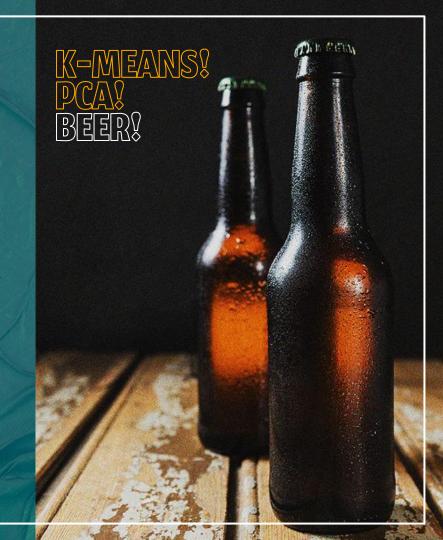
// Spring 2021

# COGS 118B FINAL PROJECT

Analysis of Beer Reviews as Predictors of Beer Type



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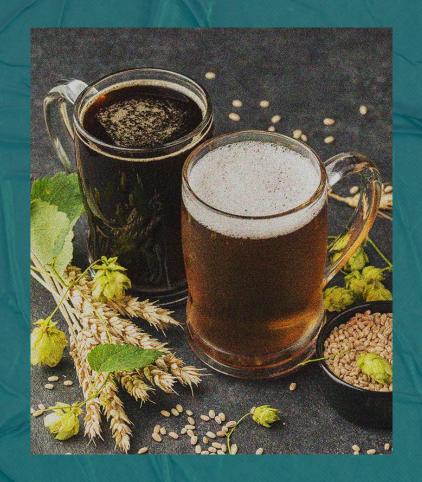
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**DISCUSSION** 

Limitations with our analysis and next steps

# OUR MOTIVATION

- Thousands of years of history
- Brewing method informs taste,
   ABV, color, hoppiness and more
- Lots of review data online
- Gives insight into most valued features in types of beer



## RELATED WORK



### PCA + K-means

Class lectures aided in application of PCA and implementation of K-means grouping

# CHEERS



### **RoboBEER**

Beer quality assessed through ML/CV-aided color, foamability and foam stability



	Percent of Reviews
Dark ale	20.68%
Pale ale	28.85%
Lager	15.01%
Stout	16.10%
Other   	19.36%

# **METHODS: Data Cleaning**

First, we split 100+ unique beer styles into 5 major categories:

- Dark Ales (1)
- Pale Ales (2)
- Lagers (3)
- Stouts (4)
- Other (0)

This split the beer by brewing method, separating the data by color & ABV

# **METHODS: Review Data**

TASTE







**AROMA** 

**PALATE** 

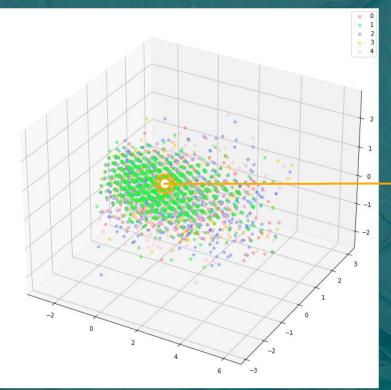


**OVERALL SCORE** 



**APPEARANCE** 

# **METHODS: Principal Component Analysis**



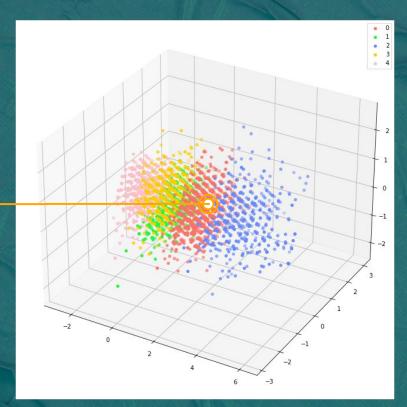
Using the sklearn package, we applied PCA to the cleaned dataset.

Unfortunately, each group has equal spread over the graph, with category 2 showing in the highest layer.

## **METHODS: K-means**

We implemented K-means from scratch, using lecture slides as a reference.

After applying the function to the PCA-processed data, it yielded the following graph.



## RESULTS

- K-means clusters only matched with our 5 categories 20% of the time
- This implies random selection, demonstrating no correlation between beer reviews and beer category
- Utilizing PCA & K-means was not enough to uncover a relationship between reviews and beer
- This result is useful for brewers and sellers, allowing them to focus on the craft and marketing without worrying about the brewing method



# **DISCUSSION: Limitations**







#### **CLASSIFICATION**

A more rigorous classification system could have improved groups

#### MORE FEATURES

More numerical features or encoding of extra categorical data could have led to more conclusive results

#### **PLOT CHANGES**

3-D plots are complex, and rotation features could help better interpret data

## **DISCUSSION: Future Work**

- A recommendation system utilizing unsupervised learning could be ideal
- Users could receive recommendations for new beers they'd like based on reviews
- Brewers could analyze the most popular beers and styles to inform future brewing and marketing choices



