



Fake News Detection Project

Submitted by:
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ACKNOWLEDGMENT

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<https://www.google.com/>

<https://www.youtube.com/>

https://scikit-learn.org/stable/user_guide.html

<https://github.com/>

<https://www.analyticsvidhya.com/>

INTRODUCTION

Business Problem Framing

The project is concerned with identifying a solution that could be used to detect and filter out sites containing fake news for purpose of helping users to avoid being lured by clickbait. It is imperative that such solutions are identified as they will prove to be useful to both readers and tech companies involved in the issue.

Conceptual Background of the Domain Problem

The idea of fake news is not a novel concept. Notably, the idea has been in existence even before the emergence of the Internet as publishers used false and misleading information to further their interests. Following the advent of the web, more and more consumers began forsaking the traditional media channels used to disseminate information for online platforms. Not only does the latter alternative allow users to access a variety of publications in one sitting, but it is also more convenience and faster. The development, however, came with a redefined concept of fake news as content publishers began using what has come to be commonly referred to as a clickbait

Review of Literature

If we look at some scholar work shows the issue that the fake news has been major concerned amongst scholar from various background. For instance, some authors have observed that fake news is no longer a preserve of the marketing and public relations departments. Instead, there is an increasing risk of IT security, therefore, IT department is premised on the idea that it would help avert the various risks associated with the problem. So, if we good deeply into it we could find that the hackers use clickbait with the help of fake news and make some professional of the organization downloads their malicious exploits in their system or leak sensitive information, albeit in an indirect manner.

The user may, for instance, be tricked into believing that they are helping to disseminate the news further when, in the actual sense, they are providing the perpetrators with access to their emails, and we can also see that the fake news are worked extensively as they are using videos with original message and uses their facial structure to replace the message with false message they want us to believe, these fake news issues is bigger day by day and we need to implement more our research and extensive knowledge to solve the problem.

Motivation for the Problem Undertaken

To build an application which can detect the fake and true news.

Analytical Problem Framing

Mathematical/ Analytical Modelling of the Problem

Cleaned Data by removing irrelevant features

Pre-processing of text using NLP processing

Used Word Counts

Used Character Counts

Used Tf-Idf Vectorizer

Split data into train and test

Built Model

Hyper parameter tuning

Data Sources and their formats

The data-set is in csv format: **fake_news.csv** and **true_news.csv**. Features of this dataset are:

title
text (containing news)
subject
date

Data Preprocessing Done

Data pre-processing is the process of converting raw data into a well- readable format to be used by Machine Learning model. Data pre- processing is an integral step in Machine Learning as the quality of data and the useful information that can be derived from it directly affects the ability of our model to learn; therefore, it is extremely important that we pre- process our data before feeding it into our model. I have used following pre-processing steps:

Importing necessary libraries and loading dataset as a data frame.

Used pandas to set display maximum columns ensuring not to find any truncated information.

Checked some statistical information like shape, number of unique values present, info, finding zero values etc.

Checked for null values and did not find any null values.

Dropped some unwanted columns as they are of no use for prediction.

Added a new label as '0' for fake news dataset and '1' for true news dataset.

Merge both fake and true dataset to a single dataset as news.

Visualized each feature using seaborn and matplotlib libraries by plotting count plot.

Separate feature and label data.

Data Inputs- Logic- Output Relationships

Text Pre-Processing

```
#Creating a function to process the texts
def wordopt(text):
    text = text.lower()
    text = re.sub('[\.\*\?\\]', '', text)
    text = re.sub("\\W", "", text)
    text = re.sub('https?://\\S+|www\\.\\S+', '', text)
    text = re.sub('<.*?>+', '', text)
    text = re.sub('[%s]' % re.escape(string.punctuation), '', text)
    text = re.sub('\\n', '', text)
    text = re.sub('\\w*\\d\\w*', '', text)
    return text
```

