



GA DSI 26


Project 3:

Wine and Beer

By: Lim Zhi Yong

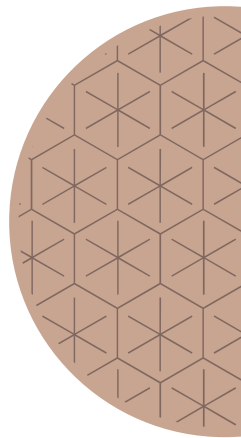


Task:

- Understand consumer patterns
 - Identify if the consumer wants winemaking or homebrewing info
 - Train model with subreddit posts
- 

Data Description

- 1,000 posts from each subreddit
 - r/winemaking
 - r/homebrewing
- Cleaned punctuation, stopwords, delimiters
- Considered both unigrams and bigrams



Notes

“wine”

“Wine” was top classified word for winemaking, but second in misclassified posts

Seeking advice

'first time' comes up relatively frequently

Usual suspects

- hop', 'malt', and 'keg' for beer
- 'grape', 'skin', 'age' for wine

Types

Wine has more types (strawberry, elderberry, banana) than beer (pale ale, ginger) in top 20

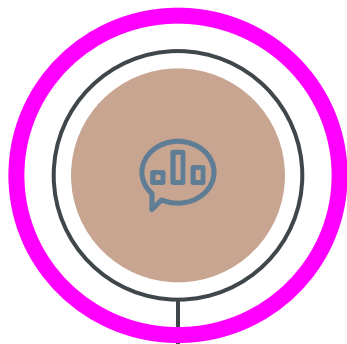
Overlapping words

- Sugar
- Yeast
- Ferment

Tokenize

Bigrams had more unique tokens than unigrams

Model Workflow



Data Cleaning

- Missing values
- Vectorization



Modelling

- Building models



Testing

- Scoring models



Recommendations

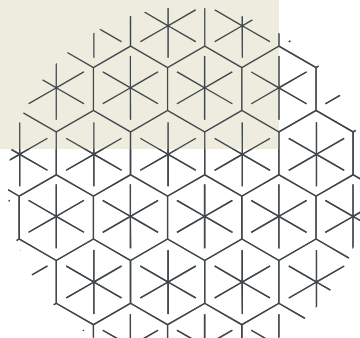
- Important features

Missing Values

There are different types of missing values:

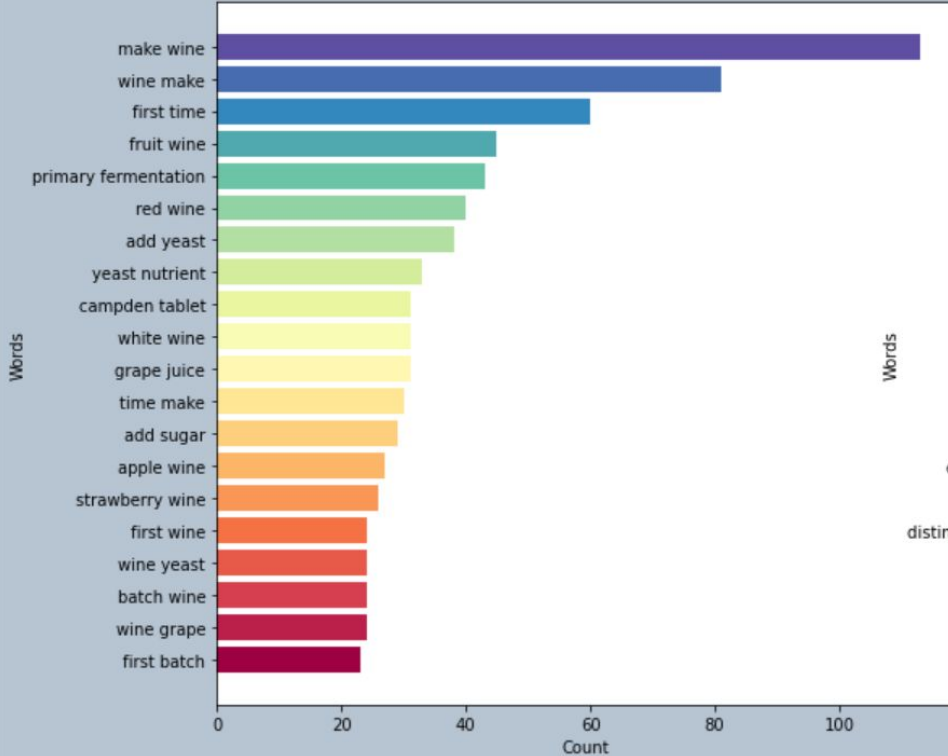
- Duplicate posts were removed
- Null and removed texts were replaced with the empty string
- One deleted post was miscategorized, we removed it

