



Cinematica Sentiment Analysis

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Received: 29 January 2022

Accepted: 19 April 2022

Published: 30 May 2022

Abstract: *Sentiment analysis, a subfield of natural language processing, holds significant importance in understanding the emotions and opinions expressed in textual data. This project focuses on applying sentiment analysis techniques to movie reviews, aiming to develop an efficient model for automatically classifying sentiments as positive, negative, or neutral. The primary goal of this project is to create a robust sentiment analysis of accurately categorizing the sentiment conveyed in movie reviews. By leveraging machine learning algorithms and natural language processing techniques, the model aims to provide insights into audience reactions and contribute to the broader field of sentiment analysis.*

Keywords: *Cinematica Sentiment, Sentiment Analysis, Natural Language Processing, Machine Learning.*

1. INTRODUCTION

In today's digital age, online movie reviews have become an invaluable resource for both filmmakers and Moviegoers alike. These reviews offer a wealth of information about audience preferences, providing Insights into what resonates with viewers and what falls short. However, manually analyzing the sentiment of countless reviews can be a daunting task. This is where movie review sentiment analysis comes in. Movie Review sentiment analysis is a subfield of Natural Sentiment expressed in movie reviews. This involves automatically classifying reviews as positive, negative, or neutral based on the language used. By analyzing the sentiment of reviews, we can gain valuable insights Into Overall audience reception: Analyze the general sentiment towards a particular movie, identifying areas of Strength and weakness. Public opinion on specific aspects: Determine how viewers perceive various aspects of the movie, such as the plot, acting, directing, and cinematography. Predicting movie success: Gauge Potential audience reaction and box office performance based on early reviews. Personalizing Recommendations: Recommend movies tailored to individual preferences by analyzing past review history & sentiment.



Literature Survey

The rise of online platforms has facilitated the widespread sharing of movie reviews, providing a wealth of Data on audience preferences and film reception. Analyzing this data manually is time-consuming and inefficient. Sentiment analysis offers a solution by automatically classifying the sentiment of reviews as Positive, negative, or neutral, providing valuable insights for various stakeholders in the film industry. Early Research on movie review sentiment analysis primarily focused on lexicon-based methods.

These methods Employ dictionaries containing lists of positive, negative, and neutral words to determine the overall Sentiment of a review based on the frequency of specific words. While simple and efficient, lexicon-based Methods often struggle with sarcasm, double negatives, and domain-specific language.

The emergence of machine learning has significantly advanced the field of sentiment analysis. Algorithms Like Naïve Bayes, Support Vector Machines (SVM), and Logistic Regression have been successfully applied To movie review classification, achieving high accuracy levels. These algorithms learn from labeled datasets Of reviews, allowing them to identify patterns and perform sentiment classification more effectively than Lexicon-based methods.

Aurangzeb Khan, 2011 [19] outlined a rule-based method for sentiment analysis of software and customer evaluations that leverages Senti Word Net to achieve more accuracy than a lexicon-based approach. 91% of documents in the proposed system are accurate, and 86% of sentences are accurate.

Seven Rill et al., 2014 [13] suggested the "PoliTwi" program, which demonstrates the impact on concept-level sentiment analysis and the early identification of developing political issues on Twitter. Even before "Google Trends," hash tags are utilized in this paper's Twitter account to ascertain the outcomes of the US election. Data is gathered via the Twitter API and then subjected to an algorithm for sentiment analysis.

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2. METHODOLOGY

Movie review sentiment analysis is the process of using natural language processing (NLP) and machine learning techniques to automatically analyze the sentiment of online movie reviews and categorize them as positive, negative, or neutral. This information can be valuable for filmmakers, studios, and moviegoers alike. Here's an overview of the methodology involved in movie review sentiment analysis:



Data Acquisition and Preprocessing

The first step is to collect a dataset of labeled movie reviews. Several publicly available datasets exist, such as IMDB, Rotten Tomatoes, and Amazon Reviews. These datasets typically include the review text and a corresponding sentiment label (positive, negative, or neutral).

Once the data is collected, it needs to be preprocessed to prepare it for analysis. This involves tasks such as:

- **Cleaning:** Removing irrelevant characters and symbols, such as punctuation, special characters, and HTML tags.
- **Tokenization:** Breaking down the text into individual words or tokens.
- **Normalization:** Converting all words to lowercase or uppercase for consistent analysis.
- **Stemming or lemmatization:** Reducing words to their base form to improve accuracy.
- **Stop word removal:** Removing common words that do not contribute significantly to the sentiment, such as articles, prepositions, and conjunctions.

Feature Engineering

After preprocessing, the next step is to extract features from the text that are relevant for sentiment analysis. These features can be broadly categorized into: Linguistic features: Word n-grams, part-of-speech tags, sentiment lexicons (containing lists of positive and negative words), and stylistic features (e.g., exclamation points, capitalization). Domain-specific features: Features specific to the domain of movie reviews, such as character names, film titles. Naive Bayes: A simple and efficient algorithm that calculates the probability of a review belonging to a specific sentiment class based on the presence of keywords.