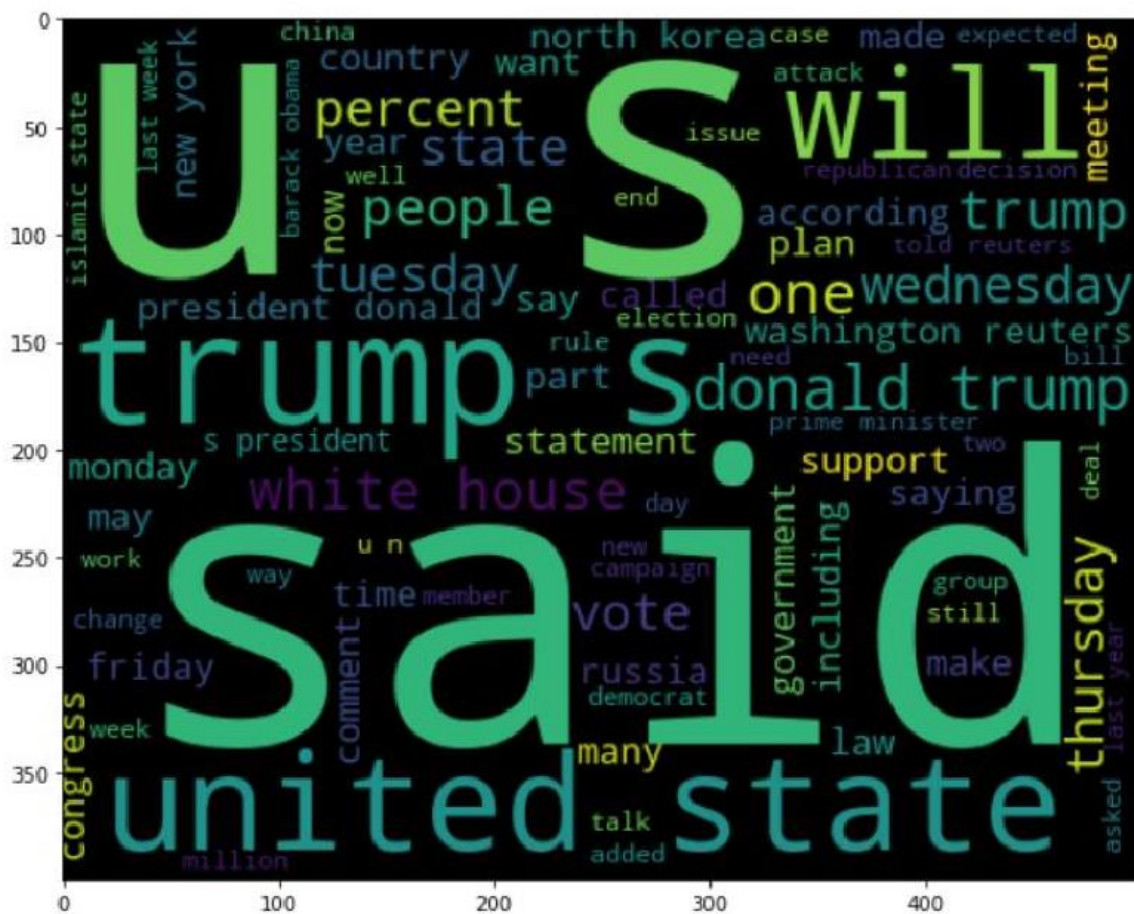
```
news['text'] = news['text'].apply(wordopt)
news.head()
```

	text	label
0	donald trump just couldn t wish all americans ...	0
1	house intelligence committee chairman devin nu...	0
2	on friday it was revealed that former milwauk...	0
3	on christmas day donald trump announced that ...	0
4	pope francis used his annual christmas day mes...	0

