importing libraries

```
In [26]:
         import nltk
          import pandas as pd
          import numpy as np
          import warnings
          warnings.filterwarnings("ignore")
```

Data gathering

```
In [9]:
          df=pd.read_table("SMSSpamCollection", header=None, encoding='utf-8')
           df.head()
 Out[9]:
                 0
           0
              ham
                      Go until jurong point, crazy.. Available only ...
                                       Ok lar... Joking wif u oni...
              ham
             spam
                    Free entry in 2 a wkly comp to win FA Cup fina...
              ham
                      U dun say so early hor... U c already then say...
              ham
                      Nah I don't think he goes to usf, he lives aro...
          df.info()
In [10]:
           <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 5572 entries, 0 to 5571
          Data columns (total 2 columns):
                Column Non-Null Count Dtype
                         -----
                         5572 non-null
                                           object
           1
                1
                         5572 non-null
                                           object
          dtypes: object(2)
          memory usage: 87.2+ KB
          # check class distribution
In [18]:
           classes = df[0]
           df[0].value_counts()
                   4825
Out[18]:
                    747
           spam
          Name: 0, dtype: int64
```

Pre-processing of data

Preprocessing the data is an essential step in natural language process. In the following cells, we will convert our class labels to binary values using the LabelEncoder from sklearn, replace email addresses, URLs, phone numbers, and other symbols by using regular expressions, remove stop words, and extract word stems.

```
# converting class label to binary values, 0=ham, 1=spam
In [21]:
          from sklearn.preprocessing import LabelEncoder
          encoder = LabelEncoder()
          Y=encoder.fit_transform(classes)
```

```
print(Y[:20])
         [0 0 1 0 0 1 0 0 1 1 0 1 1 0 0 1 0 0 0 1]
In [23]: # now store SMS message data
         text_messages= df[1]
         print(text messages[:15])
         0
               Go until jurong point, crazy.. Available only ...
         1
                                   Ok lar... Joking wif u oni...
         2
               Free entry in 2 a wkly comp to win FA Cup fina...
               U dun say so early hor... U c already then say...
         3
         4
               Nah I don't think he goes to usf, he lives aro...
         5
               FreeMsg Hey there darling it's been 3 week's n...
               Even my brother is not like to speak with me. ...
         6
         7
               As per your request 'Melle Melle (Oru Minnamin...
         8
               WINNER!! As a valued network customer you have...
         9
               Had your mobile 11 months or more? U R entitle...
               I'm gonna be home soon and i don't want to tal...
         10
         11
               SIX chances to win CASH! From 100 to 20,000 po...
         12
               URGENT! You have won a 1 week FREE membership ...
         13
               I've been searching for the right words to tha...
                              I HAVE A DATE ON SUNDAY WITH WILL!!
         14
         Name: 1, dtype: object
 In [ ]: # Regular Expressions
         we want to switch some things like if have email address in text messages
         # replace email address with "email"
In [30]:
          processed = text_messages.str.replace(r'^.+@[^\.].*\.[a-z]{2,}$',"emailaddress")
          # replae URL with "webaddress"
          processed = processed.str.replace(r'^http\://[a-zA-Z0-9\-\.]+\.[a-zA-Z]{2-3}(/\S*)?$',
          # Replace money symbols with 'moneysymb' (£ can by typed with ALT key + 156)
          processed = processed.str.replace(r'f|\$', 'moneysymb')
          # Replace 10 digit phone numbers (formats include paranthesis, spaces, no spaces, dash
          processed = processed.str.replace(r'^\(?[\d]{3}\)?[\s-]?[\d]{3}[\s-]?[\d]{4}$',
                                            'phonenumbr')
          # Replace numbers with 'numbr'
          processed = processed.str.replace(r'\d+(\.\d+)?', 'numbr')
          # Remove punctuation
          processed = processed.str.replace(r'[^\w\d\s]', ' ')
          # Replace whitespace between terms with a single space
          processed = processed.str.replace(r'\s+', ' ')
          # Remove Leading and trailing whitespace
          processed = processed.str.replace(r'^\s+|\s+?$', '')
In [31]:
         # change word with lower case
         processed = processed.str.lower()
         processed
```

```
go jurong point crazi avail bugi n great world...
Out[38]:
         1
                                            ok lar joke wif u oni
         2
               free entri numbr wkli comp win fa cup final tk...
                              u dun say earli hor u c alreadi say
         3
                             nah think goe usf live around though
         4
         5
                freemsg hey darl numbr week word back like fun...
         6
                    even brother like speak treat like aid patent
         7
                per request mell mell oru minnaminungint nurun...
         8
                winner valu network custom select receivea mon...
         9
                mobil numbr month u r entitl updat latest colo...
         10
                gonna home soon want talk stuff anymor tonight...
         11
                six chanc win cash numbr numbr numbr pound txt...
         12
                urgent numbr week free membership moneysymbnum...
         13
                search right word thank breather promis wont t...
         14
                                                      date sunday
         15
               xxxmobilemovieclub use credit click wap link n...
         16
                                                       oh k watch
         17
               eh u rememb numbr spell name ye v naughti make...
         18
                                       fine way u feel way gota b
         19
                england v macedonia dont miss goal team news t...
         Name: 1, dtype: object
```

