Sommeliers and BERT: Grape Variety Classification

The Process

01 02 03

Problem Collection Exploration

06 05 04

Conclusions Modeling Theory

01 Problem

Text to Tool

Problem Statement

Can Sommelier reviews be used as predictive features for a multiclass classification model to distinguish specific grape varieties as a means to educate those new to wine and facilitate growth as a wine consumer?

02 Collection

Scraping winemag.com

Reviews

There's a mineral-laced start to the nose of this bottling, with hints of asphalt and graphite, followed by darker stewed fruit on the back. The palate is layered in creamy, lavish tannins, as flavors of cassis and caramel aim to impress through the clean acidity. —MATT KETTMANN

RATING 92

POINTS

PRICE \$45, BUYNOW

DESIGNATION Reserve

VARIETY <u>Cabernet Sauvignon</u>

APPELLATION Paso Robles, Central Coast,

California, US

WINERY Smith & Hook

03 Exploration

Overall and Text

Exploration

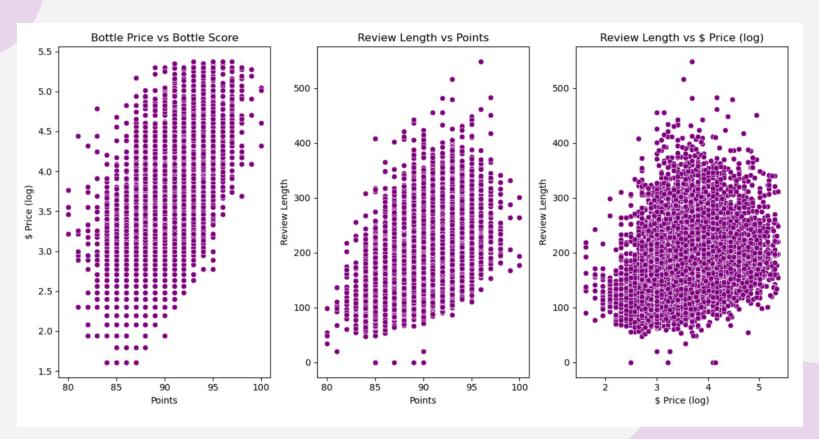
Overview

- Price
- Points
- Countries
- Regions
- Wineries

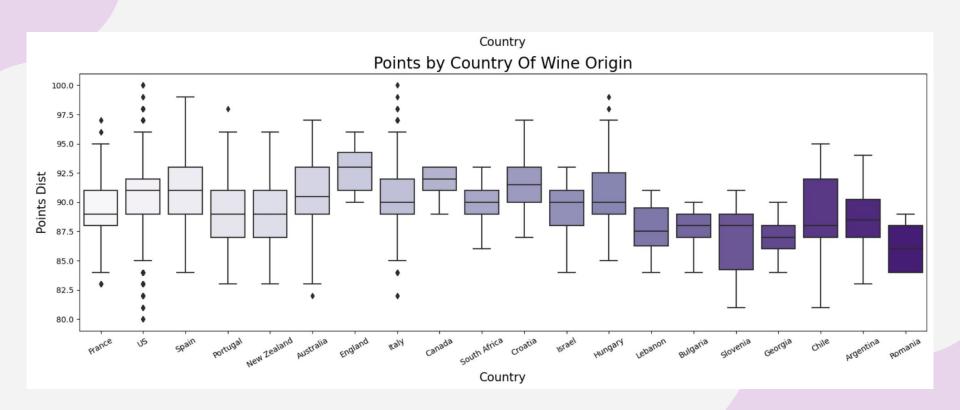
Focus

- Reviews
- Varieties

- Unique Sommeliers: 18
- Unique countries: 23
- Unique varieties: 328
- Unique wineries: 4496
- Unique bottles: 13782



Points	Country	Bottle	Variety
95	Spain	Rolland & Galarreta 2014 Tempranillo (Rioja)	Tempranillo
95	Australia	Chambers Rosewood Vineyards NV Muscat (Rutherglen)	Muscat Blanc à Petits Grains
95	Italy	Paltrinieri 2020 Radice (Lambrusco di Sorbara)	Lambrusco di Sorbara
94	Australia	El Vinculo 2018 Crianza Tempranillo (La Mancha)	Tempranillo



