

# Effective Mining of Unstructured Twitter Data for Detecting User Persecution

A. Afiya, Shaik Javed Parvez, S. Arun

**Abstract---** *With the increase in social media platform there has been heave in user spawn data. These data has generated in remarkable amount on daily basis through all micro blogging sites and most of the data are unstructured. Social media platform like Twitter has become a major platform for people to share their daily thoughts, opinions, suggestions and also pave a way for abusing other users verbally. We propose to investigate user oppression detection on twitter. In the paper we propose data mining techniques for mining unstructured twitter data and apply deep learning concept on tweets. We also present a case revise to exemplify the effectives of proposed system. In this paper we have tried to point the opportunities of future work by providing a expansive perspective on open forum for data mining.*

**Keywords---** *Data Mining, Classification Algorithm, Unstructured Data, Deep Learning.*

## I. INTRODUCTION

Predominant increase in network connectivity has promised for wide expansion in resource sharing and various alliances. Social platforms like Facebook, YouTube and Twitter has gained much popularity among the people through Web 2.0. Most of 80% data are in unstructured form [1]. This allows people to build strong connection, share different kind of information like pictures, videos and their thoughts on social networks. It is apparent that advent of this real time application has caused people life more convenient also threatening at times due to user oppression. Abusive tweets sharing, commenting and posting cause menace to others users. Social Media source provides immense amount of data. User generates text data on daily basis through tweets and comments. Dataset taken is through Twitter API. Sample amount of twitter words taken for sampling is around 10000 words which contains all kind of tweets. On daily basis huge amount of unstructured data is generated through social mediums and other sources. Most of these data move unused. Information Retrieval (IR) has gained more insight in recent years due to availability of data in digital form and flexibility to access those data. As the name

implies it is a process of retrieving significant information from database.

Data mining subset of is a process of retrieving information from dataset and making into a more understandable format. Through data mining interesting hidden patterns can be discovered. The process of getting deeper inside the data is also termed as Data Mining. It involves following six basic classes of task like

1. Anomaly Detection : Identifying the data of interest or identifying odd data
2. Association Rule Learning: Understanding dependencies between variables
3. Clustering: Discovering groups and similar arrangements in data
4. Classification is method of used to predict output based on data analyzes.
5. Regression is process used to predict numbers.
6. Summarization is visualization of results

A single label Naïve Bayes algorithm is used for classification on online discussion data to understand the reflective thinking of teachers [2]. The methodology works by initially collecting the teachers online discussion data, data storage, data sampling and classification modeling and result analyzing. They implemented the initially 2000 post with 70 percent for training and 30 percent of 2000 for tests data. And found Naïve Bayes classifier had improved performance than SVM Classifier.

### 1.1 Classification

Classification is an important phase of data mining it is data analysis task. It predicts the data model based on given train and test set of data. Classification is a process of defining a group or class of data into one predefined set of data based on each item in dataset. The main aim of classification is predict accurately the target class for each case in data. Before classification there are few steps required to convert the raw data. For example consider a medical student want to analyze the cancer data and predict which one of the treatments must be given to patient among the treatment1, treatment2 and treatment3 is termed as classification and prediction. It is a data analysis process where a classifier is build to predict the data model. It has two steps learning and classifying.

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be always possible in that case the data is split into train and test sets. Sklearn library is imported for splitting dat. Shuffle Split allows randomly splitting test and training data.

```

File Edit View Insert Cell Kernel Help
In [11]: #data, train
len(data[data['label']==1]), len(data[data['label']==0])
Out[11]: (2502, 4311)

In [12]: from sklearn.model_selection importKFold, StratifiedKFold
X=data['tweet'].values
y=data['label'].values

skf = StratifiedKFold(n_splits=10)
skf.get_n_splits(X, y)
for train_index, test_index in skf.split(X, y):
    print("TRAIN: ", train_index, "TEST: ", test_index)
    X_train, X_test = X[train_index], X[test_index]
    y_train, y_test = y[train_index], y[test_index]
# For train_index, test_index in skf.split(X, y):
#     print("TRAIN: ", train_index, "TEST: ", test_index)
    
```

Fig. 5: Data Split

- Step 1: Split the Twitter Dataset into two different sets
- Step 2: x for independent variables and y for dependent variables.
- Step 3: Split the x variable into xtrain and xtest
- Step 4: Split the y variable in-to ytrain and ytest
- Step 5: The data is split into 80-20 ratios

III. ALGORITHM IMPLEMENTATION & RESULTS

Neural Networks are part of Machine Learning Algorithms. Most of Deep Learning uses neural network architecture but conventionally neural network has two to three hidden while deep learning networks has more than hidden layers and it is often termed as deep neural networks.

