

Sentiment Analysis for Product Review using Machine Learning and Deep Learning Methods

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Introduction

- ❑ In today's digitalized world, e-commerce is growing popularity since it brings things closer to customers without requiring them to leave their homes.
- ❑ Sentiment analysis helps to analyze these opinioned data and extract some important insights which will help to other user to make decision.
- ❑ In this competitive business world, various industries especially e-commerce immensely use sentiment analysis to increase productivity and make better business decisions.

Motivation

- ❑ Sentiment analysis is currently at the leading position of scientific research. Its applications are beneficial in many cases.
- ❑ This thesis investigates the sentiment classification problem for online reviews, employing supervised approaches to determine the overall semantic of customer reviews by classifying them as positive, negative, or neutral.
- ❑ A model could be assist possible clients with settling on an informed decision on their purchase and organizations to improve their items or services .

Objective

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- ❑ My research aims to do this by conducting sentiment analysis on customer reviews and classifying them as positive, negative, or neutral.
- ❑ I believe that my suggested approach would reduce consumer frustration while shopping online since they will be able to examine product reviews based on the ratio of previous customers positive, negative, and neutral comments.
- ❑ It can also be beneficial to the vendor because it allows him to spot product flaws and give better customer service.

State Of The Art

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| Title | Author | Date & Publisher | Findings |
|--|--|---|---|
| 1. Sentiment Analysis For Product Review | Najma Sultana, Pintu Kumar, Monika Rani Patra, Sourabh Chandra and S.K. Safikul Alam | International Journal of Soft Computing, April 2019 | Algorithm : Naive Bayes, Logistic Regression, Linear SVC ,Decision Tree Accuracy : 89.85%.(NB) Dataset : Stanford dataset for 50,000 movie reviews. Limitation: They use machine learning algorithm.The preprocess part have only Punctution remove and used one kinds of feature extraction approach. |
| 2. Sentiment Analysis For Amazon.Com Reviews | Levent Guner, Emilie Coyne and Jim Smit | International Conference on Intelligent Computing and Control (I2C2), March 2019 | Algorithm : Linear SVM, Multinomial NB,LSTM network Accuracy : 90%.(LSTM) Dataset : Kaggle dataset and Amazon review data. Limitation: They use only three algorithm and use one feature extraction. |

State Of The Art

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| Title | Author | Date & Publisher | Findings |
|---|---|--|---|
| 3. Sentiment Analysis on Large Scale Amazon Product Reviews | Tanjim Ul Haque, Nudrat Nawal Saber and Faisal Muhammad Shah | International Conference on Innovative Research and Development (ICIRD), June 2018 | Algorithm : Linear SVM, Multinomial NB, Stochastic Gradient Descent ,Random Forest, Logistic Regression, decision Tree Accuracy : 90%.(SVM) Dataset : Amazon review data. Limitation: They use a supervised learning method and used a mix of 2 kinds of feature extractor approach. |
| 4.Sentiment Analysis on Product Reviews Using Machine Learning Techniques | Rajkumar S. Jagdale, Vishal S. Shirsat and Sachin N. Deshmukh | International Journal of Soft Computing, January 2019 | Algorithm : Naive Bayes, SVM Accuracy : 98.17%(NB) Dataset : Amazon review data Limitation: They use only Two algorithm. |

