Sentiment Analysis of Amazon Product Reviews

Tools

- Python
- Scikit-Learn
- NLTK
- NumPy, Pandas, Matplotlib

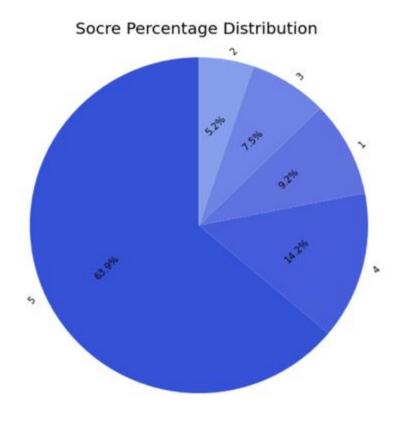
Dataset: Amazon Product Reviews

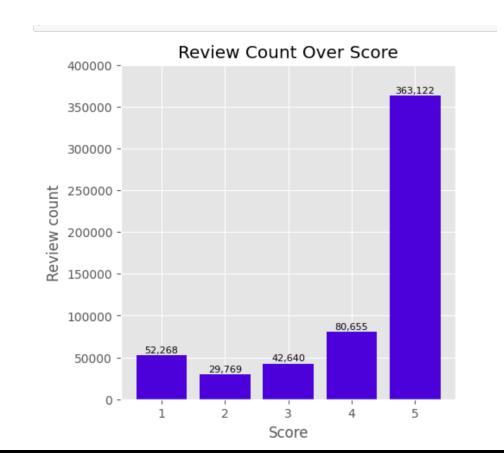
- Amazon fine food reviews
- 568,454 reviews
- 74,258 products
- 256,059 users
- Data source: https://www.kaggle.com/datasets/snap/amazon-fine-food-reviews

Introduction

- **Problem:** Building models which can identify the sentiment (positive or negative)
- Scope of work:
 - Analyzing the input text data and the corresponding response variables (ratings)
 - Performing basic pre-processing to prepare the data for modeling
 - Featurizing the reviews text
 - Building machine learning models to classify text as either positive or negative sentiment (1 or 0)

Data Exploration





Methodology

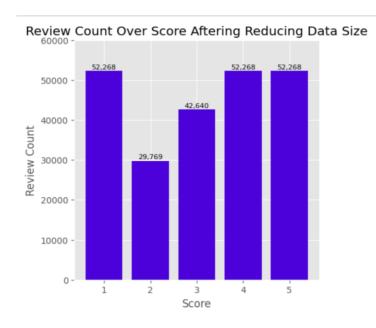
- Data pre-processing
- Machine learning models
- Confusion matrices

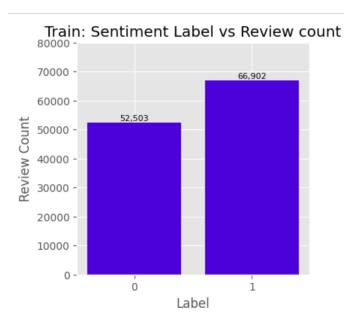
Data Pre-processing

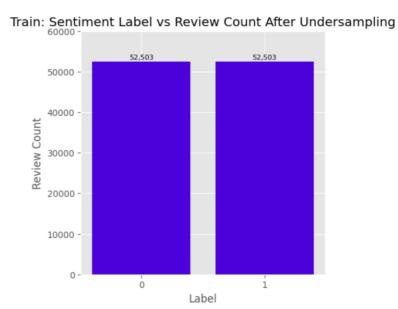
- Deleting redundant columns
- Random resampling without replacement
- Preparaing train validation and testing data
- Standardize the ratings for sentiment analysis
- Converting words to lower case
- Tokenizing the words

- Removing the special characters and the stop words
- Lemmetizing the words
- Making random undersampling to make data balanced
- Vectorizing the features
- Normalize the data

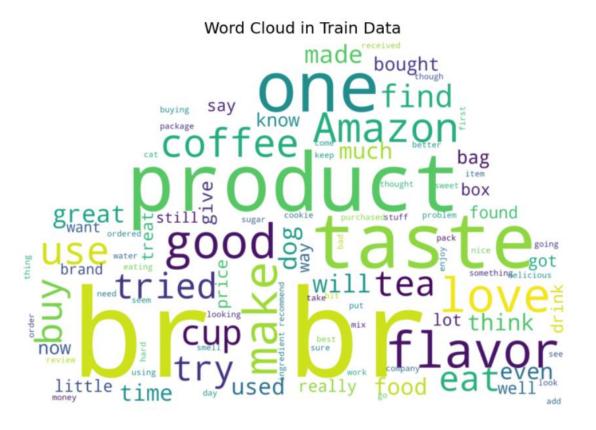
Data Preprocessing







Word Cloud



Machine Learning Models

- Bernoulli Naive Bayes
- Complement Naive Bayes
- Logistic Regression
- Support Vector Machine (SVM)
- Random Forest

Model Training and Evaluation

- I split the data into train and test sets. Then, I further split the training set into train and validation sets.
- Train the models using training set and tune hyperparameters using validation set.
- Model accuracies, precision scores, recall scores and F1-scores have been used to evaluate the model

	Best Parameters
Logistic Regression:	C=5, max_iter=30, tol=0.001
Linear Support Vector Classifier	C=0.1, max_iter=30, tol=0.01
Bernoulli Naive Bayes	alpha=0.1
Complement Naive Bayes	Alpha=10
Random Forest Classifier	Criterion="gini", max_depth=150, n_estimators=200

TF-IDF (ngram =(1 , 1))	Accuracy	Precision	Recall	F1-Score
Logistic Regression:	86.65%	89.36%	86.48%	87.90%
Linear Support Vector Classifier	87.29%	89.95%	87.03%	88.47%
Bernoulli Naive Bayes	83.39%	82.93%	88.59%	85.66%
Complement Naive Bayes	84.26%	87.69%	83.65%	85.62%
Random Forest Classifier	88.63%	89.69%	90.04%	89.87%

TF-IDF (ngram =(1,2))	Accuracy	Precision	Recall	F1-Score
Logistic Regression:	90.69%	92.78%	90.42%	91.58%
Linear Support Vector Classifier	91.15%	92.84%	91.24%	92.03%
Bernoulli Naive Bayes	87.98%	85.84%	94.05%	89.76%
Complement Naive Bayes	88.57%	91.81%	87.39%	89.54%
Random Forest Classifier	88.84%	89.23%	91.07%	90.14%

TF-IDF (ngram =(1 , 3))	Accuracy	Precision	Recall	F1-Score
Logistic Regression:	90.11%	92.27%	98.88%	91.06%
Linear Support Vector Classifier	91.21%	92.03%	91.29%	92.09%
Bernoulli Naive Bayes	88.29%	85.18%	95.76%	90.16%
Complement Naive Bayes	89.29%	92.83%	87.65%	90.16%
Random Forest Classifier	88.49%	88.51%	91.30%	89.89%

Conclusion

- In conclusion, I cleaned up and featurized an Amazon reviews dataset and built Logistic Regression, Linear Support Vector Classifier, Bernoulli Naive Bayes, Complement Naive Bayes, and Random Forest Classifier to predict sentiment (positive or negative).
- I found out that **Linear Support Vector Classifier** model got the highest accuracy (91.21) and outperformed among the proposed models.

References

- https://www.researchgate.net/publication/331291125_Sentiment_analysis_and_opinion_mining_applied_to_scientific_paper_reviews
- https://journalofbigdata.springeropen.com/articles/10.1186/s40537-015-0015-2
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