

Metaphor Detection Using logistic regression machine Learning Model

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Abstract

This study delves into metaphor detection within the realm of natural language processing, employing a Logistic Regression machine learning model. Working with a dataset comprising 1,870 training samples and 800 unseen test samples, the model achieved notable accuracy, demonstrating the efficacy of Logistic Regression in this challenging domain. This approach underscores the potential of Logistic Regression, especially its proficiency in deciphering complex linguistic features like metaphors. The results of this study offer promising insights into enhancing AI's understanding of human language, providing a solid foundation for future advancements in sophisticated NLP applications.

1 Introduction

Metaphor detection is a nuanced and intricate task in the realm of natural language processing (NLP), holding significant importance due to its widespread use in everyday language. Metaphors, which involve the use of figurative language to represent concepts or ideas indirectly, are a core element of human communication. They enrich language, convey complex ideas succinctly, and can evoke strong imagery or emotions. However, their inherent ambiguity and context-dependent nature pose substantial challenges for automated detection and interpretation.

In the digital age, where vast amounts of textual data are generated and consumed daily, the ability to automatically detect metaphors has far-reaching implications. Applications range from enhancing the performance of language models and sentiment analysis tools to contributing to fields like linguistics, psychology, and literature studies. Moreover, understanding metaphors is crucial for improving human-computer interaction, as it enables more natural and intuitive communication with AI systems.

The project at hand utilizes a Logistic Regression machine learning model to tackle the intricate task of metaphor detection. Renowned for its effectiveness in binary classification, Logistic Regression is a statistical method that excels in identifying relationships between features and outcomes. This approach is particularly suited for metaphor detection, given its ability to handle linguistic subtleties through pattern recognition in textual data.

Metaphors, with their inherent complexity and subtlety, often pose a challenge for more rudimentary rule-based or linear statistical approaches. By employing Logistic Regression, this project aims to address these challenges. Its strength lies in its straightforward interpretability and computational efficiency, which are crucial in processing and understanding the nuances of language in NLP tasks.

This endeavor is aimed at contributing to the enhancement of language understanding capabilities within NLP systems, thereby facilitating more sophisticated and human-like language processing techniques. By leveraging the logistic regression model's robust classification abilities, we seek to advance the field of NLP in deciphering complex linguistic phenomena like metaphors.

2 Dataset Overview

The effectiveness of a machine learning model, especially in tasks as intricate as metaphor detection, hinges significantly on the quality and appropriateness of the dataset used for training and testing. In this project, we utilized two distinct datasets for training and testing purposes.

2.1 Training dataset

The training dataset is a robust collection of 1,870 samples, each meticulously annotated to facilitate the training of the logistic regression model for metaphor detection. Key characteristics of the training data include:

Metaphor ID:

- Each sample is uniquely identified with a metaphor ID, allowing for precise tracking and analysis of individual instances.

Label (Boolean):

- Samples are labelled with a boolean value, where 'True' indicates the presence of a metaphor and 'False' signifies its absence. This binary classification forms the foundational basis for the model's learning process.

Text Content:

- The dataset encompasses a diverse range of text snippets. These samples vary in length and context, offering a comprehensive representation of different styles and uses of language. This variety is crucial for training the model to recognize metaphors across various textual formats and contexts.

The training data plays a pivotal role in shaping the model's ability to discern and accurately classify metaphoric language, providing a foundation for the model's learning and subsequent performance.

2.2 Testing dataset

In addition to the training set, the project utilizes a test dataset comprising 800 unseen samples. This dataset is critical for evaluating the model's effectiveness and generalizability in metaphor detection. Key aspects of the test data include:

Unseen Samples:

- These samples were not part of the model's training process, ensuring an unbiased evaluation of the model's performance.

Diverse Contexts:

- Similar to the training dataset, the test samples encompass a wide range of contexts and linguistic styles. This diversity tests the model's ability to generalize its learning to new and varied data.

Evaluation Metric:

- The primary metric for assessing the model on this test dataset is accuracy, calculated as the proportion of correctly identified instances (both metaphorical and non-metaphorical) out of the total number of samples.

The combination of the training and test datasets provides a comprehensive platform for developing and validating the logistic regression model's capability in detecting metaphors, a crucial aspect of understanding human language in computational linguistics.