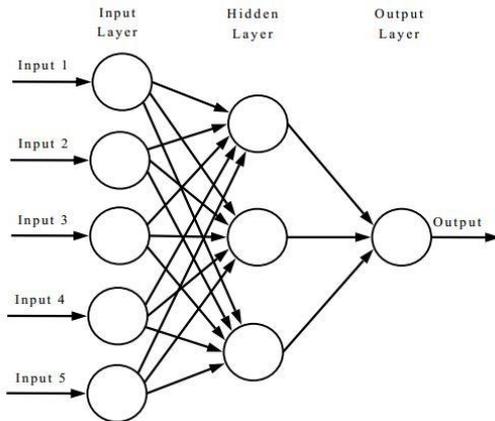


Fig. 6: Example of Neural Networks

The concept behind Deep learning networks is more data, more models leads better results. The neural networks are based on biological neurons. Example of deep learning is voice controller of mobile devices, PC's and other wireless devices.

3.1 LSTM

RNN (Recurrent Neutral Network) works by creating loop in network architecture which serves as memory state. This is modeled to remember what all learned from the beginning.

Vanishing Gradient is the problem that arises due to memory state in the model which means learning from large memory makes network more difficult to remember previous layers and tune parameters of earlier layer. LSTM (Long Short Term Memory) is type of RNN that was created to overcome the gradient issues by using tanh. LSTM is also called as cell state which updates the memory periodically.

Looping arrow is recursive and stores the previous data. There are three gates in general the forget gate modifies the cell and input modulation gate.

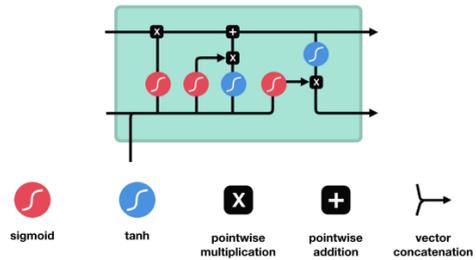


Fig. 7: LSTM Architecture (Source: Medium)

3.1.1 Forget Gate

Sigmoid can output 0 or 1 for forget or remember memory.  $x(t)$  and  $h(t-1)$  are inputs of sigmoid layers and it decides which information must be kept and deleted. For example consider a person named Peter and statement is about him "he is a good friend of lia". Now the gender of peter can be removed since the subject is about lia. The Equation of Forget Gate =  $f(t) * c(t-1)$ .

3.1.2 Input Gate

Hidden state and current input is passed into tanh (-1,1) for regulating network. The output from tanh and sigmoid are multiplied and Sigmoid decides which information must be kept and discarded.

3.1.3 Cell State

Information is collected above to built cell state. The cell state is multiplied by forget vector and drops the non required values. The output from input gate gets added to cell state. Finally the cell state has updated values.

3.1.4 Output Gate

Output Gate determines out of all values which must be forwarded to hidden layers. The new hidden state and cell is carried to next phase. Step 1: Input Layers takes sequence words as input

