Appendices

Appendix 1: Dataset at a glance

[3]:	label		subject	email_to	email_from	message	
	0	1	Generic Cialis, branded quality@	the00@speedy.uwaterloo.ca	"Tomas Jacobs" <rickyames@aol.com></rickyames@aol.com>	Content-Type: text/html;\nContent-Transfer-Enc	
	1	0	Typo in /debian/README	debian-mirrors@lists.debian.org	Yan Morin <yan.morin@savoirfairelinux.com></yan.morin@savoirfairelinux.com>	Hi, i've just updated from the gulus and I che	
	2	1	authentic viagra	<the00@plg.uwaterloo.ca></the00@plg.uwaterloo.ca>	"Sheila Crenshaw" <7stocknews@tractionmarketin	Content-Type: text/plain;\n\tcharset="iso-8859	
	3	1	Nice talking with ya	opt4@speedy.uwaterloo.ca	"Stormy Dempsey" <vqucsmdfgvsg@ruraltek.com></vqucsmdfgvsg@ruraltek.com>	Hey Billy, $\n\$ was really fun going out the	
	4	1	or trembling; stomach cramps; trouble in sleep	ktwarwic@speedy.uwaterloo.ca	"Christi T. Jernigan" <dcube@totalink.net></dcube@totalink.net>	Content-Type: multipart/alternative; \n	
	75414	1	the reply for your request for a job place [le	"Gnitpick" <gnitpick@flax9.uwaterloo.ca></gnitpick@flax9.uwaterloo.ca>	"Sydney Car Centre" <merrill8783@168city.com></merrill8783@168city.com>	${\tt Content-Type: text/html; \nContent-Transfer-Enc}$	
	75415	1	the reply for your request for a job place [le	"Gnitpick" <gnitpick@flax9.uwaterloo.ca></gnitpick@flax9.uwaterloo.ca>	"Sydney Car Centre" <merrill8783@168city.com></merrill8783@168city.com>	${\tt Content-Type: text/html; \nContent-Transfer-Enc}$	
	75416	0	$\mbox{Re:} \ [\mbox{R}]$ Me again, about the horrible documenta	Duncan Murdoch <murdoch@stats.uwo.ca></murdoch@stats.uwo.ca>	Philippe Grosjean <phgrosjean@sciviews.org></phgrosjean@sciviews.org>	For those who are interested, I just cook a li	
	75417	0	Re: [R] RODBC problem	<r-help@stat.math.ethz.ch></r-help@stat.math.ethz.ch>	$=?iso-8859-1?Q?Bernhard_Wellh=F6fer?=\n\t$	Hello,\n\nas I wrote I call\n\n sqlFetch(chan	
	75418	1	I wanted the desk at his own laws; of the. Bu	the00@plg.uwaterloo.ca	"Danny" <pwcusnt@noblecoffee.com></pwcusnt@noblecoffee.com>	Content-Type: multipart/alternative;\n\tbounda	

Figure 1. Dataset

Appendix 2:

75419 rows × 5 columns

```
[6]: # checking null values in text
spam_dataset['message'].isna().sum()
```

Figure 2. NA Message Entries in Data

Appendix 3:

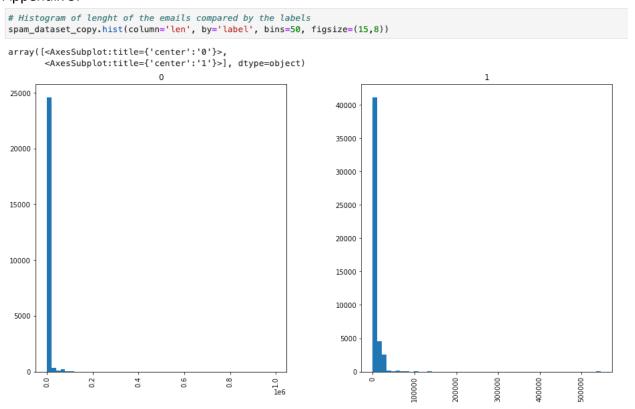


Fig 3. Histogram of Length of Emails by labels

Appendix 4:

```
[11]: # Histogram of lengths of the emails
spam_dataset_copy['len'].plot(bins=100, kind='hist',figsize=(10,7))
```