Wordcloud

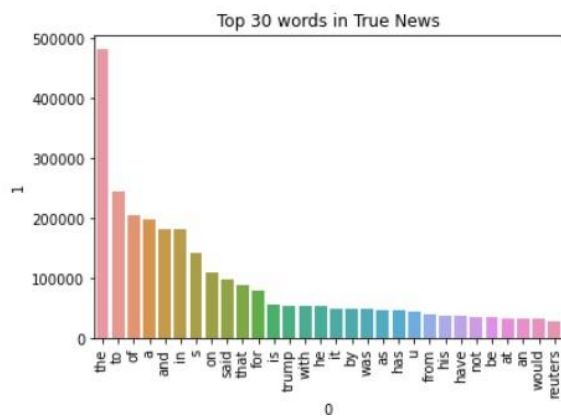
```
wc = WordCloud(width = 500, height = 400, min_font_size= 10, background_color= 'black')
```

```
#Generating Word Cloud for True News
true_wordcloud = wc.generate(news[news['label']==1]['text'].str.cat(sep = " "))
plt.figure(figsize=(12,8))
plt.imshow(true_wordcloud)
plt.show()
```




```
#Top 30 most frequently occurring words in True news
true_corpus = []
for msg in news[news['label']==1]['text'].tolist():
    for word in msg.split():
        true_corpus.append(word)
```

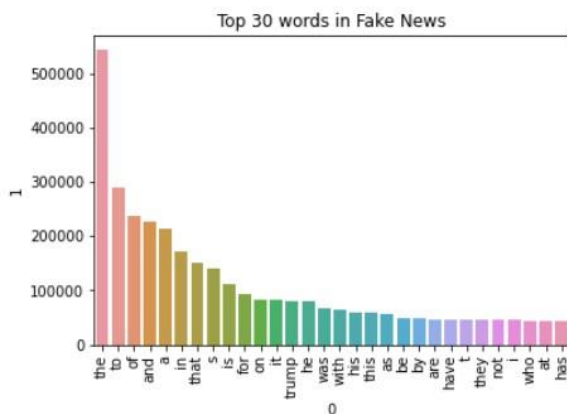
```
sns.barplot(pd.DataFrame(Counter(true_corpus).most_common(30))[0] , pd.DataFrame(Counter(true_corpus).most_common(30))[1])
plt.title("Top 30 words in True News")
plt.xticks(rotation = "vertical")
plt.show
```



```
#Top 30 most frequently occurring words in Fake news
```

```
fake_corpus = []  
for msg in news[news['label']==0]['text'].tolist():  
    for word in msg.split():  
        fake_corpus.append(word)
```

```
sns.barplot(pd.DataFrame(Counter(fake_corpus).most_common(30))[0] , pd.DataFrame(Counter(fake_corpus).most_common(30))[1])  
plt.title("Top 30 words in Fake News")  
plt.xticks(rotation = "vertical")  
plt.show
```



State the set of assumptions (if any) related to the problem under consideration

It was observed that there are two types of news: fake and true. So, must detect which news is fake and which is true.

Need to add one more column which is a target column for distinguishing fake and true news by labelling 0 for fake and 1 for true news.

It was observed that title, subject, and date column are irrelevant. So, we need to drop them. It was also observed that text column containing news have stop- words, punctuation so must replace or pre-process those values.

Also, must convert text (reviews) into vectors using Tf-Idf vectorization.

By looking into the Target Variable, it is assumed that it is a classification problem.

Hardware and Software Requirements and Tools Used

Hardware Used:

Processor — core i5 and above

RAM — 8 GB or above

SSD — 250GB or above

Software Used:

Programming language: Python

Distribution: Anaconda Navigator

Browser based language shell: Jupyter Notebook

Libraries/Packages Used:

Pandas, NumPy, matplotlib, seaborn, scikit-learn and

pandas_profiling

Model/s Development and Evaluation

Identification of possible problem

Solving approaches (methods)

In this project, we want to differentiate between comments and its categories and for this we have used these approaches:

Checked Total Numbers of Rows and Column

Checked All Column Name

Checked Data Type of All Data

Checked for Null Values

Description of Data

Dropped irrelevant Columns

Added one target column to distinguish which new is fake and which is true

Replaced special characters and irrelevant data

Checked all features through visualization.