3 Task Description

3.1 Objective

The primary objective of this project is to develop an efficient machine learning model, specifically a Logistic Regression classifier, for detecting the presence of metaphors in textual data. This endeavor is situated within the domain of natural language processing (NLP), where accurately interpreting figurative language is key for diverse applications such as sentiment analysis, automated text interpretation, and improving human-computer interactions.

3.2 Significance

Metaphors are a pervasive element in human language, offering richness and depth but also posing significant challenges in automatic language processing due to their inherent ambiguity and context dependence. Successfully identifying metaphors is vital for machines to comprehend the nuances of human language fully, a crucial step towards advanced AI language understanding.

3.3 Challenges

The task is challenging due to several factors like:

Variability and Contextual Nature:

- Metaphors can vary greatly in form and context, making them difficult to identify through rigid, rule-based methods.

Subtlety of Figurative Language:

- Metaphors often involve subtle nuances that require not just linguistic analysis but also an understanding of cultural and contextual elements.

162 **Balancing Precision and Recall:**

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164 • The model must accurately identify metaphors

165 (high precision) while minimizing the mis-

166 classification of non-metaphorical language

as metaphorical (high recall).

167 3.4 Approach

168 We chose Logistic Regression for metaphor detec-

169 tion for its clarity and efficiency in binary classifi-

170 cation. This model excels in identifying metaphors

171 by analyzing linear relationships in text features.

172 Its simplicity aids in quick computation and easy

173 interpretation, making it effective for processing

174 linguistic data.

175 3.5 Methodology

176 The project methodology includes these steps:

177 Data Preprocessing:

- 178 • Cleaning and preparing the textual data for
- 179 analysis, including tokenization, normaliza-
- 180 tion, and possibly feature extraction.

181 Model Training:

- 182 • Using the annotated training dataset to train
- 183 the random forest model, optimizing paramet-
- 184 ers for best performance.

185 Model Evaluation:

- 186 • Testing the model on an unseen dataset to eval-
- 187 uate its performance, primarily focusing on
- 188 accuracy as the key metric.

189 Forecasts:

- 190 • The model's forecasts are stored in a CSV file,
- 191 offering a comprehensive log of how well the
- 192 model performs on individual samples.

193 4 Analysis and Results

194 The analysis of the logistic regression model's per-

195 formance on metaphor detection is a critical aspect

196 of this project. The results are derived from the ap-

197 plication of the model on the training dataset, This

198 section outlines the key findings, highlighting the

199 model's accuracy and areas of effectiveness, as well

200 as identifying potential limitations.

➡ Accuracy: 85.29411764705883
Precision: 0.89568345323741
Recall: 0.9054545454545454
F1 Score: 0.9005424954792042

4.1 Model Performance 201

Accuracy: 202

- 203 • The model achieved an accuracy of 85.29%
- 204 on the test dataset. This metric is particularly
- 205 significant as it reflects the model's ability
- 206 to correctly identify both the presence and
- 207 absence of metaphors in the text.

Precision and Recall: 208

- 209 • While the primary focus of the analysis is on
- 210 accuracy, other metrics such as precision at
- 211 89.5% and recall at 90.5% were also consid-
- 212 ered to provide a more comprehensive view
- 213 of the model's performance.

F1 Score: 214

- 215 • The F1 score (90.05%) serves as a trade-off
- 216 between precision and recall, signifying ro-
- 217 bust overall model performance.

4.2 Strengths of the Model 218

Robustness in Varied Contexts: 219

- 220 • The model demonstrated strong performance
- 221 across a range of different textual contexts,
- 222 suggesting its effectiveness in handling the
- 223 diverse nature of language and metaphors.

Handling of Complex Patterns: 224

- 225 • The logistic regression algorithm's ability to
- 226 manage complex, non-linear relationships in
- 227 data was evident, contributing to its high ac-
- 228 curacy rate.

5 Conclusion 229

230 In conclusion, our exploration of Metaphor Detec-

231 tion using Logistic Regression has showcased the

232 model's robustness and accuracy in discerning com-

233 plex linguistic patterns in natural language. This

234 approach highlights the strength of logistic regres-

235 sion in handling the nuanced task of identifying

236 metaphors, balancing precision and interpretability.

237 While challenges persist, particularly in refining

238 the model for more subtle metaphors, our findings

open promising avenues for future research in enhancing AI’s linguistic understanding, making a significant contribution to the field of natural language processing.

References

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Team Members



Bhuvaneswar Sarakaranam



Teja Krishna.V



Chandresh Reddy Sura



Bhuvanesh Muppaneni