1969 rows left

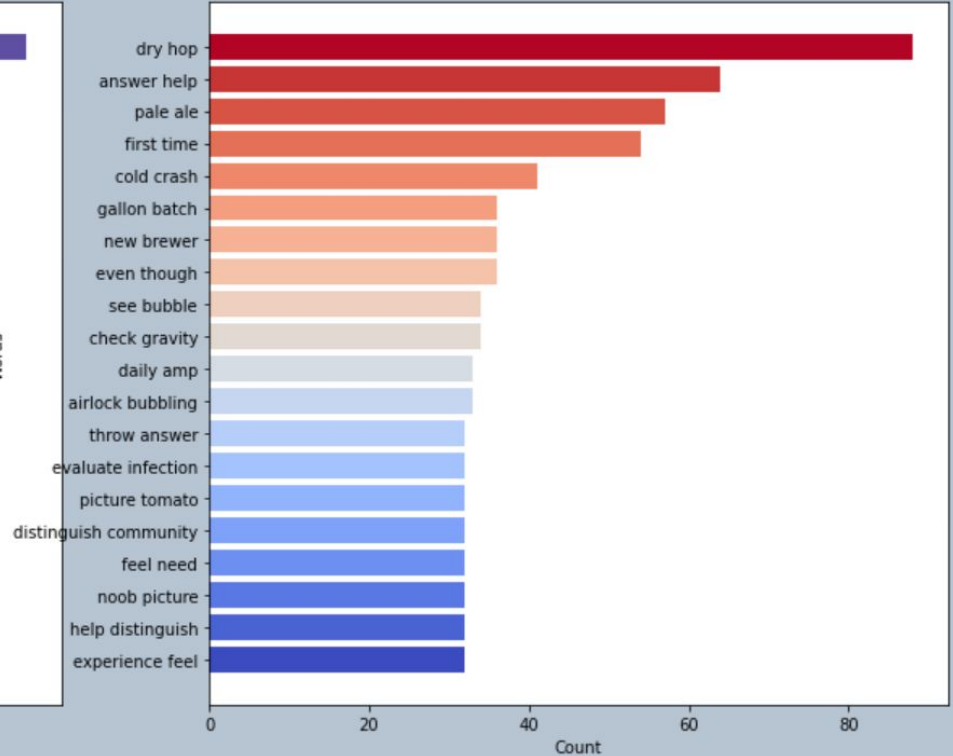


Plots of most frequent bigrams in the subreddits

Top 20 words in winemaking subreddit titles

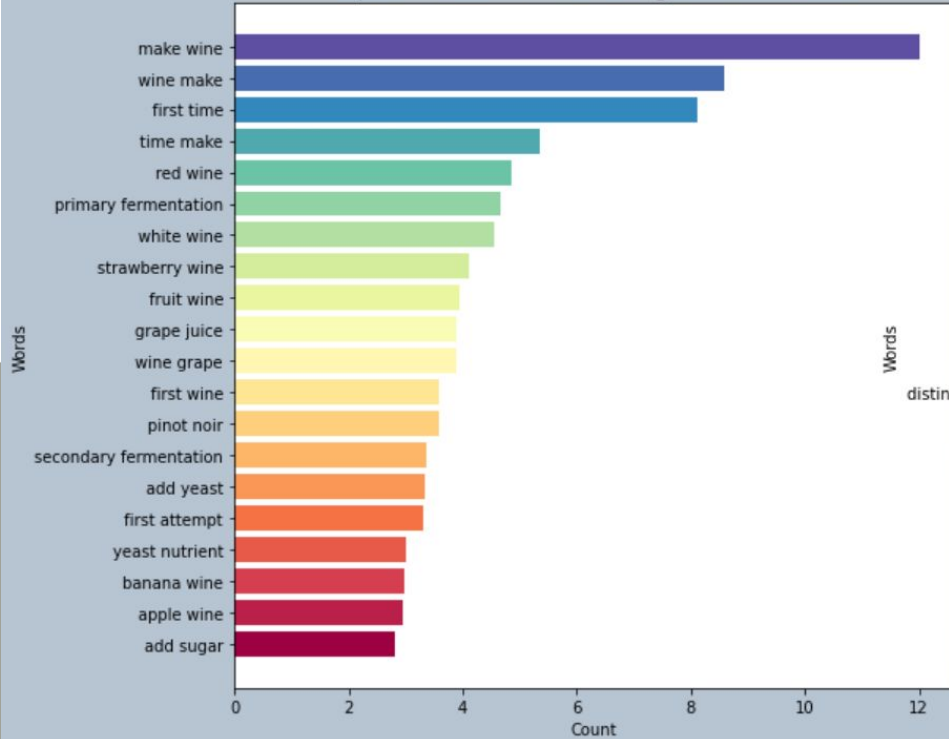


Top 20 words in homebrewing subreddit text

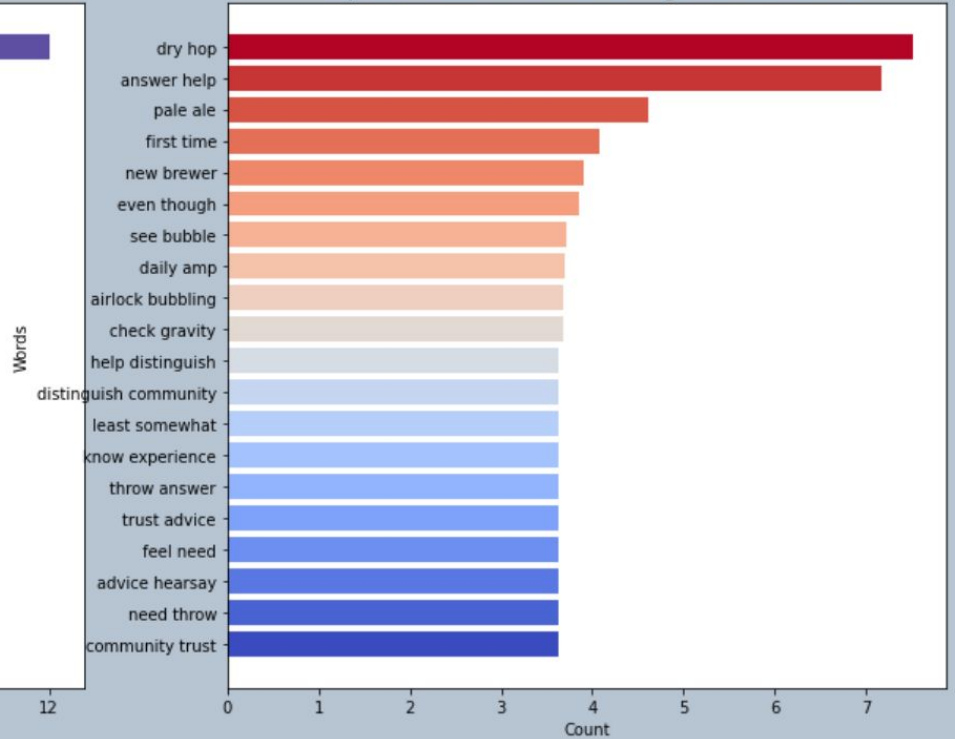


Plots of most frequent tf-idf bigrams in the subreddits

Top 20 words in winemaking subreddit titles



Top 20 words in homebrewing subreddit text

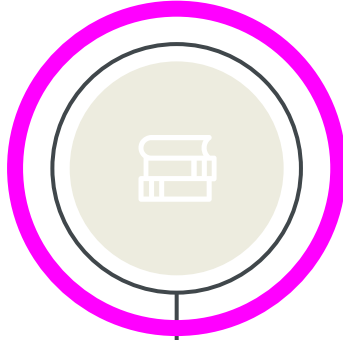


Model Workflow



Data Cleaning

- Missing values
- Vectorization



Modelling

- Building models



Testing

- Scoring models



Recommendations

- Important features

Modelling

- ❖ 8 models:
 - Logistic regression
 - Count
 - Tf-idf
 - KNN classifier
 - Count
 - Tf-idf
 - Naïve bayes
 - Count
 - Tf-idf
 - Random forest
 - Count
 - Tf-idf

Scoring

- ❖ ROC-AUC to determine best models
