# Machine Learning Nanodegree Capstone Project

Spam Detection using Naive Bayes and Support Vector Machines

By Sylvester Ranjith Francis

# **Domain Background:**

- Natural language Processing: Natural language processing is a field of computer science, artificial intelligence, and computational linguistics concerned with the interactions between computers and human (natural) languages. As such, NLP is related to the area of human—computer interaction.
- Machine Learning: Machine learning is a type of artificial intelligence (AI) that
  provides computers with the ability to learn without being explicitly
  programmed. Machine learning focuses on the development of computer
  programs that can teach themselves to grow and change when exposed to
  new data.
- Support Vector Machines: Support vector machines (SVMs, also support vector networks) are supervised learning models with associated learning algorithms that analyze data used for classification and regression analysis.
- Naive Bayes: It is a classification technique based on Bayes' Theorem with an assumption of independence among predictors. In simple terms, a Naive Bayes classifier assumes that the presence of a particular feature in a class is unrelated to the presence of any other feature.
- Text Feature Extraction: the **term** frequency inverse document frequency (also called tf-idf), is a well know method to evaluate how important is a word in a document

### **Problem Statement:**

- The Proposed project involves classifying the SMS messages received in an end-user's phone as either a Spam message or not a Spam message (Ham)
- Spam messages are a nuisance which cause various problems such as inbox storage issues
- Unwanted messages lead to delay in delivery of important messages
- It also leads to loss of bandwidth in the carriers.

• The project involves removal of spam messages by classifying them and segregating them into a separate bin

## **Datasets and Inputs:**

This project utilises the <u>SMS spam dataset</u> found in the UCI Machine Learning Repository .

• The dataset looks like this:

	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
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
5	spam	FreeMsg Hey there darling it's been 3 week's n

• The dataset consists of Tab separated values

#### **Solution Statement:**

- The solution to this problem is to classify the texts as spam and ham, and create a prediction model to do the same
- The end result being the creation of the prediction model with two different approaches namely
  - 1. Using a Naive Bayes Classifier with an accuracy of 95%
  - 2. Using a Support Vector Machine with an accuracy of 98%

# An Outline of Project Design:

- The project is implemented as an IPython notebook using python 2.7,numpy, pandas,Sklearn,Matplotlib and Textblob
- The basic goal of the project is to design a Naive Bayes Classifier to classify the text messages into Spam and Ham (not spam)
- Initially a naive bayes classifier approach was utilised.
- Step 1:Initial exploration of the data was performed
- Step 2:Once the data exploration was performed,I proceeded to convert the raw messages into vectors.
- Step 3:The mapping was not 1 to 1 hence the Bag of Words
  approach is used where each unique word in a text is represented by a
  number
- Step 4: Converting the text into a vector involves three steps namely, Counting
  the frequency of a word occurring in each message (term
  frequency), Weighting the counts, so that frequent tokens get lower weight
  (inverse document frequency), Normalizing the vectors to unit length, to
  abstract from the original text length (L2 norm)
- Reference:
   <a href="http://scikit-learn.org/stable/modules/feature\_extraction.html#the-bag-of-words">http://scikit-learn.org/stable/modules/feature\_extraction.html#the-bag-of-words</a>
   -representation
- Step 5: The model has to trained using a naive bayes classifier to check if our model works
- Step 6: Split the data into testing and training data using Sklearn's train test split
- Step 7:The testing data is fed into the designed Naive bayes classifier,to find out that the prediction engine gave an accuracy of 95%
- Step 8: To improve the performance of the prediction engine,I tried designing a Support Vector Machine after many different trials and errors to improve the accuracy
- Step 9: The resulting Support Vector Machine gave an accuracy of 98% which is a huge improvement compared to the Naive bayes classifier
- The result being that we were able to classify the messages successfully as spam and ham with an accuracy of 98%

#### References:

## **Text Feature Extraction- Bag of Words Representation:**

http://scikit-learn.org/stable/modules/feature\_extraction.html#the-bag-of-words-representation

http://scikit-learn.org/stable/auto\_examples/model\_selection/grid\_search\_text\_feature extraction.html

http://scikit-learn.org/stable/modules/generated/sklearn.feature\_extraction.text.Count Vectorizer.html

http://scikit-learn.org/stable/modules/generated/sklearn.feature\_extraction.text.TfidfTransformer.html

## Naive Bayes - Sklearn Documentation:

http://scikit-learn.org/stable/modules/naive\_bayes.html

## **Support Vector Machines -Sklearn Documentation:**

http://scikit-learn.org/stable/modules/svm.html

http://scikit-learn.org/stable/modules/generated/sklearn.svm.SVC.html

#### **Pipelines-Sklearn Documentation:**

http://scikit-learn.org/stable/modules/generated/sklearn.pipeline.Pipeline.html

#### **Cross validation-Sklearn Documentation:**

http://scikit-learn.org/stable/modules/cross\_validation.html

http://scikit-learn.org/0.17/modules/generated/sklearn.grid\_search.GridSearchCV.html

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#### Model Selection-Sklearn Documentation:

http://scikit-learn.org/stable/modules/generated/sklearn.model\_selection.StratifiedKFold.html#sklearn.model\_selection.StratifiedKFold

http://scikit-learn.org/stable/auto\_examples/model\_selection/plot\_learning\_curve.html

# **Accuracy score/Metrics-Sklearn Documentation:**

http://scikit-learn.org/stable/modules/generated/sklearn.metrics.accuracy\_score.html #sklearn.metrics.accuracy\_score

## **Decision Tree Classifier-Sklearn Documentation:**

http://scikit-learn.org/stable/modules/generated/sklearn.tree.DecisionTreeClassifier.html

# Other References:

https://www.thinkful.com/projects/building-a-text-classifier-using-naive-bayes-499/ http://blog.fliptop.com/blog/2015/03/02/bias-variance-and-overfitting-machine-learning-overview/

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https://www.analyticsvidhya.com/blog/2015/10/understaing-support-vector-machine-example-code/

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http://www.ling.upenn.edu/courses/Fall\_2007/ling001/penn\_treebank\_pos.html

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https://en.wikipedia.org/wiki/Text\_normalization

https://en.wikipedia.org/wiki/Stop\_words

http://stackoverflow.com/questions/19335165/cross-validation-and-grid-search

http://sebastianraschka.com/Articles/2014 naive bayes 1.html

#### Affirmation:

I hereby confirm that this submission is my work. I have cited above the origins of any parts of the submission that were taken from Websites, books, forums, blog posts, github repositories, etc.

Sylvester Ranjith Francis