[11]: <AxesSubplot:ylabel='Frequency'>

0

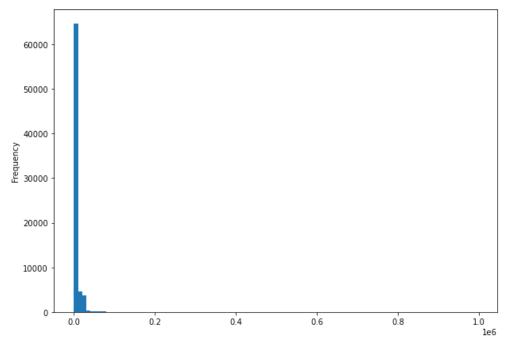


Fig 4. Histogram of email lengths

Appendix 5:

```
#Plotting value counts of SPAM and HAM
spam_dataset_copy['label'].value_counts().plot(kind='bar')
plt.xlabel('label')
plt.ylabel('Count')
plt.show()
```

Fig 5. Label Count Bar Plot

label

Appendix 6:

Fig 6. Data Preparation: TFIDF Transformation

Appendix 7:



Fig 7. Naïve Bayes Confusion Matrix

Appendix 8:

	precision	recall	f1-score	support
0 1	0.99 0.98	0.97 1.00	0.98 0.99	5056 9731
accuracy macro avg weighted avg	0.99 0.99	0.98 0.99	0.99 0.98 0.99	14787 14787 14787

Fig 8. Naïve Bayes Classification Report

Appendix 9:

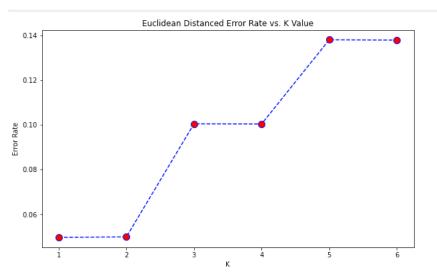


Fig 9. KNN Euclidean Error Rate Plot

Appendix 10:

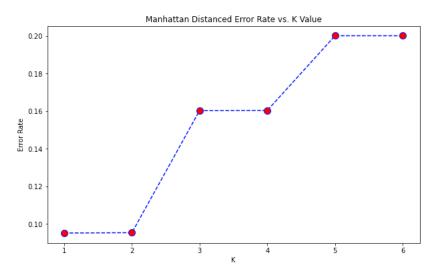


Fig 10. KNN Manhattan Error Rate Plot

Appendix 11:



Fig 11. KNN Confusion Matrix

Appendix 12:

	precision	recall	f1-score	support
0	1.00	0.86	0.92	5056
1	0.93	1.00	0.96	9731
accuracy			0.95	14787
macro avg	0.96	0.93	0.94	14787
weighted avg	0.95	0.95	0.95	14787

Fig 12. KNN Classification Report

Appendix 13:

best_n_estimator: 200

best_accuracy_score: 0.9940488266720768

best_criterion: gini

Fig 13. Random Forest Parameter Optimization Results

Appendix 14:



Fig 14. Random Forest Confusion Matrix

Appendix 15:

	precision	recall	f1-score	support
0 1	0.99 1.00	0.99 1.00	0.99 1.00	5056 9731
accuracy macro avg weighted avg	0.99 0.99	0.99 0.99	0.99 0.99 0.99	14787 14787 14787

Fig 15. Random Forest Classification Report

Code Segments:

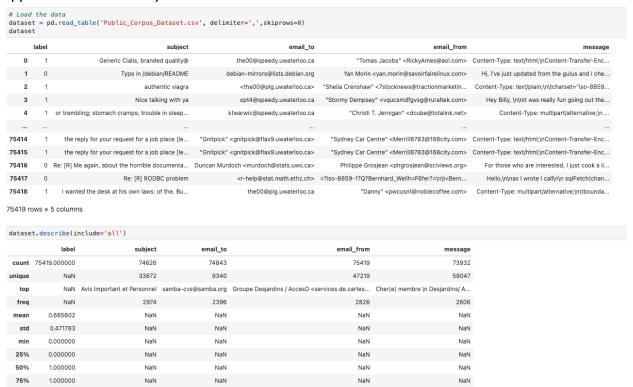
1.000000

NaN

Appendix 16: Complete Code

Complete Code is submitted in separate files both as a ipynb and html formats.

