# **Detection of Insulting Comments in Online Discussion**

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#### **Abstract**

In this project, we aim to detect the comments that may appear insulting to other participants in an online discussion or conversation. We will use text processing techniques to generate features and then we will use supervised learning techniques to detect insulting vs non-insulting comments from the text. We are planning to use naive Bayes and logistic regression to set a baseline. After setting the baseline, we are aiming to use models like support vector machine, random forests. Finally we will use adaptive ensemble classifier to maximize accuracy. In order to avoid over-fitting, we will tune and evaluate our model via K-folds cross validation. The dataset is obtained from the popular data science competition portal, Kaggle.

#### 1 Introduction

With the prosperity of the Internet, ample amount of ways or spaces for public discussion has emerged which changes how do people communicate with each other. Either it is a section for comments or reviews for a news article or a forum to discuss some aspects of a particular product or event, these online discussion gives us opportunity to share our opinion and findings, as well as to know about the thoughts of others. Even though these discussions are expected to be productive, it allows people to post insulting or inappropriate comments. As a result, these comments can often create a hostile or uncomfortable environment for some users, who might stop visiting the site in future. So, our aim is to focus on insulting comments of the online (ex. blog/forum) discussion.

However, the comments containing insults but are targeted to a non-participant of conversation (like a celebrity etc.) are not marked as insults. Insults are of many types like: Taunts, reference to handicaps, improper language, slurs and racism which are aimed to attack the other person in an online discussion. While some other type of insults which mainly aim to embarrass the reader (not an attack) like crude language, provocative words, sarcasm, indirect reference. An effective solution to mitigate this problem is to build a system that can detect whether a comment is insulting or not. So, we are primarily interested in detecting comments that are intended to be insulting to other participants in an online discussion. The objective of this project is to do: build a machine learning system that can accurately classify online comments as insulting to other participants or not.

### 2 Previous solutions for the target task

Different attempts have been made to classify the insulting comments in online discussions. The work by Ellen Spertus [1] used static dictionary approach to build a feature vector for training but it suffers from high false positive rates. Another work by Altaf Mahmud et. al [2] used semantic rules but couldn't distinguish between the insults directed to non-participant and participant of conversation. The work by Razavi et. al [3] makes use of insulting and abusing language dictionary on top of bag-of words features in their proposed three-level classification machine learning model.

Priya Goyal and Kalra [4] used support vector machine and logistic regression to train their model in their work but there were some false positives. In the work by Heh [5], logistic regression and stochastic gradient descent is used for classification and training their parameter vector respectively. Prashant Ravi found SVM, the best classifier in terms of the most number of correctly classified instances in his work. So, in our work, we aim to build an efficient ensemble classifier excelling previous works to detect insults involving machine learning approaches.

# **3** Why is this related to machine learning?

As we all know machine learning explores the study and construction of algorithms for learning to do tasks. The learning that is being done is always based on some sort of observations or data, such as examples (the most common case in this course), direct experience, or instruction. So in general, machine learning is about learning to do better in the future based on what was experienced in the past.

The emphasis of machine learning is on automatic methods. In other words, the goal is to devise learning algorithms that do the learning automatically without human intervention or assistance. Since we are trying to build an automated system to detect insults, we need to particularly investigate and analyze different techniques in the field of machine learning to come up with an efficient and accurate classifier. So, this work is entirely based on machine learning techniques.

# 4 Proposed Method

We will use text processing techniques to extract features and then we will use supervised learning techniques to detect insulting and neutral comments from the text. We are planning to use naive Bayes and logistic regression to set a baseline. After setting the baseline, we are aiming to use models like support vector machine, random forests. Finally we will use adaptive ensemble classifier to maximize accuracy. In order to avoid over-fitting, we will tune and evaluate our model via K-folds cross validation.

#### 5 Details of the dataset

We obtained data for training and testing from Kaggle [6], which is a popular website that hosts machine learning competitions. The training data set contains 3947 examples and the testing data set contains 2647 examples, each of which consists of the text of a particular comment and its desired label. A label of 1 represents an insulting comment, while a label of 0 represents a neutral comment. For instance, two examples from the training data set are:

- Text: "Either you are fake or extremely stupid...maybe both...", Label: 1
- Text: "We afford what we HAVE to afford, Marco.", Label: 0

In the training data, total 1049 of the examples are labeled as "insulting", while the remaining 2898 examples are labeled as "neutral". In the testing data, total 693 of the examples are labeled as "insulting", while the remaining 1954 examples are labeled as "neutral".

# 6 Why are you the right person/team for implementing this plan?

We are doing this project for our machine learning course. We are learning machine learning techniques and how they can be employed in different applications. To grasp these concepts cogently, we need to implement some of these techniques in a practical work. Through this project, we will be able to learn how ensemble classifier works and some of the well established machine learning techniques like support vector machine or logistic regression to handle text data. With the implementation of this project, we will improve our breadth in the field of machine learning.

### References

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