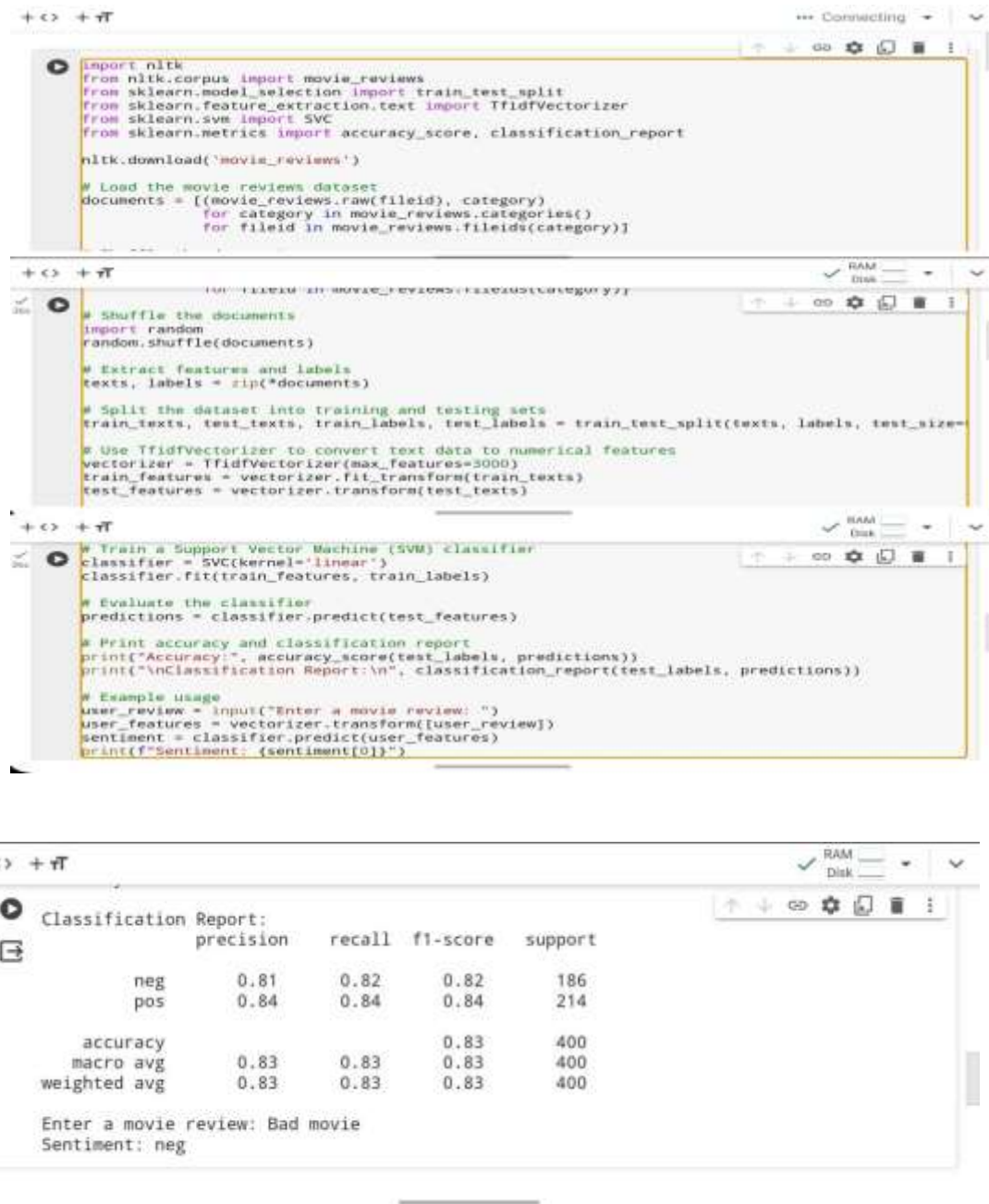
3. RESULT

Select an appropriate machine learning, and algorithm for regression tasks. Common choices include:

1. Linear Regression
2. Decision Trees

Model Deployment

If the model performs well in validation and testing, consider deploying it in a real-world setting for diabetes risk predictions. This can involve integrating the model with your application or healthcare system, predict diabetes risk for new individuals based on their health and lifestyle data.



```
import nltk
from nltk.corpus import movie_reviews
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report

nltk.download('movie_reviews')

# Load the movie reviews dataset
documents = [(movie_reviews.raw(fileid), category)
              for category in movie_reviews.categories()
              for fileid in movie_reviews.fileids(category)]

# Shuffle the documents
import random
random.shuffle(documents)

# Extract features and labels
texts, labels = zip(*documents)

# Split the dataset into training and testing sets
train_texts, test_texts, train_labels, test_labels = train_test_split(texts, labels, test_size=0.2)

# Use TfidfVectorizer to convert text data to numerical features
vectorizer = TfidfVectorizer(max_features=3000)
train_features = vectorizer.fit_transform(train_texts)
test_features = vectorizer.transform(test_texts)

# Train a Support Vector Machine (SVM) classifier
classifier = SVC(kernel='linear')
classifier.fit(train_features, train_labels)

# Evaluate the classifier
predictions = classifier.predict(test_features)

# Print accuracy and classification report
print("Accuracy:", accuracy_score(test_labels, predictions))
print("\nClassification Report:\n", classification_report(test_labels, predictions))

# Example usage
user_review = input("Enter a movie review: ")
user_features = vectorizer.transform([user_review])
sentiment = classifier.predict(user_features)
print(f"Sentiment: {sentiment[0]}")
```

	precision	recall	f1-score	support
neg	0.81	0.82	0.82	186
pos	0.84	0.84	0.84	214
accuracy			0.83	400
macro avg	0.83	0.83	0.83	400
weighted avg	0.83	0.83	0.83	400

Enter a movie review: Bad movie
Sentiment: neg

4. CONCLUSION

Movie review sentiment analysis has emerged as a powerful tool with vast potential to revolution in the film industry and enhance the moviegoer experience. By extracting valuable insights from online reviews, it offers a wealth of benefits for filmmakers, studios, online platforms, and movie goers alike



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