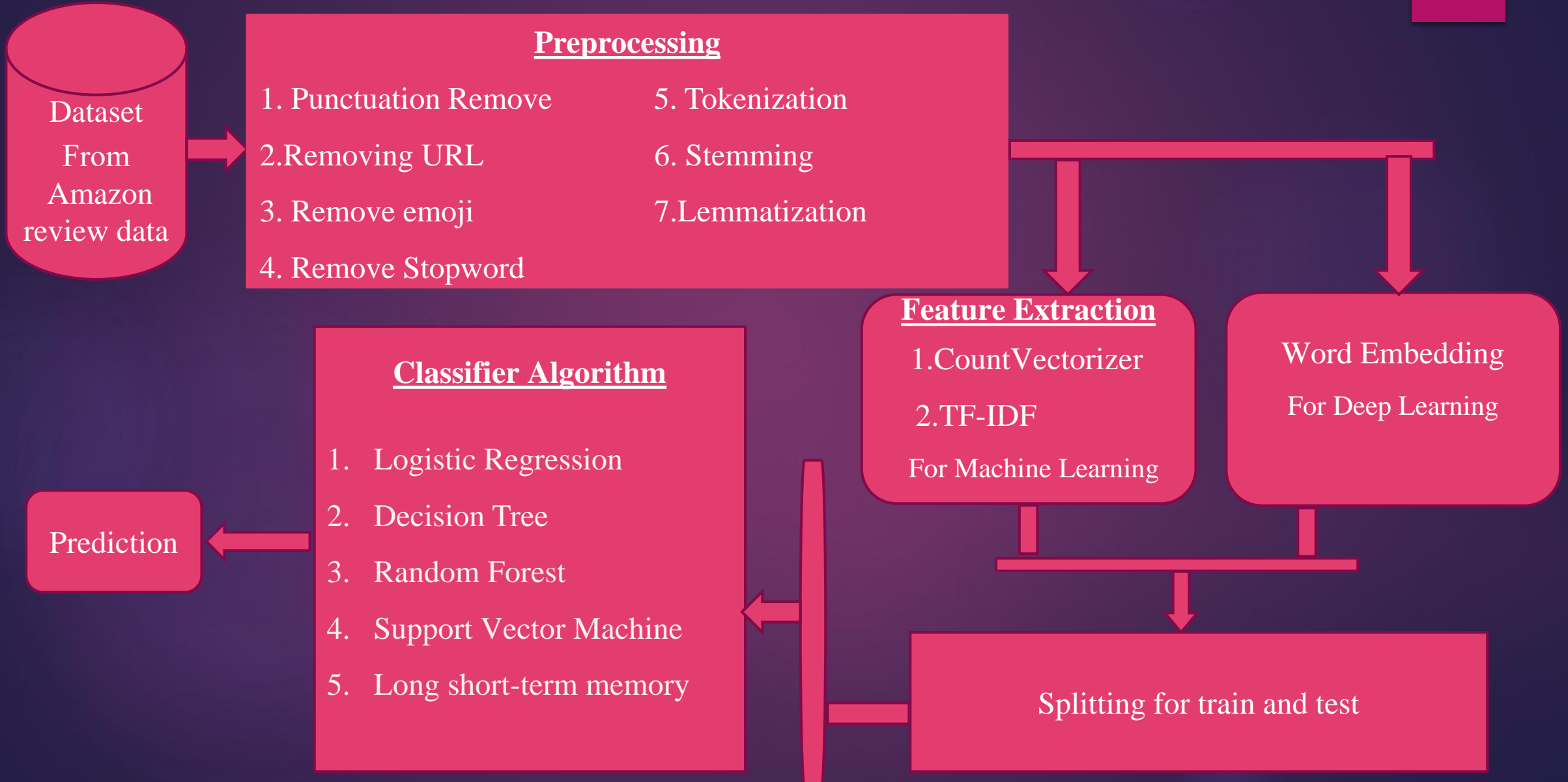
State Of The Art

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| Title | Author | Date & Publisher | Findings |
|--|---|--|---|
| 5. Sentiment Analysis Of Customer Product Reviews Using Machine Learning | Zeenia Singla, Sukhchandran Randhawa and Sushma Jain | International Conference on Intelligent Computing and Control (I2C2), June 2017 | Algorithm : Naive Bayes, SVM ,Decision Tree Accuracy : 81.77%. Dataset : Amazon review data. Limitation: Accuracy is low. |

Methodology

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Tools & Technologies

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- I. Python
- II. Pandas
- III. NumPy
- IV. Matplotlib
- V. Scikit-learn
- VI. Google Colab

Dataset Collection

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I collect the Amazon Review dataset from Julian McAuley,UCSD website. "Clothing, Shoes, and Jewelry" is the name of this dataset. This dataset contains 278 677 reviews. From this dataset, I took a total of 55000 reviews, which I used on my thesis. This dataset has been labeled with three classes, which are positive, negative, and neutral.