Converted all messages to lower case

Removed special characters

Removed punctuations

Removed Stop-Words

Used Word Counts

Used Character Counts

Checked loud word using Word-Cloud

Converted text into vectors using Tf-Idf Vectorization

Testing of Identified Approaches (Algorithms)

Since label is my target variable, which is categorical in nature, from this I can conclude that it is a classification type problem hence I have used following classification algorithms

Logistic Regression
Decision Tree Classifier
Gradient Boosting Classifier
Random Forest Classifier
Linear Support Vector Classifier
Bernoulli NB
Multinomial NB
SGD Classifier
LGBM Classifier
XGB Classifier

Run and evaluate selected models

I used a total of 10 classification Models after choosing the random state amongst 1-100 number. I have used Decision Tree Classifier to find best random state and the code is as below. The code for the models is listed below.

Random State:

```
maxAccu = 0
maxRs = 0
for i in range(1,100):
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25,random_state=i)
    LR = LogisticRegression()
    LR.fit(x_train,y_train)
    pred_LR = LR.predict(x_test)
    acc = accuracy_score(y_test,pred_LR)
    if acc>maxAccu:
        maxAccu=acc
        maxRs = i
print(f'Best Accuracy is {maxAccu} on Random_state {maxRs}')
```

Best Accuracy is 0.98913140311804 on Random_state 32

Model Building:

```
# Defining the Classification Machine Learning Algorithms
lr = LogisticRegression(solver='lbfgs')
dtc = DecisionTreeClassifier()
gbc = GradientBoostingClassifier(random_state=0)
rfc = RandomForestClassifier(random_state=0)
svc = LinearSVC()
bnb = BernoulliNB()
mnb = MultinomialNB()
sgd = SGDClassifier()
lgb = LGBMClassifier()
xgb = XGBClassifier(verbosity=0)
```

```

# Creating a function to train and test the model with evaluation metrics
def BuiltModel(model):
    print('*'*30+model.__class__.__name__+'*'*30)
    model.fit(x_train, y_train)
    y_pred = model.predict(x_train)
    # Prediction
    pred = model.predict(x_test)

    # Accuracy Score
    accuracy = accuracy_score(y_test, pred)*100
    print(f"ACCURACY SCORE PERCENTAGE:", accuracy)

    # Mean Absolute Error(MAE)
    print('Mean Absolute Error(MAE)',mean_absolute_error(y_test,pred))

    # Mean Squared Error(MSE)
    print('Mean Squared Error',mean_squared_error(y_test,pred))

    # Root Mean Squared Error (RMSE)
    print('Root Mean Squared Error',np.sqrt(mean_squared_error(y_test,pred)))

    # Classification report
    print(f"CLASSIFICATION REPORT: \n {classification_report(y_test, pred)}")

    # Confusion matrix and
    print(f"CONFUSION MATRIX: \n {confusion_matrix(y_test, pred)}\n")

    print("-"*120)
    print("\n")

```

```

for model in [lr,dtc,gbt,rfc,svc,bnb,mnb,sgd,lgb]:
    BuiltModel(model)

```

*****LogisticRegression*****

ACCURACY SCORE PERCENTAGE: 98.913140311804

Mean Absolute Error(MAE) 0.01086859688195991

Mean Squared Error 0.01086859688195991

Root Mean Squared Error 0.1042525629515165

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.99	0.99	0.99	5861
1	0.99	0.99	0.99	5364
accuracy			0.99	11225
macro avg	0.99	0.99	0.99	11225
weighted avg	0.99	0.99	0.99	11225

CONFUSION MATRIX:

[[5794 67]

[55 5309]]