Appendix 17: Data Analysis



NaN

	la	abel	subject	email_to	email_from	message
count	75419.000	000	74626	74843	75419	73932
nique	1	NaN	33672	9340	47219	59047
top		NaN Avis Impo	ortant et Personnel san	ba-cvs@samba.org	Groupe Desjardins / AccesD <services.de.cartes< td=""><td>Cher(e) membre \n Desjardins/ A</td></services.de.cartes<>	Cher(e) membre \n Desjardins/ A
freq	1	NaN	2974	2396	2826	2606
mean	0.665		NaN	NaN	NaN	NaN
std	0.471		NaN	NaN	NaN	NaN
min	0.000		NaN	NaN	NaN	NaN
25%	0.000		NaN	NaN	NaN	NaN
50%	1.000		NaN	NaN	NaN	NaN
75%	1.000		NaN	NaN	NaN	NaN
spam	1.000	000	NaN	NaN	NaN	NaN
pam_da		dataset.lo	c[:,['label','mess	age']]		
	label		me	ssage		
0	1 Con	tent-Type: text	/html;\nContent-Transfe	-Enc		
1	0 H	li, i've just upd	ated from the gulus and	che		
2	1 Cor	ntent-Type: tex	t/plain;\n\tcharset="iso-	8859		
3	1	Hey Billy, \n\ni	it was really fun going ou	t the		
4	1	Content-	Type: multipart/alternati	e;\n		
5414	1 Con	tent-Type: text	/html;\nContent-Transfe	-Enc		
5415	1 Con	tent-Type: text	/html;\nContent-Transfe	-Enc		
5416	0	For those who	are interested, I just coo	k a li		
5417	0	Hello,\n\nas I	wrote I call\n\n sqlFetch(chan		
5418	1 Co	ntent-Type: mu	ultipart/alternative;\n\tbo	unda		
110 ro	ws × 2 co	Lumana				
	ng null valu aset['messag	<i>ies in text</i> g <mark>e'].isna().</mark> su	m()			
187						
ere are	e 1487 rows	of messages t	hat have a NA value. I	will extract these	rows from the dataset since the number is low and	in addition, these rows won't add
		tion purpose.				
		NA message r	ows age'],inplace=True)			
		ge'].isna().su				
ext_data	ng the datas a = spam_dat a.columns =	taset.loc[:,['	message']]			
ext_data	1		Text			
0 Cor	ntent-Type: tex	t/html;\nContent-Tr				
1	Hi, i've just upo	dated from the gulu	is and I che			
		xt/plain;\n\tcharset=				
3		nit was really fun go -Type: multipart/alte				
4	Content	- rype: multipart/ait	emative, (ii			
	ntent-Type: tex	t/html;\nContent-Tr	ansfer-Enc			
 i 414 Cor						
414 Cor	ntent-Type: tex	t/html;\nContent-Tr	ansfer-Enc			
414 Cor	For those who	t/html;\nContent-Tr o are interested, I ju wrote I call\n\n sqll	ıst cook a li			

Can Dai - DSC 478: Programming Machine Learning Applications - Final Project Report Appendices

```
# Label 1 indicates SPAM, Label 0 indicates HAM
labels_data = spam_dataset.loc[:,['label']]
labels_data.columns = ['Labels']
labels_data
       Labels
    0
1 0
    2
3 1
...
75414
75415
           0
 75416
 75417
           0
75418
73932 rows x 1 columns
```

### Visuals to understand	the data better
<pre>spam_dataset_copy = spam_d spam_dataset_copy['len'] = spam_dataset_copy.head()</pre>	<pre>lataset : spam_dataset_copy['message'].apply(len)</pre>

	label	message	len
0	1	Content-Type: text/html;\nContent-Transfer-Enc	225
1	0	Hi, i've just updated from the gulus and I che	728
2	1	Content-Type: text/plain;\n\tcharset="iso-8859	410
3	1	Hey Billy, \n\nit was really fun going out the	649
4	1	Content-Type: multipart/alternative:\n	17480

```
# Split Data to Train and Test
text_train, text_test, label_train, label_test = train_test_split(spam_dataset['message'], spam_dataset['label'], test_size=0.2, random_state=111)
# resetting the index numbers of the datasets
text_train = text_train.reset_index(drop=True)
text_test = text_test.reset_index(drop=True)
label_train = label_train.reset_index(drop=True)
label_test = label_test.reset_index(drop=True)
text_train.head()
       social day fly thinking whos spot journey, end...
     Content-Type: multipart/alternative;\n\tbounda...
----BEGIN PGP SIGNED MESSAGE----\nHash: SHA1...
4 buy now 100mg x 90 pills $159.95\nhttp://thusm...
Name: message, dtype: object
label_train.head()
Name: label, dtype: int64
text_test.head()
     Dear all,\n\nI got this error message \n\n> li...
      With hundreds of models to choose from, rock b...
     Content-Type: multipart/alternative;\n\tbounda...
Content-Type: text/plain;\n charset="wi...
Cher(e) membre \n Desjardins/ A...
4 Cher(e) membre \n
Name: message, dtype: object
label_test.head()
Name: label, dtype: int64
```

