

# Fake News Detection Using Traditional ML & Modern DL

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## Abstract

This report shows fake news detections using various ML approaches like Naive Bayes classifier, Logistic Regression, Decision Tree Classifier and SVM. The report also contains a roadmap to the planned baseline implementation based on few research papers and how we plan to contribute and improve it further. All the above mentioned models are trained and tested on dataset which we acquired from Kaggle. Received results that are discussed in subsequent sections suggest that fake news detection problem can be addressed with artificial intelligence methods. This is in coherence with what we have gathered from various research papers as well.

## 1 Project Introduction

Internet and social media made the access to the news information much easier and comfortable. Often Internet users can follow the events of their interest in online mode, and spread of the mobile devices make this process even easier.

Mass media have a huge influence on the society, and as it often happens, there is someone who wants to take advantage of this fact. Sometimes to achieve some goals mass-media may manipulate the information in different ways. This leads to producing of the news articles that are not completely true or even completely false. There even exist lots of websites that produce fake news almost exclusively. They deliberately publish hoaxes, propaganda and disinformation purporting to be real news—often using social media to drive web traffic and amplify their effect. The main goal of fake news websites is to affect the public opinion on certain matters (mostly political). Fake news is now a global issue as well as a global challenge. (Mykhailo Granik, 2017)

## 2 Importance Of The Project

With the increasing rate in people using social media, they are exposed to new information every day but using social media as a medium for news updates is a double-edged sword. The main advantage is social media provides free access and spread of information at an impressive rate. The main disadvantage is social media provides the ideal place for creation and spread of fake news which can become highly influential and has ability to spread at exponential rate. The problem is this misinformation is hard to correct and may cause problems. For example, on October 03, 2008 Apple's stock took a 10-point hit because of a false report surfaced on CNN's iReport that Steve Jobs had a heart attack. The fake news will lead to a lot of confusion and uncertainty because in a short time it will spread a lot. The fake news will cause an imbalance in news ecosystem, for instance "The 2016 US Presidential Election", the most popular fake news was even more widespread on Facebook. This describes how people will pay more attention to misinformation rather than authenticated facts. The fake news can change the way the people interpret and respond to real news. To mitigate the negative effects caused by fake news both to benefit the society

and new ecosystem it's important to develop methods that will automatically detect fake news on social media. (T S, 2020)

### **3 Definitions**

#### **3.1 Count Vectors**

Count Vectors: Count Vector represents a notation in the form of a matrix data set matrix notation in which corpus document is represented by each row, each column represents a corpus term, and each cell represents the frequency count of a particular term in a particular document.

#### **3.2 Bag of Words**

The Bag of Words model is simplifying the representation of text (such as sentence or text dataset or document), which is represented in bag (multiset) of its words, disregarding the grammar and even word order but keeping multiplicity. In simple words Bag of words is a simplest form of representing text in terms of numbers. This is often used in Text Classification because of its simplicity and less expensive to compute.

#### **3.3 TF-IDF**

TF-IDF is an acronym of Term Frequency-Inverse Document Frequency. TF-IDF is a numerical statistic that is intended to reflect how important a word/term is to a document in a corpus. TF-IDF is the product of the TF and IDF scores of the term. TF means the number of times the word appears in the document to the total number of words in the document. IDF means the number of times a word occurs in a corpus of documents. In simple words TF-IDF is a score to highlight each word's relevance in the entire document. So using TF model makes sense of important words in a document and using IDF model makes sense of important words throughout all documents. (George, 2020)

#### **3.4 Stemming**

Stemming is NLP technique used to normalize the word by truncating the word to its stem word. Stemming may not give us a dictionary, grammatical word for a particular set of words it only gives the stem word.

#### **3.5 Lemmatization**

Lemmatization technique is also like stemming. However, what makes it different from stemming is that lemmatization finds the dictionary word or meaningful word instead of truncating to the original word. Because of this Lemmatization is slower than stemming.

#### **3.6 Doc2vec**

Doc2vec is an unsupervised algorithm which computes a feature vector for every document in the corpus. These vectors can be used to find similarity between sentence or paragraph documents.

#### **3.7 Word2vec**

Word2vec is a combination of models used to represent distributed representations of words in a corpus. Word2vec works on the idea of Distributional semantics which means that we can understand the meaning of a word by understanding the context that a word keeps. In simple words, Word vector algorithms use the context of the words to learn numerical representations for words, so that words used in the same context have similar looking word vectors.