*****DecisionTreeClassifier*****

ACCURACY SCORE PERCENTAGE: 99.55456570155901

Mean Absolute Error(MAE) 0.004454342984409799

Mean Squared Error 0.004454342984409799

Root Mean Squared Error 0.06674086442659999

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.99	1.00	1.00	5861
1	1.00	0.99	1.00	5364
accuracy			1.00	11225
macro avg	1.00	1.00	1.00	11225
weighted avg	1.00	1.00	1.00	11225

CONFUSION MATRIX:

```
[[5842  19]
 [  31 5333]]
```

*****GradientBoostingClassifier*****

ACCURACY SCORE PERCENTAGE: 99.59910913140313

Mean Absolute Error(MAE) 0.004008908685968819

Mean Squared Error 0.004008908685968819

Root Mean Squared Error 0.06331594337896909

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	1.00	0.99	1.00	5861
1	0.99	1.00	1.00	5364
accuracy			1.00	11225
macro avg	1.00	1.00	1.00	11225
weighted avg	1.00	1.00	1.00	11225

CONFUSION MATRIX:

```
[[5830  31]
 [  14 5350]]
```

*****RandomForestClassifier*****

ACCURACY SCORE PERCENTAGE: 99.75055679287304

Mean Absolute Error(MAE) 0.0024944320712694877

Mean Squared Error 0.0024944320712694877

Root Mean Squared Error 0.049944289676293205

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	5861
1	1.00	1.00	1.00	5364
accuracy			1.00	11225
macro avg	1.00	1.00	1.00	11225
weighted avg	1.00	1.00	1.00	11225

CONFUSION MATRIX:

```
[[5851  10]
 [  18 5346]]
```

*****LinearSVC*****

ACCURACY SCORE PERCENTAGE: 99.49220489977728

Mean Absolute Error(MAE) 0.005077951002227171

Mean Squared Error 0.005077951002227171

Root Mean Squared Error 0.07125974320910208

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.99	1.00	1.00	5861
1	1.00	0.99	0.99	5364
accuracy			0.99	11225
macro avg	0.99	0.99	0.99	11225
weighted avg	0.99	0.99	0.99	11225

CONFUSION MATRIX:

```
[[5839  22]
 [  35 5329]]
```


*****BernoulliNB*****

ACCURACY SCORE PERCENTAGE: 96.09799554565701

Mean Absolute Error(MAE) 0.039020044543429847

Mean Squared Error 0.039020044543429847

Root Mean Squared Error 0.1975349198076883

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.97	0.96	0.96	5861
1	0.96	0.96	0.96	5364
accuracy			0.96	11225
macro avg	0.96	0.96	0.96	11225
weighted avg	0.96	0.96	0.96	11225

CONFUSION MATRIX:

[[5623 238]

[200 5164]]

*****MultinomialNB*****

ACCURACY SCORE PERCENTAGE: 93.21158129175946

Mean Absolute Error(MAE) 0.06788418708240535

Mean Squared Error 0.06788418708240535

Root Mean Squared Error 0.2605459404450688

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.94	0.93	0.93	5861
1	0.93	0.93	0.93	5364
accuracy			0.93	11225
macro avg	0.93	0.93	0.93	11225
weighted avg	0.93	0.93	0.93	11225

CONFUSION MATRIX:

[[5477 384]

[378 4986]]

*****SGDClassifier*****

ACCURACY SCORE PERCENTAGE: 99.26057906458797

Mean Absolute Error(MAE) 0.0073942093541202675

Mean Squared Error 0.0073942093541202675

Root Mean Squared Error 0.08598958863792912

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.99	0.99	0.99	5861
1	0.99	0.99	0.99	5364
accuracy			0.99	11225
macro avg	0.99	0.99	0.99	11225
weighted avg	0.99	0.99	0.99	11225