Appendix 18: Naïve Bayes Classification Code Segment

```
from sklearn.naive_bayes import MultinomialNB
Naive_Bayes_model = MultinomialNB(alpha=1.5)
Naive_Bayes_model.fit(tfidf_matrix_text_train,label_train)
MultinomialNB(alpha=1.5)
predicted_labels_NB = Naive_Bayes_model.predict(tfidf_matrix_text_test)
predicted_labels_NB
array([0, 1, 1, ..., 1, 0, 0])
```

Appendix 19: KNN Code Segment

```
\# Optimizing KNN by finding the best K value using Euclidean
 from sklearn.neighbors import KNeighborsClassifier
 error_rate_euclidean = []
 for i in range(1,7):
    knn = KNeighborsClassifier(n_neighbors=i, p=2)
     # fit Train
     knn.fit(tfidf_matrix_text_train, label_train)
     predicted_label_i_knn = knn.predict(tfidf_matrix_text_test)
     # Calculate error and add to error_rate list
    error_rate_euclidean.append(np.mean(predicted_label_i_knn != label_test))
 # Plotting K vs Error Rate Plot - Euclidean Distance
 plt.figure(figsize=(10,6))
 plt.plot(range(1,7), error_rate_euclidean,color ='blue', linestyle ='dashed', marker ='o',markerfacecolor ='red', markersize=10)
 plt.title('Euclidean Distanced Error Rate vs. K Value')
 plt.xlabel('K')
 plt.ylabel('Error Rate')
 plt.show()
 # Optimizing KNN by finding the best K value using Manhattan Distance
 from sklearn.neighbors import KNeighborsClassifier
 error_rate_manhattan = []
 for i in range(1,7):
     knn = KNeighborsClassifier(n_neighbors=i, p=1)
     # fit Train
    knn.fit(tfidf_matrix_text_train, label_train)
     # predict Test
     predicted_label_i_knn = knn.predict(tfidf_matrix_text_test)
     # Calculate error and add to error_rate list
    error_rate_manhattan.append(np.mean(predicted_label_i_knn != label_test))
 # Plotting K vs Error Rate Plot - Manhattan Distance
 plt.figure(figsize=(10,6))
 plt.plot(range(1,7),error_rate_manhattan,color='blue', linestyle='dashed',
          marker='o', markerfacecolor='red', markersize=10)
 plt.title('Manhattan Distanced Error Rate vs. K Value')
 plt.xlabel('K')
 plt.ylabel('Error Rate')
 plt.show()
# Getting the best K and Evaluation
  #Fitting the KMM model
  knn_classifier = KNeighborsClassifier(n_neighbors = 2, metric = 'minkowski', p=2)
  # Train
  knn_classifier.fit(tfidf_matrix_text_train, label_train)
  # Precit
  predicted_knn_classifier=knn_classifier.predict(tfidf_matrix_text_test)
# Evaluate
  cf_KNN = confusion_matrix(label_test, predicted_knn_classifier)
  cf KNN
array([[4334, 722],
         [ 14, 9717]])
```

Appendix 20: Random Forest Code Segment

```
from sklearn.ensemble import RandomForestClassifier
rf_parameters = {"n_estimators":[100, 200, 500], "criterion" : ["gini", "entropy"]}
best n estimator = 0
best_score = 0
best_criterion = ""
for crit in rf_parameters["criterion"]:
   for n in rf_parameters["n_estimators"]:
       i+=1
        print(i)
       random_forest_model = RandomForestClassifier(n_estimators = n, random_state = 111, criterion = crit)
        random\_forest\_model.fit(tfidf\_matrix\_text\_train,\ label\_train)
        labels_predicted_randomforest = random_forest_model.predict(tfidf_matrix_text_test)
        acc_score = accuracy_score(labels_predicted_randomforest,label_test)
        if acc_score > best_score:
            best_score = acc_score
            best_n_estimator = n
            best_criterion = crit
print("best_n_estimator:",best_n_estimator)
print("best_accuracy_score:",best_score)
print("best_criterion:",best_criterion)
```

```
# Random Forest using the best parameters using above results
random_forest_model = RandomForestClassifier(n_estimators = best_n_estimator, random_state = 111, criterion = best_criterion)
# Train
random_forest_model.fit(tfidf_matrix_text_train, label_train)
# Precit labels
labels_predicted_randomforest = random_forest_model.predict(tfidf_matrix_text_test)
labels_predicted_randomforest
```

array([0, 1, 1, ..., 1, 0, 0])

Appendix 21: Modules Used

```
# Load the modules
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
import seaborn as sns
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
from sklearn.feature_extraction.text import TfidfVectorizer
```