#### **3.8 GloVe**

Global Vectors for Word Representation or GloVe is an unsupervised learning algorithm for obtaining vector representations for words. GloVe is a count-based model, which takes the advantage of global count statistics instead of only local information. GloVe uses co-occurrence matrix to derive semantic relationships between words. (Jeffrey Pennington, 2014)





Model	Accuracy
NaiveBayes	56.60
LogisticRegression	55.20
DecisionTree	50.80
LinearSVM	54.50
RBFSVM	55.70

Table2:TF-IDF&Lemmatization

### 7.1 NaiveBayes

Naive Bayes are mostly used in natural language processing. Naive Bayes classifier algorithm is a family of algorithms which use Bayes Theorem. It uses the naive assumption that all the features are independent of each other. Bayes theorem calculates the probability  $P(c|x)$  where  $c$  is the class of possible outcomes and  $x$  is the given instance which has to be classified.

### 7.2 LogisticRegression

Logistic Regression is one of the classification algorithm that is used to predict the probability of a categorical Target variable. In logistic regression, the Target variable is a binary variable that contains data coded as 1 (yes, True) or 0 (no, fake). Logistic Regression is quite good in solving binary classifications due to its predictive power in probability values. Logistic Regression can also be used for solving multi-class classification.

### 7.3 DecisionTree

Decision Tree is one of the most widely used classifiers. It is a supervised learning algorithm. Decision tree can perform both classification and regression. Decision Tree classifiers are more popular because tree analysis is easy to understand. It divides the given data set into small parts and a decision tree is incrementally constructed. The leaf nodes of a decision tree represent the classification. Decision trees are comfortable with numeric and categorical data.

### 7.4 SupportVectorMachine

Support Vector Machine or SVM is a linear model for classification and regression problems. SVM model takes the data in the training set, and maps it to data points in spaces so that there is a clear gap between points belonging to different categories.

This gap is made as wide as possible to improve the performance of the model. Whenever a new data point is given to the model, it maps the point to the same space, and predict the category based on the side of the gap on which they fall.

## 8 Model Analysis-Deep Learning

Model	Accuracy
LSTM	98.59
BiLSTM	98.21

Table3:Word2Vec

Model	Accuracy
LSTM	95.98
BiLSTM	96.31

Table4:Glove

Model	Accuracy
BERT	95.57

Table5:KtrainBERT

### 8.1 LongShort-TermMemory

Long short-term memory is an artificial recurrentneural network architecture used in the field ofdeeplearning.

Unlikestandardfeedforwardneuralnetworks,LSTMhasfeedbackconnections.Itcannotonlyprocesssingledatapo ints,butalsoentiresequencesofdata.(Wikipedia)

### 8.2 BidirectionalRNNwithLSTMs

As a final RNN model, a bidirectional recurrent net- workwithLSTMcellsalsoimplemented.Thefinalstatesoftheforwardandbackwardlayersareconcatenated and passed through an affine trans-formbeforebeinginputtothesigmoidtogeneratetheprediction.

### 8.3 BERT

BERT stands for Bidirectional Encoder Representa-tions from Transformers. It is designed to pre- traindeepbidirectional representations from unlabeled

text by jointly conditioning on both left and rightcontext. As a result, the pre-trained BERT modelcan be fine- tuned with just one additional outputlayer to create state-of-the-art models for a widerange of NLP tasks.BERT is pre-trained on alargecorpusofunlabelledtextincludingtheentireWikipediaandBookCorpus.

## 9 Conclusion

Even though there is a bit of success in detectionof fake news using some of the Machine learningtechniques yet since with every passing day char-acteristics and definition and types of fake newsin social media networks is changing drasticallycausingachallengeinclassificationoffakenews

. But with the advent of deep learning techniquesand applications in recent past, most of researchworks is implementing deep learning methods, likeCNN,, Deep neural network and Deep auto en-coder model,LSTM, RNN , in various applications,like audio and speech processing, Natural languageprocessing and modelling, information retrieval,objective recognition and computer vision, hencepost midsem we worked particularly on modernDL techniques like LSTM ,attention models, Bi-Lstmandwesawahugeimprovementinaccuracythan our old traditional ML techniques as we cansee from the tables itself. (HenilChopada, 2020)(Bajaj,2017)

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