CONFUSION MATRIX:

```
[[5822  39]
 [  44 5320]]
```

*****LGBMClassifier*****

ACCURACY SCORE PERCENTAGE: 99.74164810690424

Mean Absolute Error(MAE) 0.0025835189309576837

Mean Squared Error 0.0025835189309576837

Root Mean Squared Error 0.05082832803622094

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	1.00	1.00	1.00	5861
1	1.00	1.00	1.00	5364
accuracy			1.00	11225
macro avg	1.00	1.00	1.00	11225
weighted avg	1.00	1.00	1.00	11225

CONFUSION MATRIX:

```
[[5842  19]
 [  10 5354]]
```

Cross validation score for best score models

```
def cross_val(model):
    print('*'*30+model.__class__.__name__+''*30)
    scores = cross_val_score(model,x,y, cv = 3).mean()*100
    print("Cross validation score:", scores)
    print("\n")

for model in [lr,dtc,gbt,rfc,svc,bnb,mnb,sgd,lgb]:
    cross_val(model)
```

*****LogisticRegression*****

Cross validation score: 96.95754822041071

*****DecisionTreeClassifier*****

Cross validation score: 99.1135462604125

*****GradientBoostingClassifier*****

Cross validation score: 99.35854603768543

*****RandomForestClassifier*****

Cross validation score: 99.42536415875985

*****LinearSVC*****

Cross validation score: 98.48768319301527

*****BernoulliNB*****

Cross validation score: 90.40714508441356

*****MultinomialNB*****

Cross validation score: 88.3669651209408

*****SGDClassifier*****

Cross validation score: 97.80391108735355

*****LGBMClassifier*****

Cross validation score: 99.55900040090873

Hyper Parameter Tuning

Linear SVC with GridSearchCV

```
# Lets select the different parameters for tuning our best model (Linear SVC)
grid_params = {'C':(0.001, 0.01, 0.1, 1, 10),
               'penalty':('l1','l2'),
               'loss':('hinge','squared_hinge')}

# Train the model with given parameters using GridSearchCV
LSVC = GridSearchCV(svc, grid_params, cv=3)
LSVC.fit(x_train, y_train)
```

```
GridSearchCV(cv=3, estimator=LinearSVC(),
             param_grid={'C': (0.001, 0.01, 0.1, 1, 10),
                          'loss': ('hinge', 'squared_hinge'),
                          'penalty': ('l1', 'l2')})
```

```
# Selecting the best parameters found by GridSearchCV
LSVC.best_params_
```

```
{'C': 1, 'loss': 'squared_hinge', 'penalty': 'l2'}
```

```

# Final Model with the best chosen parameters list
best_model = LinearSVC(C= 1, loss= 'squared_hinge', penalty= 'l2')
best_model.fit(x_train,y_train) # fitting data to the best model
pred = best_model.predict(x_test)
accuracy = accuracy_score(y_test, pred)*100
# Printing the accuracy score
print("ACCURACY SCORE:", accuracy)
# Printing the classification report
print(f"\nCLASSIFICATION REPORT: \n {classification_report(y_test, pred)}")
# Printing the Confusion matrix
print(f"\nCONFUSION MATRIX: \n {confusion_matrix(y_test, pred)}")

```

ACCURACY SCORE: 99.49220489977728

CLASSIFICATION REPORT:

	precision	recall	f1-score	support
0	0.99	1.00	1.00	5861
1	1.00	0.99	0.99	5364
accuracy			0.99	11225
macro avg	0.99	0.99	0.99	11225
weighted avg	0.99	0.99	0.99	11225

