

Here we represent our feature extractor and classical classifiers

Available Classifiers

- Naive Bayes
- Logistic Regression
- XGBoost

To Train

1. Download [Quora Question Pairs Dataset](#).
2. Place it into `dataset` directory in the parent directory.
3. Install imblearn.

```
pip install imblearn
```

4. extract features from data

```
python main.py --do_data
```

5. Train the classifier

```
python main.py -classifier <classifier version (naive_bayes,  
logistic_regression, xgboost)>
```

e.g. to train `naive_bayes` classifier

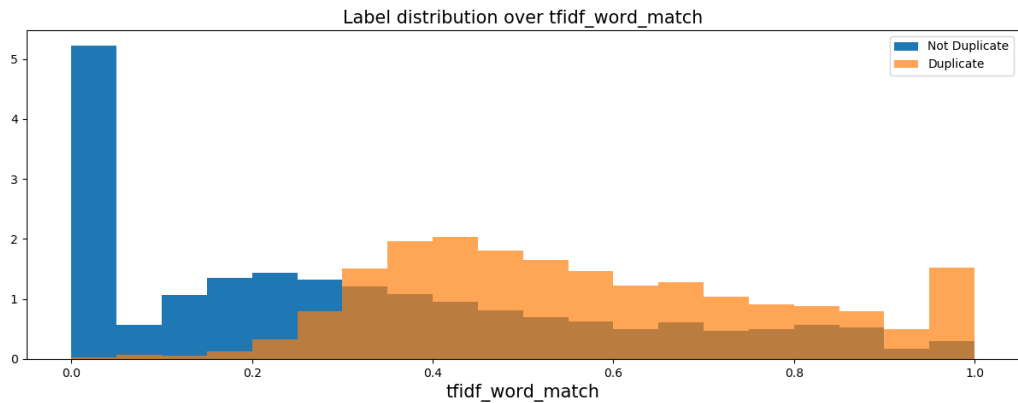
```
python main.py -classifier naive_bayes
```

Feature Extractor Details

Set of Features

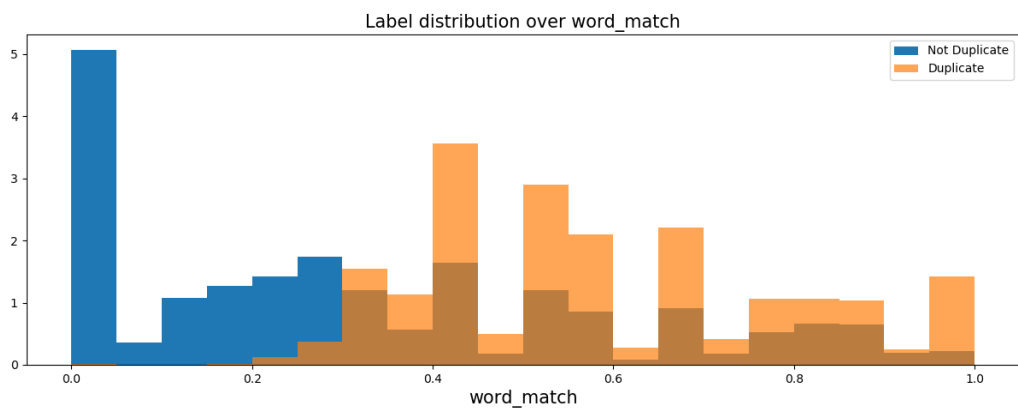
1. **Tfidf on shared words** For each question pair,
 - Stop words are removed.
 - Shared words are extracted.
 - Tfidf feature is extracted from the shared words only.

- This allows not to be biased to the common words that are more likely to be shared, because of their commonalities.




2. Word Match Share For each question pair,

- Stop words are removed.
- Shared words are extracted.
- Ratio of shared words is calculated, $R = (2 * \text{number_of_shared_words}) / (\text{number_of_words_in_question1} + \text{number_of_words_in_question2})$.




3. Jaccard For each question pair,

- Stop words are removed.
- Shared words are extracted (intersection).
- All set of words in both questions are extracted (union).
- Ratio of shared words is calculated, $R = \text{number_of_shared_words} / \text{number_of_union_words}$. 









4. Word Count Difference For each question pair, absolute difference between number of words in questions is calculated.

5. Word Count Ratio For each question pair,

- Word Count of both questions are calculated.
- Ratio of Counts is calculated, $R = \text{min_word_count} / \text{max_word_count}$. 

6. Unique Word Count Difference For each question pair, absolute difference between number of unique words in questions is calculated.

7. Unique Word Count Difference without Stop words For each question pair,

- Stop words are removed.
 - Absolute difference between number of unique words in questions is calculated. 
8. **Word Match Count** For each question pair, number of shared words is calculated. 
9. **Unique Word Count** For each question pair, number of unique words from both questions is calculated. 
10. **Unique Word Count Ratio** For each question pair,
- Unique Word Count of both questions are calculated.
 - Ratio of Counts is calculated, $R = \text{min_word_count} / \text{max_word_count}$. 
11. **Unique Word Count Ratio without Stop words** For each question pair,
- Stop words are removed.
 - Unique Word Count of both questions are calculated.
 - Ratio of Counts is calculated, $R = \text{min_word_count} / \text{max_word_count}$. 
12. **Same Start Word** For each question pair, check whether both questions start with the same word or not. 
13. **Character Count Difference** For each question pair, absolute difference between number of characters in questions is calculated. 
14. **Character Count Ratio** For each question pair,
- Character Count of both questions are calculated.
 - Ratio of Counts is calculated, $R = \text{min_character_count} / \text{max_character_count}$ 




Feature Selection

We applied a **Sequential Backward Selection** Approach to select the best representative features, and it ended-up with these features:

- Word Match Share
- Word Count Difference
- Word Count Difference without Stop words
- Unique Word Count Difference
- Unique Word Count Difference without Stop words
- Unique Word Count
- Unique Word Count Ratio
- Same Start Word
- Character Count Difference
- Character Count Ratio

Classifiers

We splitted the dataset as **90% training** and **10% validation**, and here are the results

- **Naive Bayes** Training Accuracy: 59% Validation Accuracy: 58.8% Validation AUC: 0.612 ROC Curve: image
- **Logistic Regression** Training Accuracy: 70.27% Validation Accuracy: 70.46% Validation AUC: 0.783 ROC Curve: image
- **XGBoost** Training Accuracy: 78.26% Validation Accuracy: 78.32% Validation AUC: 0.874 ROC Curve: image