- Step 2: Using LSTM Units compute the output layer
- Step 3: The activation of certain neutrons are turned off to prevent over fitting
- Step 4: Compute the new best output
- Step 5: Run for five epochs

```

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In [19]: model.fit(x_shuffled, y_shuffled, batch_size=batch_size,
                epochs=10, validation_data=(x_val, y_val))
Train on 7000 samples, validate on 800 samples
Epoch 1/10
7000/7000 [#####] - loss: 0.7034 - val_loss: 4.5774 - val_acc: 0.7028
Epoch 2/10
7000/7000 [#####] - loss: 0.4822 - val_loss: 3.3216 - val_acc: 0.7041
Epoch 3/10
7000/7000 [#####] - loss: 0.7019 - val_loss: 3.8980 - val_acc: 0.7029
Epoch 4/10
7000/7000 [#####] - loss: 0.5085 - val_loss: 3.7482 - val_acc: 0.7041
Epoch 5/10
7000/7000 [#####] - loss: 0.4787 - val_loss: 3.6117 - val_acc: 0.7041
Epoch 6/10
7000/7000 [#####] - loss: 0.6204 - val_loss: 3.6129 - val_acc: 0.7041
Epoch 7/10
7000/7000 [#####] - loss: 0.7193 - val_loss: 3.6201 - val_acc: 0.7041
Epoch 8/10
7000/7000 [#####] - loss: 0.6143 - val_loss: 3.6594 - val_acc: 0.7041
Epoch 9/10
7000/7000 [#####] - loss: 0.6214 - val_loss: 3.6594 - val_acc: 0.7041
Epoch 10/10
7000/7000 [#####] - loss: 0.7174 - val_loss: 3.6449 - val_acc: 0.7041
Out[19]: <keras.callbacks.History at 0x13855126>

In [20]: model.evaluate(x_shuffled, y_shuffled)
{'loss': 0.4294, 'acc': 0.7151}

In [21]: from keras.models import Sequential, Model

In [22]: graph_in = Input(shape=(sequence_length, embedding_dim))
        LSTM = []
        for filter in filters_sizes:
            conv = Conv1D(filters_size, filter_length,
                          filter_length=filter_length,
                          border_mode='valid')
    
```

Fig. 8: Implementation of LSTM using Python

3.2 CNN

Convolutional neural network is a class of deep neural networks precisely used in image related problems or image visionary problems. Recently applied to NLP problems [9],[10] and the results are promising. There are three layers in general convolutional layer, pooling layer and classification layer.

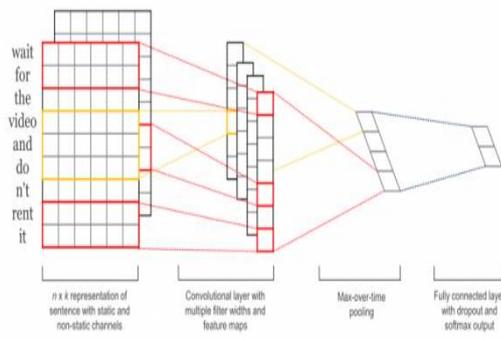


Fig. 9: Illustration of CNN (Source: Wildml)

Step 1: Input test is represented in matrix form  $n \times k$  where maximum number of words refers to  $n$  and length of embedding refers to  $k$

Step 2: Apply convolution multiple times with  $k$  as width along different height

Step 3: Apply Max pooling across the output filter and select one output from each filter ( for detecting any feature presences)

Step 4: Last is fully connected layer and output of softmax. Softmax is used for probability distribution.

```

File Edit View Insert Cell Kernel Help Python 3.0
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Epoch 5/10
7000/7000 [#####] - loss: 0.4787 - val_loss: 3.6117 - val_acc: 0.7041
Epoch 6/10
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7000/7000 [#####] - loss: 0.7174 - val_loss: 3.6449 - val_acc: 0.7041
Out[19]: <keras.callbacks.History at 0x13855126>

In [20]: model.evaluate(x_shuffled, y_shuffled)
{'loss': 0.4294, 'acc': 0.7151}

In [21]: from keras.models import Sequential, Model

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```

Fig. 10: Implementation of CNN using Python

Method	Accuracy Rate
LSTM	69.04
CNN	71.61

Fig 11: Accuracy of Classification Algorithms

IV. CONCLUSIONS

In data mining the traditional classification method learns from the data and makes decision based on the learned data. While Deep Learning classification algorithms which are detachment of ML but functioning capabilities are different and precise. LSTM and CNN create neural networks and make an intelligent decision. The main aim of classification algorithm is to generate accurate and better results. Henceforth there are different classification algorithms and the each works for different data. There are more deep algorithms for different requirement and may work well for text mining concepts.

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