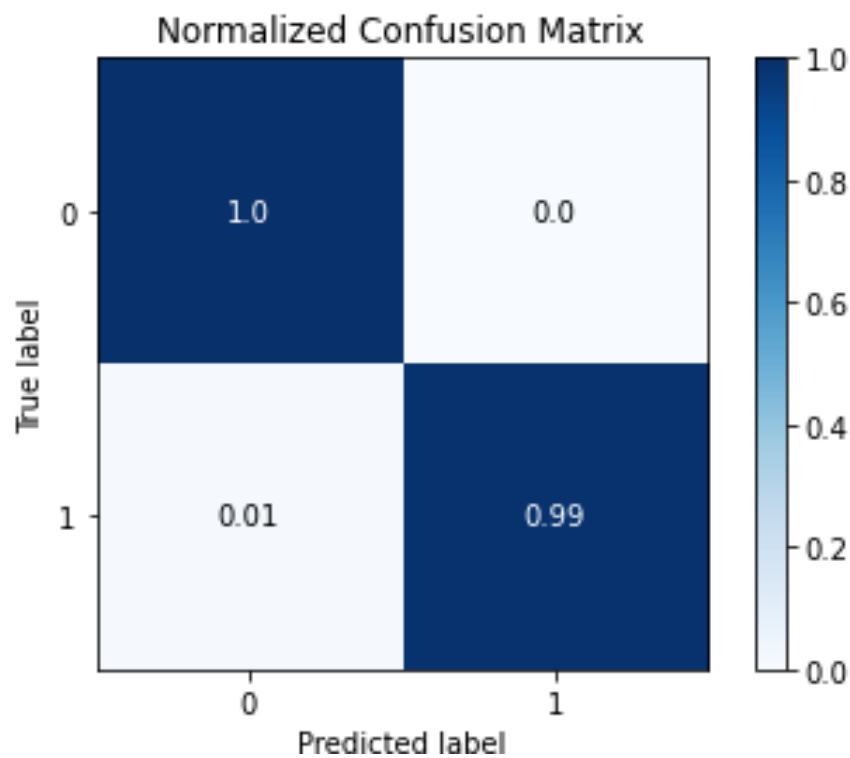
CONFUSION MATRIX:

```

[[5839  22]
 [ 35 5329]]

```

```
# Creating a normalized confusion matrix here  
skplt.metrics.plot_confusion_matrix(y_test, pred, normalize=True)
```



Saving the Final model

```
# Saving the model using joblib Library
import joblib
joblib.dump(best_model, "Fake_News_Prediction.pkl")
```

```
['Fake_News_Prediction.pkl']
```

Loading the saved model and predicting

```
# Loading the saved model
Model=joblib.load("Fake_News_Prediction.pkl")

#Prediction
prediction = Model.predict(x_test)
# Analysing Predicted vs Actual results
Fake_News_Detection = pd.DataFrame()
Fake_News_Detection['Predicted Fake News'] = prediction
Fake_News_Detection['Actual Fake News'] = y
Fake_News_Detection
```

	Predicted Fake News	Actual Fake News
0	0	0
1	1	0
2	1	0
3	1	0
4	0	0
...
11220	0	0
11221	1	0
11222	0	0
11223	1	0
11224	1	0

11225 rows × 2 columns

Key Metrics for success in solving problem under Consideration

The key metrics used here were Accuracy Score, Precision, Recall, F1 score, Cross Validation Score, Roc AUC Score and Confusion Matrix. We tried to find out the best parameters and to increase our scores by using Hyperparameter Tuning and used RandomizedSearchCV method.

Accuracy score means how accurate our model is that is the degree of closeness of the measured value to a standard or true value. It is one metric for evaluating classification models. Accuracy is the ratio of number of correct predictions into number of predictions.

Precision is the degree to which repeated measurements under the same conditions are unchanged. It is amount of information that is conveyed by a value. It refers to the data that is correctly classified by the classification algorithm.

Recall is how many of the true positives were recalled (found). Recall refers to the percentage of data that is relevant to the class. In binary classification problem recall is calculated as below:

$$\text{Recall} = \frac{\text{Number of True Positives}}{(\text{Total number of True Positives} + \text{Total number of False Negatives})}$$

F1 Score is used to express the performance of the machine learning model (or classifier). It gives the combined information about the precision and recall of a model. This means a high F1-score indicates a high value for both recall and precision.

