THENEXT DRINK'S ON THEM!

Recommend their next **favorite** bottle of wine for customer happiness and retention



THE PROBLEM

- Lack of customer loyalty
- Few repeat purchases
- Recommendation system failing

THE ROOT CAUSES



Collaborative recommender system

Niche preferences are overlooked



Customers aren't usually experts

Hard to tell from just a label if you'll like it



THE SOLUTION



Content-based recommender system

Customers have unique preferences



Beginners welcome

Customers don't have to know **why** they like a bottle



Concrete features

Varietal, flavors, dryness/sweetness, body



THE IMPACT



Customers love their wines!

Expert suggestions build trust

Repeat bottle purchases

Repeat purchases per customer Customers making repeat purchases

Customer retention

1

Purchases per customer

WHAT'S IN A WINE?

• Varietal isn't everything

 Analysis performed on Kaggle dataset of 129,000+ expert

wine reviews









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- Grab nouns & adjectives
- Standardize & lemmatize





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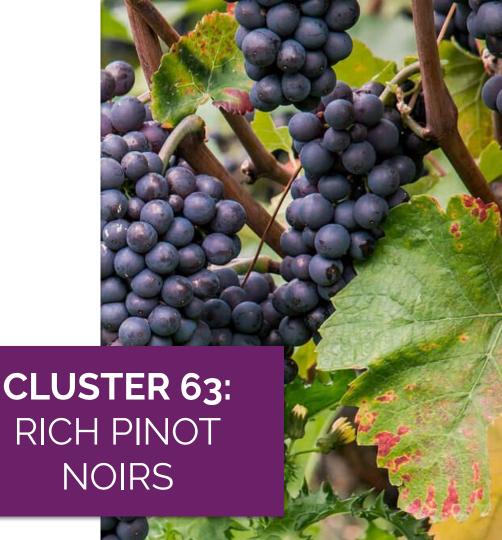
stylecloud

Understand the wines in a given cluster





STRUCTURE







THE MODEL



At least one 'like'

Most common 'liked' cluster

Most recent 'liked' cluster

1-3 random recommendations from chosen cluster

"Fine acidity and a bracing **minerality**, like a lick of cold **steel** on a winter day, mark this Chardonnay. It has oak notes, but they're wisely kept in the background, allowing the citrus, **green** apple and peach flavors to star. This beautiful wine continues the winery's string of great reserve Chardonnays, and is their best since the 2006 vintage."





"An initially wood-laden wine that only slowly reveals the rich fruit character. There is wood certainly, but the spice integrates well with the **green** fruits and the intense **minerality**. The great **steely** backbone on this wine indicates it will age."





NUMBERS TO TWEAK

130

Total clusters

3

Recs per email

1

Cluster per email



FOR BEST RESULTS...



Data	Experiment	Focus groups	Inform	Encourage sign-ups
Good, reliable descriptions	A/B test recommender logic - i.e. wines from different clusters in same email	Test out model recommendations in person	Tell users why the wine is recommended	In-store iPads to recommend wines

Filter recommendations by price!

