
194
    clf = MultinomialNB()
195
    clf.fit(x_train_df,train_set.Label)
196
    predicted_values_NB = clf.predict(x_test_df)
197
    predictions = dict()
198
    accuracy = accuracy_score(test_set.Label, predicted_values_NB)
199
    predictions['Naive Bayes'] = accuracy * 100
200
    confusionmatrix = confusion_matrix(test_set.Label, predicted_values_NB)
201
    print("The accuracy of Naive Bayes clasifier is {}*".format(accuracy * 100))
202
    print("\n", confusionmatrix)
203
    skplt.metrics.plot_confusion_matrix(test_set.Label, predicted_values_NB, normalize = True)
204
    plt.show()
```

```
The accuracy of Naive Bayes clasifier is 97.87928221859707%

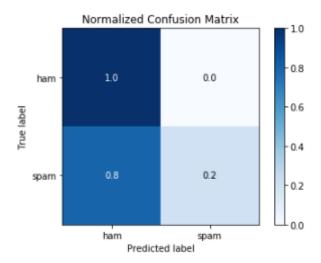
[[1581 6]
[33 219]]
```



K-Nearest Neighbors algorithm:

```
#KNN = KNeighborsClassifier(metric = 'euclidean')
kNN = KNeighborsClassifier()
kNN.fit(x_train_df, train_set.Label)
predicted_values_KNN = KNN.predict(x_test_df)
print(predicted_values_KNN)
accuracy_KNN = accuracy_score(test_set.Label, predicted_values_KNN)
predictions['K-Nearest Neighbors algorithm'] = accuracy_KNN * 100
print("\nThe accuracy of K-Nearest Neighbors algorithm is {}%".format(accuracy_KNN * 100))
confusion_matrix_KNN = confusion_matrix(test_set.Label, predicted_values_KNN)
print("\n", confusion_matrix_KNN)
skplt.metrics.plot_confusion_matrix(test_set.Label, predicted_values_KNN, normalize = True)
plt.show()
```

```
['ham' 'ham' 'ham' ... 'ham' 'ham' 'ham']
The accuracy of K-Nearest Neighbors algorithm is 89.07014681892332%
[[1587     0]
[201     51]]
```



Decision Tree learning:

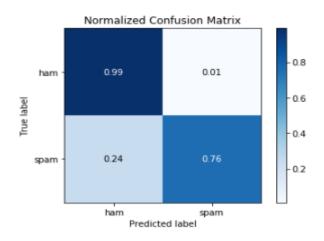
```
#### Decision Tree learning :

223

224   DT = DecisionTreeClassifier()
225   DT.fit(x_train_df, train_set.Label)
226   predicted_values_DT = DT.predict(x_test_df)
227   print(predicted_values_DT)
228   accuracy_DT = accuracy_score(test_set.Label, predicted_values_DT)
229   predictions['Decision Tree learning'] = accuracy_DT * 100
230   print("\nThe accuracy of Decision Tree learning is {}%".format(accuracy_DT * 100))
231   confusion_matrix_DT = confusion_matrix(test_set.Label, predicted_values_DT)
232   print("\n", confusion_matrix_DT)
233   skplt.metrics.plot_confusion_matrix(test_set.Label, predicted_values_DT, normalize = True)
234   plt.show()
235
236
```

```
['ham' 'ham' 'spam' ... 'ham' 'ham' 'spam'] The accuracy of Decision Tree learning is 96.08482871125612%
```

[[1576 11] [61 191]]



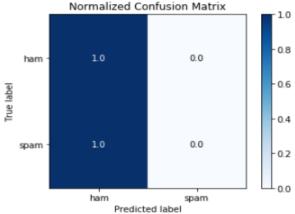
Support Vector Machine (SVM):

```
['ham' 'ham' 'ham' ... 'ham' 'ham' 'ham']

The accuracy of Support Vector Machine (SVM) is 86.2969004893964%

[[1587 0]
[252 0]]

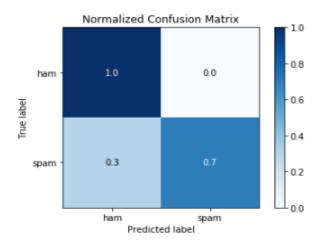
Normalized Confusion Matrix
```



Random Forest:

['ham' 'ham' 'ham' 'ham' 'ham' 'ham'] The accuracy of Random Forest is 95.86731919521479%

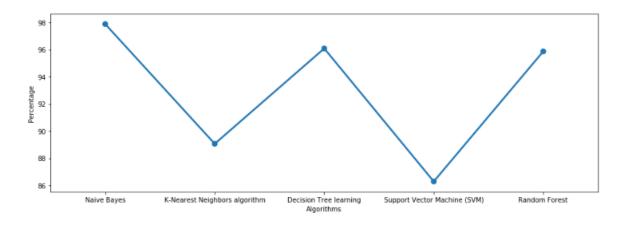
[[1587 0] [76 176]]



Méthodes Comparison:

```
271
272  # ##### Method Comparison:
273
274  fig, (axl) = plt.subplots(ncols = 1, sharey = True,figsize = (15,5))
275  df = pd.DataFrame(list(predictions.items()),columns = ['Algorithms','Percentage'])
276  display(df)
277  sns.pointplot(x = "Algorithms", y = "Percentage", data = df,ax = axl);
278
```

	Algorithms	Percentage
0	Naive Bayes	97.879282
1	K-Nearest Neighbors algorithm	89.070147
2	Decision Tree learning	96.084829
3	Support Vector Machine (SVM)	86.296900
4	Random Forest	95.867319



ROC Accuracy:

```
#pr, tpr, thresholds = roc_curve(testset.vl,predicted_values_XGB, pos_label=2)

test_prediction = test_set.Label.tolist()

predicted_values = predicted_values_NB.tolist()

test_prediction = [l if pred=="spam" else 0 for pred in test_prediction]

predicted_values = [l if pred=="spam" else 0 for pred in predicted_values]

fpr, tpr, thresholds = roc_curve(test_prediction,predicted_values)

roc_auc = auc(fpr, tpr)

print("The ROC Accuracy is {}".format(roc_auc))
```

The ROC Accuracy is 0.9326334503555676

```
plt.title('Receiver Operating Characteristic')

plt.plot(fpr, tpr, 'b',

label='AUC = %0.2f'% roc_auc)

plt.legend(loc='lower right')

plt.plot([0,1],[0,1],'r--')

plt.xlim([-0.1,1.2])

plt.ylim([-0.1,1.2])

plt.ylabel('True Positive Rate')

plt.xlabel('False Positive Rate')

plt.show()

306

307
```