Feature generating

Feature engineering is the process of using domain knowledge of the data to create features for machine learning algorithms. In this project, the words in each text message will be our features. For this purpose, it will be necessary to tokenize each word. We will use the 1500 most common words as features.

```
In [46]: from nltk.tokenize import word tokenize
         # creating bag of words
         all words = []
          for message in processed:
             words = word_tokenize(message)
             for w in words:
                  all words.append(w)
          all words= nltk.FreqDist(all words)
         # print total number of words and the 15 most common words
In [49]:
         print("Number of Words",len(all words))
          print("Most common words",all_words.most_common(20))
         Number of Words 6579
         Most common words [('numbr', 2648), ('u', 1207), ('call', 674), ('go', 456), ('get',
         451), ('ur', 391), ('gt', 318), ('lt', 316), ('come', 304), ('moneysymbnumbr', 303),
         ('ok', 293), ('free', 284), ('day', 276), ('know', 275), ('love', 266), ('like', 26
         1), ('got', 252), ('time', 252), ('good', 248), ('want', 247)]
         # lets use most common 1500 words as features
In [65]:
         word_features = list(all_words.keys())[:1500]
In [66]:
         len(word_features)
         1500
Out[66]:
```

```
In [ ]: word_features
         # The find_features function will determine which of the 1500 word features are contai
In [78]:
          def find_features(message):
              words = word tokenize(message)
              features = {}
              for word in word features:
                  features[word] = (word in words)
              return features
          # Lets see an example!
          features = find_features(processed[0])
          for key, value in features.items():
              if value == True:
                  print (key)
         go
         jurong
         point
         crazi
         avail
         bugi
         great
         world
         la
         е
         buffet
         cine
         got
         amor
         wat
         processed[0]
In [79]:
          'go jurong point crazi avail bugi n great world la e buffet cine got amor wat'
Out[79]:
          # Now lets do it for all the messages
In [82]:
          messages = zip(processed, Y)
In [84]:
         # call find features function for each SMS message
          featuresets = [(find_features(text), label) for (text, label) in messages]
         # split the featureset into trainin and testing datasets using sklearn
In [86]:
          from sklearn import model selection
          training,testing = model_selection.train_test_split(featuresets,test_size=0.25)
In [87]: print(len(training))
         4179
In [88]:
         print(len(testing))
         1393
```

Model evaluation

Now that we have our dataset, we can start building algorithms! Let's start with a simple linear support vector classifier, then expand to other algorithms. We'll need to import each algorithm we plan on using from sklearn. We also need to import some performance metrics, such as accuracy_score and classification_report.

```
from nltk.classify.scikitlearn import SklearnClassifier
         from sklearn.svm import SVC
         model = SklearnClassifier(SVC(kernel="linear"))
          #train the model
         model.train(training)
         # test on testing datasets
          accuracy=nltk.classify.accuracy(model,testing)*100
          print("SVC Accuracy",accuracy)
         SVC Accuracy 97.84637473079684
         from sklearn.neighbors import KNeighborsClassifier
In [104...
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
          from sklearn.linear model import LogisticRegression
          from sklearn.naive_bayes import MultinomialNB
          from sklearn.svm import SVC
         from sklearn.metrics import classification report,accuracy score,confusion matrix
        # define models to train
In [108...
         names = ["K Nearest Neighbors", "Decision Tree", "Random Forest", "Logistic Regressior
                   "Naive Bayes", "SVM Linear"]
          classifiers = [KNeighborsClassifier(),
             DecisionTreeClassifier(),
             RandomForestClassifier(),
             LogisticRegression(),
             MultinomialNB(),
             SVC(kernel = 'linear')]
         models=zip(names,classifiers)
          for name, model in models:
             nltk model = SklearnClassifier(model)
             nltk_model.train(training)
             accuracy = nltk.classify.accuracy(nltk model,testing)*100
              print("{} Accuracy:{}".format(name,accuracy))
         K Nearest Neighbors Accuracy:92.46231155778895
         Decision Tree Accuracy:96.91313711414213
         Random Forest Accuracy:97.63101220387652
         Logistic Regression Accuracy:97.91816223977028
         Naive Bayes Accuracy:98.20531227566404
         SVM Linear Accuracy:97.84637473079684
In [110... # make class label prediction for testing set
         txt_features, labels = zip(*testing)
         prediction = nltk model.classify many(txt features)
```

```
# print a confusion matrix and a classification report
 In [111...
           print(classification_report(labels, prediction))
           pd.DataFrame(
               confusion_matrix(labels, prediction),
               index = [['actual', 'actual'], ['ham', 'spam']],
               columns = [['predicted', 'predicted'], ['ham', 'spam']])
                         precision
                                      recall f1-score
                                                          support
                      0
                              0.98
                                         1.00
                                                   0.99
                                                             1190
                      1
                              0.98
                                         0.87
                                                   0.92
                                                              203
                                                   0.98
                                                             1393
               accuracy
                              0.98
                                         0.93
                                                   0.95
             macro avg
                                                             1393
          weighted avg
                              0.98
                                         0.98
                                                   0.98
                                                             1393
Out[111]:
                          predicted
                        ham
                             spam
           actual
                  ham 1186
                  spam
                          26
                               177
  In [ ]:
```