Data Preprocessing

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Punctuation Remove

Tokenization

Removing URL

Stemming

Remove emoji

Lemmatization

Remove Stopword

Data Preprocessing

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- Punctuation Remove : !"#\$%&'()*+,-./:;<=>?@[\\]^_`{|}~

"love these glasses! perfect size"



love these glasses perfect size

- Removing URL : https://.*|http://.*

Go to: https://example.com



Go to

- Remove Stopwords :

I like this bag



like , bag

- Tokenization : break sentences into parts

I love these glasses



[I, love, these , glasses]

- Stemming:

Operator, Going



Operate, Go

- Lemmatization : Finding the same root

[Connection, Connected, Connecting]



[Connect]

Feature Extraction

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- Count Vectorizer : Vectorization is a process of converting the text data into a machine-readable form.

Text Input :

[' this pair is great ']



Vocabulary :

{ 'this': 3, 'pair': 2, 'is': 1, 'great': 0 }



Encoded Text is:

[[1 1 1 1]]

- TF-IDF : It is used to tokenize the documents learn the vocabulary and inverse the document frequency weightings, and allow to encode new documents.

Text Input :

['this pair is great', 'i love these glasses']

Vocabulary :

{ 'this': 6, 'pair': 4, 'is': 2, 'great': 1, 'love': 3, 'these': 5, 'glasses': 0 }

Encoded Text is:

[1.40546511 1.40546511 1.40546511 1.40546511 1.40546511 1.40546511
1.40546511]

Classification Algorithm

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- Logistic Regression (LR)
- Decision Tree (DT)
- Random Forest (RF)
- Support Vector Machine (SVM)
- Long short-term memory (LSTM)

Experimental Output

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```
Input Text: Bad qualilty..... material qualilty is very poor  
Output : Negative
```

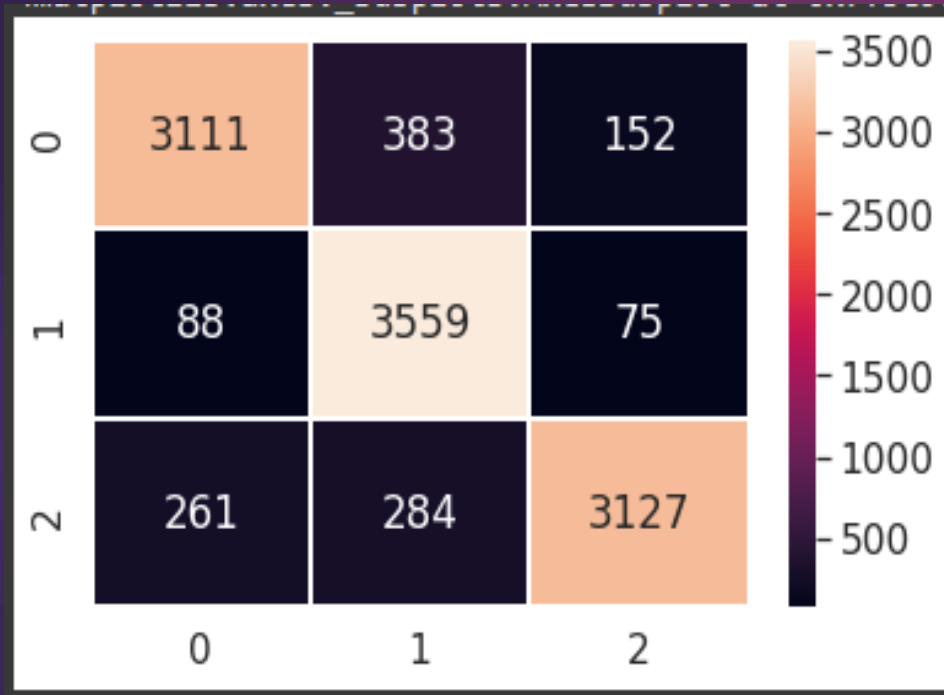
```
Input Text : ['Great quality']  
Output :Positive
```

