Extreme Gradient Boosting Based Question Similarity Predictor

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1.INTRODUCTION

Nowadays question answer websites like Quora, stack overflow, Yahoo Answers, Google scholars have become a very popular medium to search questions for laymen as well as domain experts. Many people ask similar worded questions multiple times. Multiple questions with the same intent can cause users to spend more time seeking the best answer for their question. Similarly, writers should not feel the need to answer the same question multiple times. We have used a machine learning based approach for searching questions on the website. The main problem in these tasks is to identify that the questions asked are duplicates of questions that have already been asked. This could be useful to instantly provide answers to questions that have already been answered. We are tasked with predicting whether a pair of questions are duplicates or not.

2. LITERATURE SURVEY

We've Compared Multiple Models, but the Random Models is the worst-case scenario. The test Log-Loss using Random Model is 0.8865. whereas when the data is linearly separable i.e., while using Logistic Regression we encountered test Log-Loss of 0.4237. In linear support vector machines where the data is separated using 2 plains, we got a test Log-Loss of 0.5078.

3. DATA ACQUISITION

We have used the Quora data set for implementing machine learning algorithms. This data set has more than 4 lakh data points. Each data point consists of two questions and a binary value which states whether they are similar or not.

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ebsites. In this paper we have compared multiple machine arning algorithms to predict similar questions using real		id c	jid1	qid2	question1	question2	is_duplicate
	0	0	1	2	What is the step by step guide to invest in sh	What is the step by step guide to invest in sh	0
	1	1	3	4	What is the story of Kohinoor (Koh-i-Noor) Dia	What would happen if the Indian government sto	0
	2	2	5	6	How can I increase the speed of my internet co	How can Internet speed be increased by hacking	0
	3	3	7	8	Why am I mentally very lonely? How can I solve	Find the remainder when [math]23^{24}[/math] i	0
ay Words, Machina Laarning Linear Degression	4	4	9	10	Which one dissolve in water quikly sugar, salt	Which fish would survive in salt water?	0

Chart -1: Example Points from Dataset

4. EXPLORATORY DATA ANALYSIS

The data set is imbalanced, 63% of the data points are not similar i.e., more than 2.5 lakh question pairs are not similar.

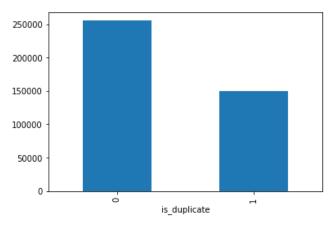


Chart -2: Imbalance Dataset

5. PRE-PROCESSING

5.1 Data Cleaning

In the process of cleaning of data, we have included steps like

- Removing of stop words
- **Removing HTML tags**
- **Removing punctuations**
- **Expanding Contractions**
- **Performing Stemming**
- Filling N.A. Values

5.2 Feature Extraction

We have constructed new features based on the length and frequency of the questions and number of words contained in the question. We also have appended features which are a derivative of above features. we have also created NLT and fuzzy features^[1].

5.3 Featurization

Featurization is a way to change text data into numerical vectors. Questions are represented in TF idf weighted word2vec format. We have featurized each question into 300 vectors. Therefore, in total we have 600 features.

6. MODEL TRAINING, EVALUATION AND INTERPRETATION

In total we have 794 features and one binary output feature for training the model. We have done a random train test split of 70:30 where the training data has equal class distribution. The key performance indicator used is log-loss. lower is the loss, better is the performance of the model.

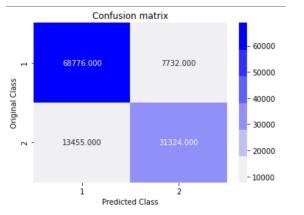
$$H_p(q) = -\frac{1}{N} \sum_{i=1}^{N} y_i \cdot log(p(y_i)) + (1 - y_i) \cdot log(1 - p(y_i))$$

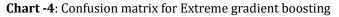
Binary Cross-Entropy / Log Loss

Chart -3: Formula to Calculate Log Loss

We have tuned the hyper parameters (alpha) in the range of 10^{-5} to 10^{1} .

This XGboost^[5] model is trained with an objective as "binary logistic". whereas the maximum depth of the tree is set to 4 and maximum rounds to stop training is 400. the train and test Log-Loss after 400 rounds is 0.3511 and 0.3517 respectively.





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Nebsite Output search Get Question results Create Question Merge pairs Question & Answers Data Cleaning Server Sort Results Feature Extraction Filter Results Featurization lodel F Predict Similarity Chart -5: Swimlane Diagram of Implemented System

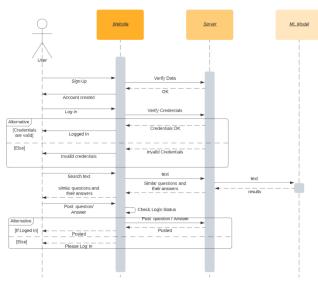


Chart -6: Sequence Diagram of Implemented System

7. VISUAL UNDERSTANDING OF SYSTEM



8. RESULT AND ANALYSIS



Chart -6: Screenshot of website with search results.

The implemented system performs superior to the model implemented with logistic regression or linear SVM as the log loss for the implemented system using Extreme Gradient Boosting is less.

9. CONCLUSIONS

Thus, we have implemented the above system on the website using Flask and we have achieved satisfactory results on the search question.

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