Classification models on subreddits:'AskWomen' and 'AskMen'

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Yonghe



Only one post came from "AskMen"

Our final model can predict them all correctly

Data Collection

- Fetched 2,000 posts from each subreddit
- Only consider columns 'title' and 'selftext'
- Balance Dataset

NLP

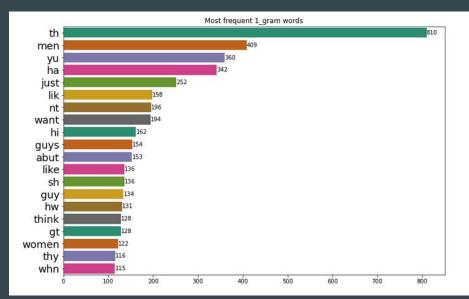
- Tried Count Vectorizer and TF-IDF Vectorizer
- Tried different NLP techniques during hyperparameter tuning
 - Tokenizer:
 - Regex_tokenizer r'(?u)\b\w\w+\b'
 - Regex_tokenizer + Lemmatizer
 - Regex_tokenizer + Stemmer
 - Stop_words: [None, 'english']
 - Ngram range: (1,1),(1,2),(1,3)
 - o max_features: [400,500,600,700,900]

Most Common 1_gram Words (excluding stopwords)

AskWomen

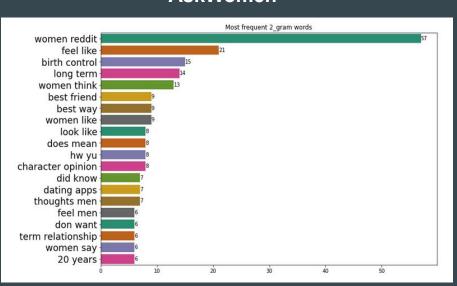
Most frequent 1_gram words women like feel did think man men does partner guy just relationship reddit th know want woman things date 200 250 300

AskMen

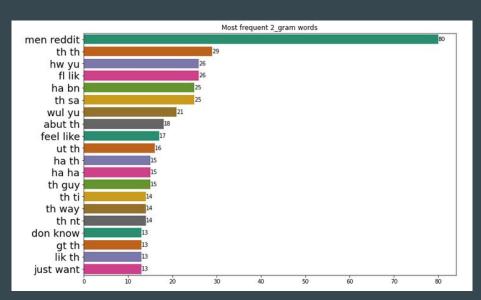


Most Frequent 2_gram Words (excluding stopwords)

AskWomen



AskMen



Top 200 Most Common Words

AskWomen

AskMen



- Abbreviations occur more frequently in 'AskMen' than 'AskWomen'
- Many abbreviations in 'AskMen' are stopwords, such as 'whn', 'hw'
- Comparing abbreviations and stopwords will be critical to classify these two subreddits. We can verify it thought feature importance analysis

Metrics

- Roc_auc_score
 - baseline score: 0.5

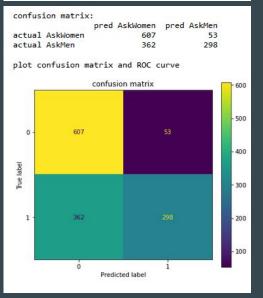
- Accuracy score:
 - baseline score: 0.5 (balance dataset)

perc_diff < 5%

Model 1: CountVectorizer + RandomForest

```
roc_auc_score on training set: 0.837
roc_auc_score on testing set: 0.819
perc_diff: 2.2 %

accuracy_score on training set: 0.705
accuracy_score on testing set: 0.686
perc_diff: 2.8 %
```



```
best_params:
    classifier__max_depth : 6
    classifier__n_estimators : 500
    tvec__max_features : 350
    tvec__ngram_range : (1, 3)
    tvec__stop_words : None
    tvec__tokenizer : None
```

Metrics:

- roc_auc_score increase from 0.5 to 0.82,
 - o perc_diff: 2.2%
- accuracy score increases from 0.5 to 0.68
 - o perc_diff: 2.8%

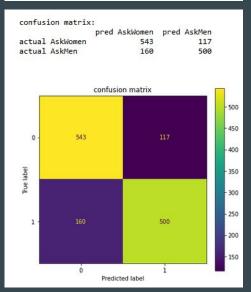
Observations of best_params:

- Only use 350 features
- Better result without removal of stopwords
- Better result without lemmatizing or stemming