```
Input Text : ['I buy this product from bangladesh']  
Output : Neutral
```


Experimental result

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► Confusion Matrix of Logistic Regression

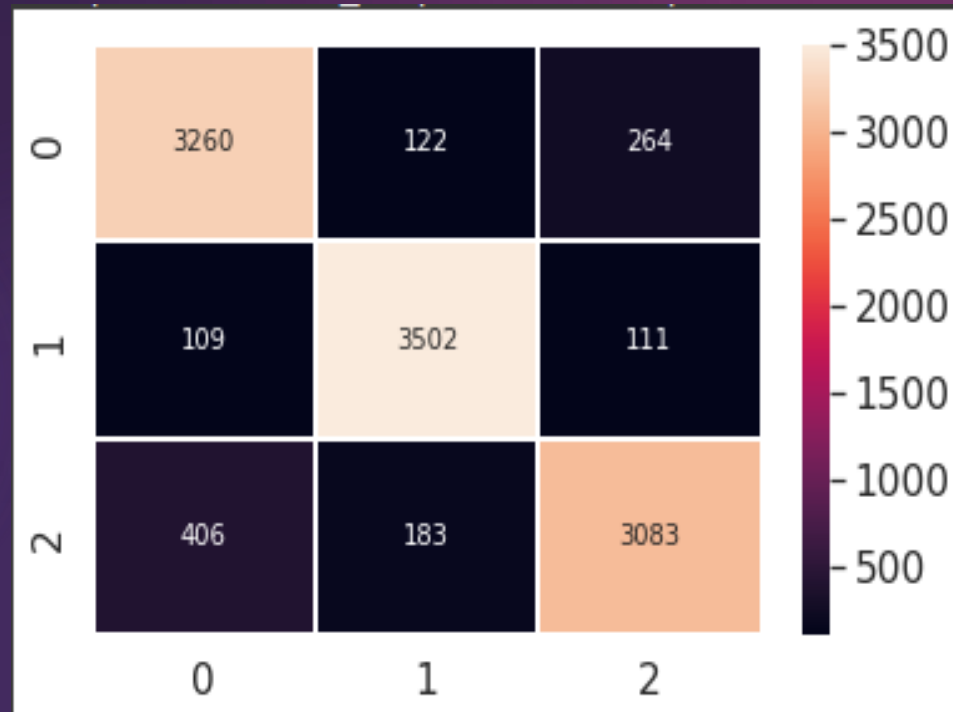


| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.90 | 0.85 | 0.88 | 3646 |
| 1 | 0.84 | 0.96 | 0.90 | 3722 |
| 2 | 0.93 | 0.85 | 0.89 | 3672 |
| accuracy | | | 0.89 | 11040 |
| macro avg | 0.89 | 0.89 | 0.89 | 11040 |
| weighted avg | 0.89 | 0.89 | 0.89 | 11040 |

Experimental result

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► Confusion Matrix of Decision Tree

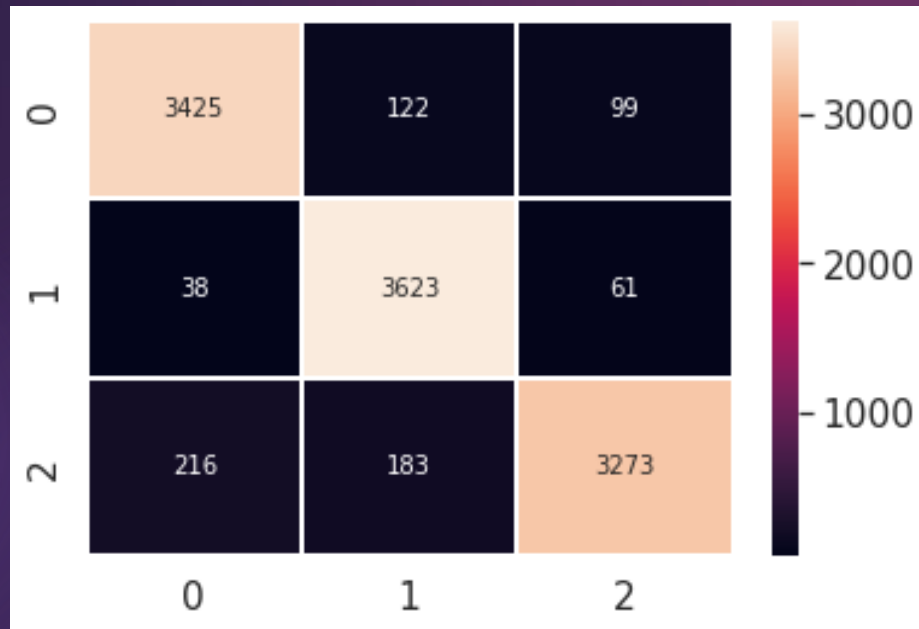


| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.86 | 0.89 | 0.88 | 3646 |
| 1 | 0.92 | 0.94 | 0.93 | 3722 |
| 2 | 0.89 | 0.84 | 0.86 | 3672 |
| accuracy | | | 0.89 | 11040 |
| macro avg | 0.89 | 0.89 | 0.89 | 11040 |
| weighted avg | 0.89 | 0.89 | 0.89 | 11040 |

Experimental result

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► Confusion Matrix of Random Forest

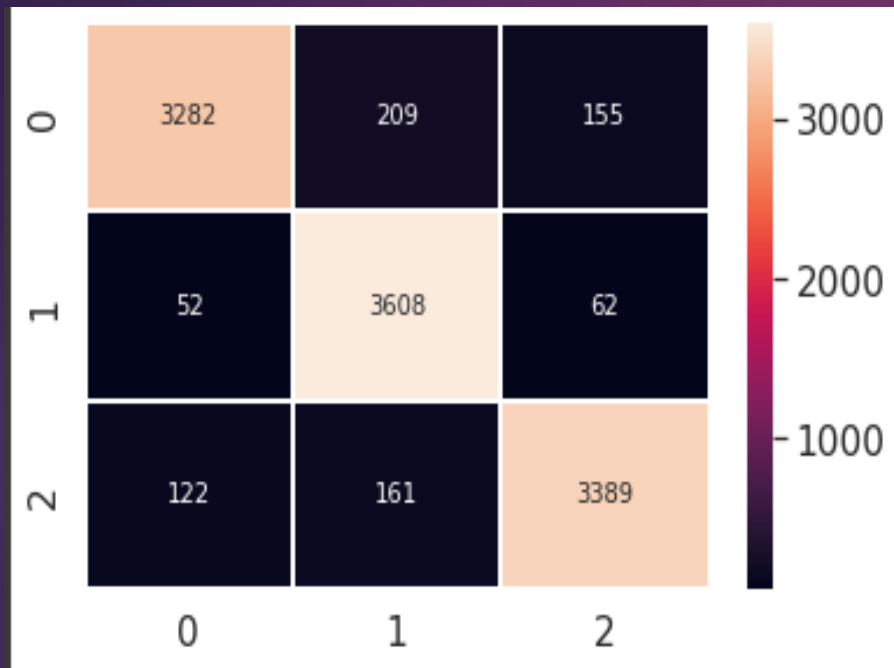


| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.93 | 0.94 | 0.94 | 3646 |
| 1 | 0.92 | 0.97 | 0.95 | 3722 |
| 2 | 0.95 | 0.89 | 0.92 | 3672 |
| accuracy | | | 0.93 | 11040 |
| macro avg | 0.94 | 0.93 | 0.93 | 11040 |
| weighted avg | 0.94 | 0.93 | 0.93 | 11040 |

Experimental result

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► Confusion Matrix of Support Vector Machine