Points	Country	Bottle	Variety
92	Hungary	Dúzsi Tamás 2020 Kékfrankos Rosé (Szekszárd)	Rosé
91	Hungary	Royal Tokaji 2019 The Oddity Dry (Tokaji)	Furmint
91	Hungary	Gál Tibor 2020 Egri Csillag White (Eger)	White Blend

Focus

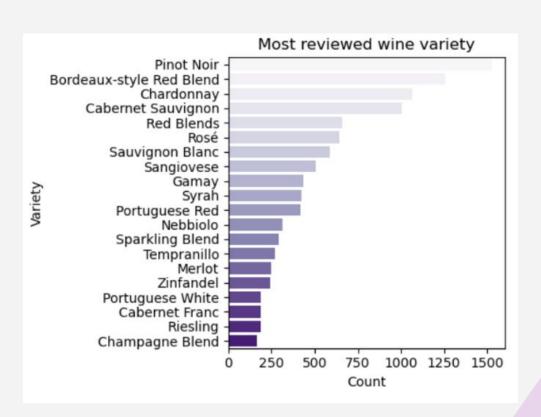
Varieties

- Unique wine varieties: 328
- Limit to those with most reviews for data
- Remove "grouped" varieties(ie: blends)

Text

 Is there information to suggest that wine varieties can be distinguished from one another?

Focus



04 Theory

Vectorization and BERT

Text Vectorization

Conversion of text to numerical values for model integration

Statistics based

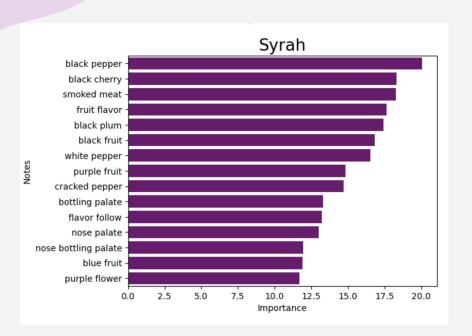
- OHE:
- Count Based:

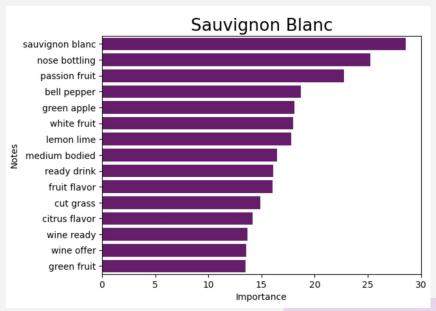
Predictive Based

- Sequential RNN:
- Transformers:

Focus

TF-IDF Vectorization





BERT

"Bidirectional Encoder Representations from Transformers"

- Stacked Transformers
- Better Contextual Relations
- Pretrained
- Fine-Tuning

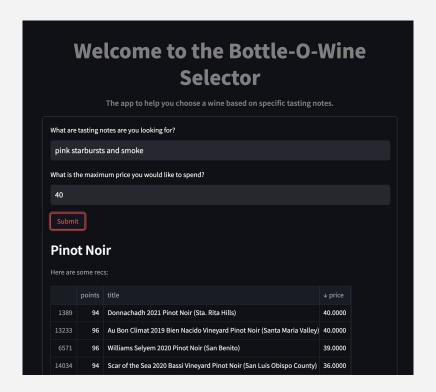
05 Models

Statistical, Untrained and Pre-trained

Models

Model	Train Accuracy	Test Accuracy
Multinomial Naive Bayes	0.908	0.819
Sequential RNN	0.824	0.7026
BERT	0.987	0.8952

DEMO



06 Conclusions

Improvements and Further Research

Conclusions

- BERT model increased accuracy by almost 10% from Naive Bayes
- Exponential increase of computing power and memory
- Some data leakage occurred since names of the grapes were mentioned within the review.
- Good use case established for continued development of an app

Further Research

- Improve model by adding more data to expand classifying to 300+ types
- Try out different pre-trained models
- Use clustering grapes before variety classification
- Add additional features like grape color (red, white, orange)

Thanks Questions?