Model 2: TfidfVectorizer + Logistic Regression

```
roc_auc_score on training set: 0.859
roc_auc_score on testing set: 0.851
perc_diff: 0.9 %

accuracy_score on training set: 0.785
accuracy_score on testing set: 0.790
perc_diff: 0.7 %
```



```
best_params:
    classifier__C : 1
    classifier__penalty : 11
    tvec__max_features : 1000
    tvec__ngram_range : (1, 3)
    tvec__stop_words : None
    tvec__tokenizer : None
```

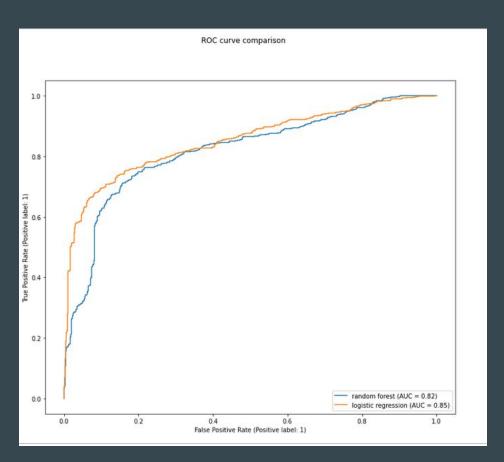
Metrics:

- Roc_auc_score increase from 0.82 to 0.85
 - o perc_diff: 0.9%
- accuracy_score increases from 0.68 to 0.79
 - o perc_diff: 0.7%

Observations of best_params:

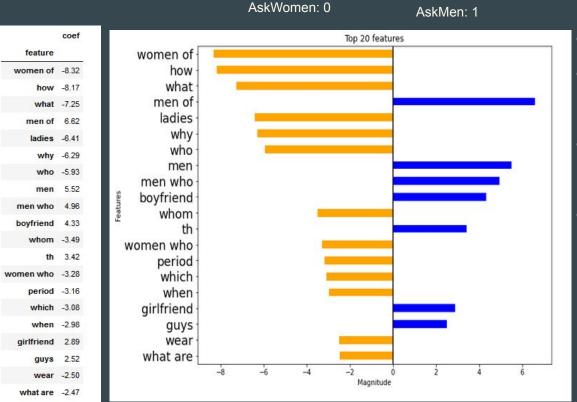
- Better result without removal of stopwords
- Better result without lemmatizing or stemming

Model performance comparison



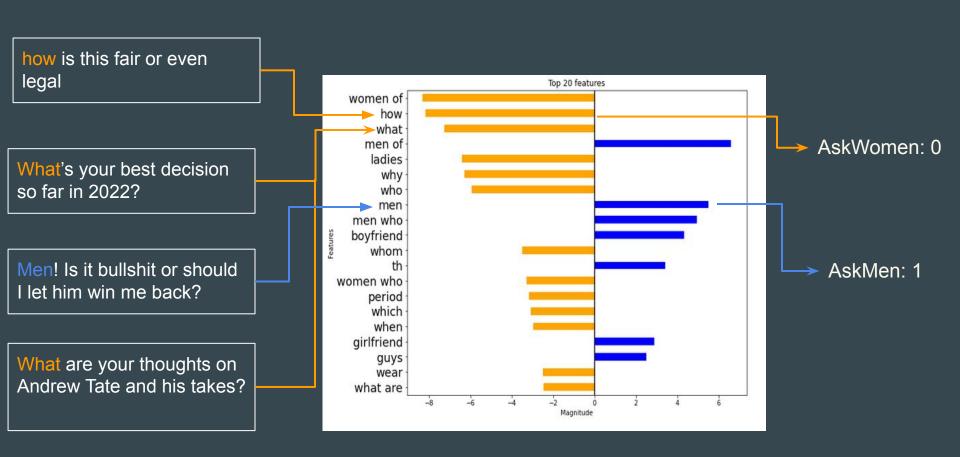
- Logistic Regression Model outperforms Random Forest Model in almost all thresholds
- Logistic Regression Model is chosen to be final model

Feature Importance Analysis for Logistic Regression Model



- We mapped "AskMen": 1, "AskWomen": 0
- Positive coefs, shown in blue color in the chart, indicate the features are more common in "AskMen"
- Negative coefs, shown in orange color in the chart, indicate the features are more common in "AskWoMen"
- Top features contains a lot of stopwords, mostly because abbreviations of stopwords such as 'whn','hw' occurs much more frequently in "AskMen".

Important Feature Demonstration



Q & A