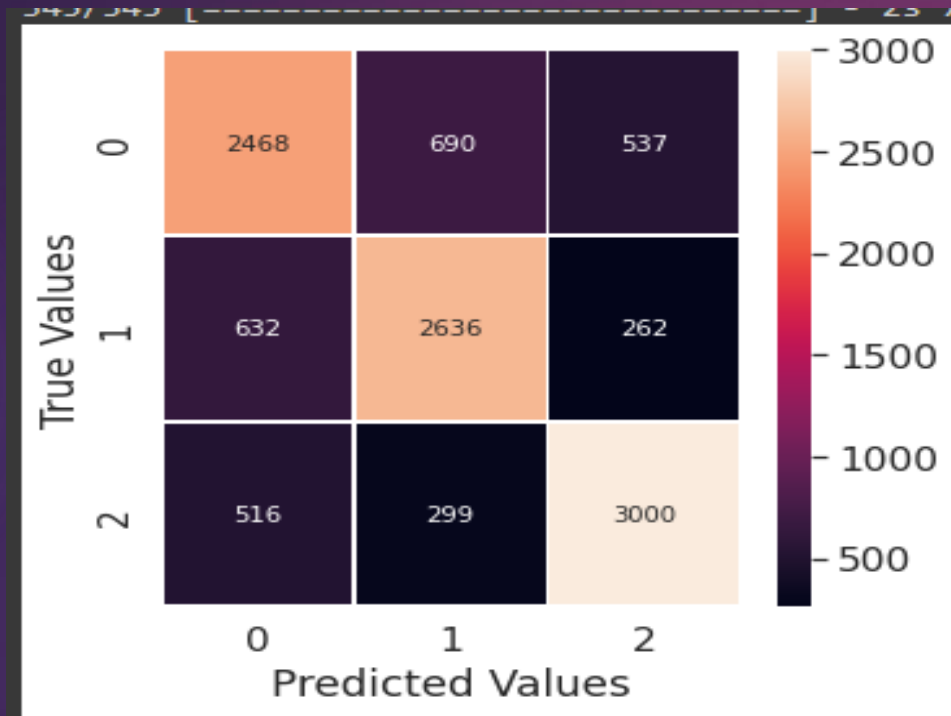


| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.95 | 0.90 | 0.92 | 3646 |
| 1 | 0.91 | 0.97 | 0.94 | 3722 |
| 2 | 0.94 | 0.92 | 0.93 | 3672 |
| accuracy | | | 0.93 | 11040 |
| macro avg | 0.93 | 0.93 | 0.93 | 11040 |
| weighted avg | 0.93 | 0.93 | 0.93 | 11040 |

Experimental result

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► Confusion Matrix of Long Short-Term Memory



| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.68 | 0.67 | 0.67 | 3616 |
| 1 | 0.76 | 0.71 | 0.74 | 3625 |
| 2 | 0.80 | 0.77 | 0.78 | 3799 |
| micro avg | 0.75 | 0.72 | 0.73 | 11040 |
| macro avg | 0.74 | 0.72 | 0.73 | 11040 |
| weighted avg | 0.75 | 0.72 | 0.73 | 11040 |
| samples avg | 0.72 | 0.72 | 0.72 | 11040 |

Experimental result

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Performance comparison of different classification algorithms

| Algorithm | Accuracy | Precision | Recall | F1-Score |
|-------------------------------|----------|-----------|--------|----------|
| Logistic Regression (LR) | 89% | 89% | 89% | 89% |
| Decision Tree (DT) | 89.14% | 89% | 89% | 89% |
| Random Forest (RF) | 93.55% | 94% | 93% | 93% |
| Support Vector Machine (SVM) | 93.13% | 93% | 93% | 93% |
| Long Short-Term Memory (LSTM) | 73.41% | 74% | 72% | 73% |

Experimental result

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Performance comparison of the target paper dataset and my dataset

| Algorithm | My Dataset | Target Paper Dataset |
|-------------------------------|------------|----------------------|
| Logistic Regression (LR) | 89% | 89.51% |
| Decision Tree (DT) | 89.14% | 71.91% |
| Random Forest (RF) | 93.55% | 85% |
| Support Vector Machine (SVM) | 93.13% | 89.88% |
| Long Short-Term Memory (LSTM) | 73.41% | 87.97% |

Conclusion

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- My research aims to do this by conducting sentiment analysis on customer reviews and classifying them as positive, negative, or neutral.
- Out of the five classifiers, i.e. RF, DT, SVM, LR and LSTM, predictive accuracy of RF is found to be the best. The highest value of accuracy achieved was 93.55% for RF among the five models.
- I believe that my suggested approach would reduce consumer frustration while shopping online since they will be able to examine product reviews based on the ratio of previous customers' positive, negative, and neutral comments. It can also be beneficial to the vendor because it allows him to spot product flaws and give better customer service.

Future work

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This Work can be extended in the following manner in future

- ▶ Try to use more feature extraction.
- ▶ To use more Deep Learning Models.
- ▶ Improve deep learning model accuracy
- ▶ Adding more data.

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Thank You