- ❖ F1 score to compare baseline score
- ❖ Accuracy to determine whether overfit

Model Workflow



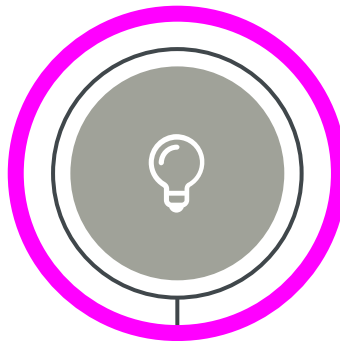
Data Cleaning

- Missing values
- Vectorization



Modelling

- Building models



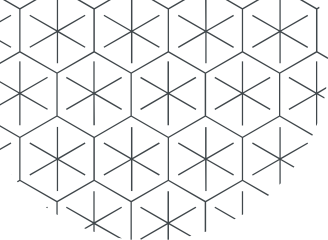
Testing

- Scoring models



Recommendations

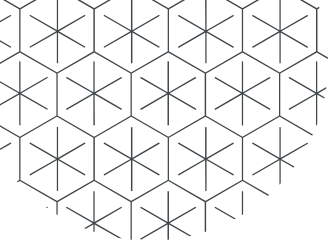
- Important features



models	vectorizer	accuracy score	auc score
Logistic Regression	count	0.897	0.961
Logistic Regression	tf-idf	0.917	0.975
KNN Classifier	count	0.720	0.851
KNN Classifier	tf-idf	0.580	0.655
Naïve Bayes	count	0.789	0.925
Naïve Bayes	tf-idf	0.789	0.925
Random Forest	count	0.890	0.964
Random Forest	tf-idf	0.888	0.966

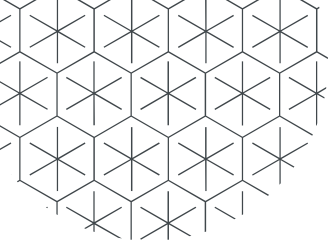
The background features several abstract, organic shapes in muted colors: a large brownish-orange shape on the left, a large light beige shape in the center, and a grey shape on the right. A small, stylized black line drawing of a leafy branch is positioned at the top center. The text is centered over the beige shape.

**Logistic Regression
performed the best**



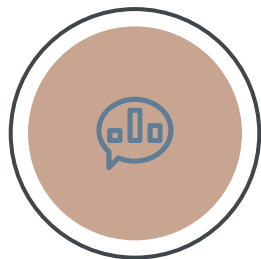
Pipeline params:

- **TfidfVectorizer(max_features=4000, min_df=2, ngram_range=(1, 2))**
 - **LogisticRegression(C=1, random_state=42, solver='liblinear')**



0.97 ROC-AUC
0.92 accuracy

Model Workflow



Data Cleaning

- Missing values
- Vectorization



Modelling

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Recommendations

- Important features

Best Model

Features

- Wordnet lemmatizer
- Tf-idf vectorizer
- Logistic regression with ridge penalty

Limitations

- Spell check before lemmatizing
- Slightly overfit, could remove more stopwords



Thanks

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