Cross Validation Score is a technique in which we train our model using the subset of the data-set and then evaluate using the complementary subset of the data-set. It is used to protect against overfitting in a predictive model, particularly in a case where the amount of data may be limited. In cross-validation, you make a fixed number of folds (or partitions) of the data, run the analysis on each fold, and then average the overall error estimate. It is used to estimate the performance of ML models.

Roc Auc Score: The Receiver Operator Characteristic (ROC) curve is an evaluation metric for binary classification problems. It is a probability curve that plots the TPR against FPR at various threshold values.

The Area Under Curve (AUC) is the measure of the ability of a classifier to distinguish between classes and is used as a summary of the ROC curve.

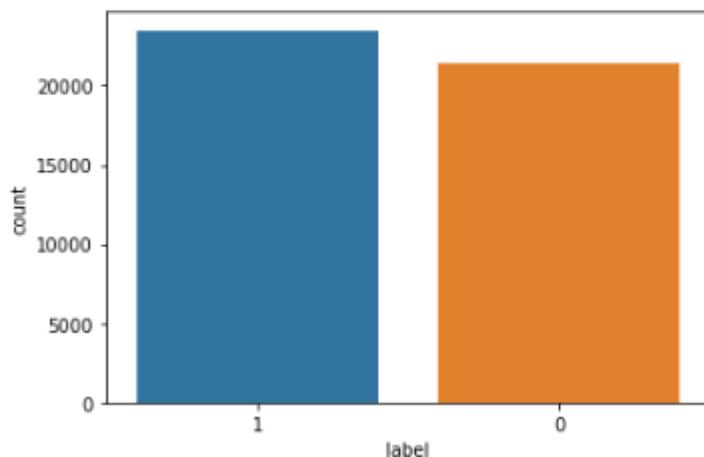
Confusion Matrix is one of the evaluation metrics for machine learning classification problems, where a trained model is being evaluated for accuracy and other performance measures. And this matrix is called the confusion matrix since it results in an output that shows how the system is confused between the two classes.

Hyperparameter Tuning: There is a list of different machine learning models. They all are different in some way or the other, but what makes them different is nothing but input parameters for the model. These input parameters are named as Hyperparameters. These hyperparameters will define the architecture of the model, and the best part about these is that you get a choice to select these for your model. You must select from a specific list of hyperparameters for a given model as it varies from model to model.

Visualizations

Using Countplot

```
sns.countplot(data=news, x='label', order=news['label'].value_counts().index)
```



Interpretation of the Results

Through Pre-processing it is interpreted that all text is converted to lower case, removed Punctuation, replaced extra space, removed stop-words, Calculated length of sentence, words and characters, converted text using Tf-Idf Vectorization.

Natural Language Processing and Machine Learning is used in this project.

Used 10 Machine Learning Algorithms for choosing one best model which is giving best accuracy than others.

By creating/building model we get best model: LinearSVC.

CONCLUSION

Key findings and conclusions of the study

In this project we have detected which news are fake news and which are true news. Then we have done different text process to eliminate problem of imbalance. By doing different EDA steps we have analyzed the text.

We have checked frequently occurring words in our data as well as rarely occurring words. After all these steps we have built function to train and test different algorithms and using various evaluation metrics we have selected Linear-SVC for our final model.

Finally, by doing hyperparameter tuning we got optimum parameters for our final model. And finally, we got improved accuracy score for our final model.

Learning Outcomes of the Study in respect of Data Science

This project has demonstrated the importance of NLP.

Through different powerful tools of visualization, we were able to analyze and interpret the huge data and with the help of count plot & word cloud, I can see the distribution of fake and true news.

Through data cleaning we were able to remove unnecessary columns, values, stop-words and punctuation from our dataset due to which our model would have suffered from overfitting or underfitting.

Limitations of this work and Scope for Future Work

As we know there are two types of news to detect. So, it is difficult to detect with higher accuracies. Still, we can improve our accuracy by fetching more data and by doing extensive hyperparameter tuning.