



Capstone Project Coffee Shop

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Problem: What features most affect the ratings of coffee reviews?

Information benefactors:

- Widely known chains for purposes of creating new coffee blends
- Newly established coffee shops or regional chain coffee shops for creating signature blends
- Coffee roasters advertising their specialty coffee roasts
- Coffee bean sourcing companies determining what coffee bean to source

Data Sourcing and Cleaning

Data taken from a [Kaggle dataset](#) with 19 features and 7041 entries

- Numeric, descriptive, and narrative features

Missing Values:

- 'with_milk' feature deleted
- 'agtron' was a descriptive feature, but had hidden unusable data

Numeric Features changed from object form:

- Target feature: 'rating' (1-100)
- 'acidity_structure' (1-10)
- 'aftertaste' (1-10)
- 'aroma' (1-10)
- 'body' (1-10)
- 'flavor' (1-10)

	count	%
with_milk	6044	85.840080
acidity_structure	4875	69.237324
bottom_line	4080	57.946314
est_price	2039	28.958955
aftertaste	872	12.384604
coffee_origin	505	7.172277
roast_level	374	5.311745
aroma	50	0.710126
flavor	16	0.227240
body	11	0.156228
notes	8	0.113620
roaster_location	3	0.042608
blind_assessment	1	0.014203
title	0	0.000000
rating	0	0.000000
agtron	0	0.000000
review_date	0	0.000000
roaster	0	0.000000
url	0	0.000000

Commentary features dropped:

- title
- blind_assessment
- bottom_line
- notes
- review_date
- url

Other dropped features:

- acidity_structure
(high correlation with the rating, 62% null)
- agtron *(hidden null values)*
- est_price *(most common value only 2.2%)*

Features with null values replaced:

- aftertaste
- aroma
- body
- flavor
- coffee_origin
- roast_level
- roaster_location

Exploratory Data Analysis

Profile Report provided a summary, visualizations, and notes for each feature.

The report also raised 28 alerts:

high cardinality
features with high correlation
features with missing values
uniformly distributed features

Preprocessing:

Dummy variables for categorical features

StandardScaler to scale the magnitude of numeric features

80/20 train/test split

- Target feature: 'rating'

Original Model Metrics:

	Model	MAE	MSE	RMSE
0	Linear Regression	1.794951e+10	5.672739e+21	7.531759e+10
1	Random Forest	1.794049e-01	8.327751e-02	2.885784e-01
2	XGBoosting	1.920313e-01	8.294503e-02	2.880018e-01
3	SVR	1.828601e-01	8.128682e-02	2.851084e-01
4	Elastic Net	4.807005e-01	4.763833e-01	6.902053e-01

Modeling:

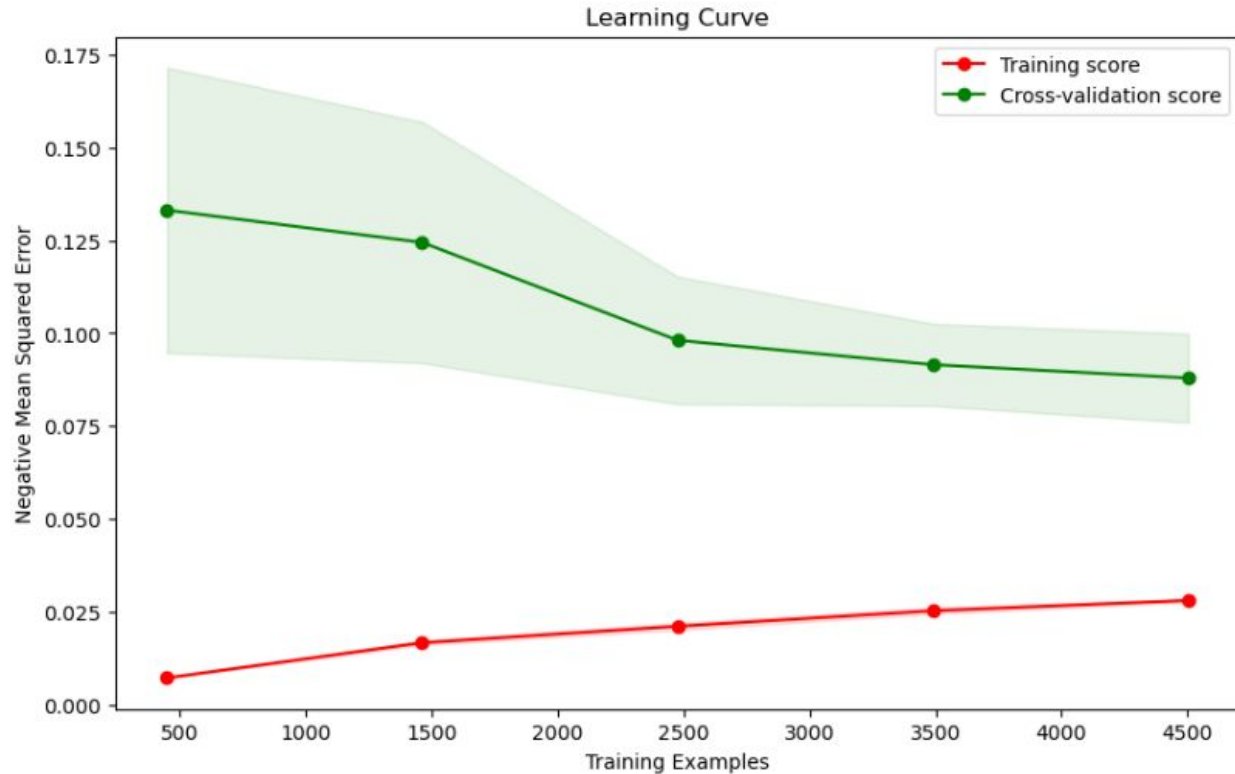
Models run with and without GridSearchCV():

- Linear Regression
- Random Forest
- Extreme Gradient Boosting (xgboost)
- Support Vector Regression (SVR)
- Elastic Net

Model Metrics with GridSearchCV():

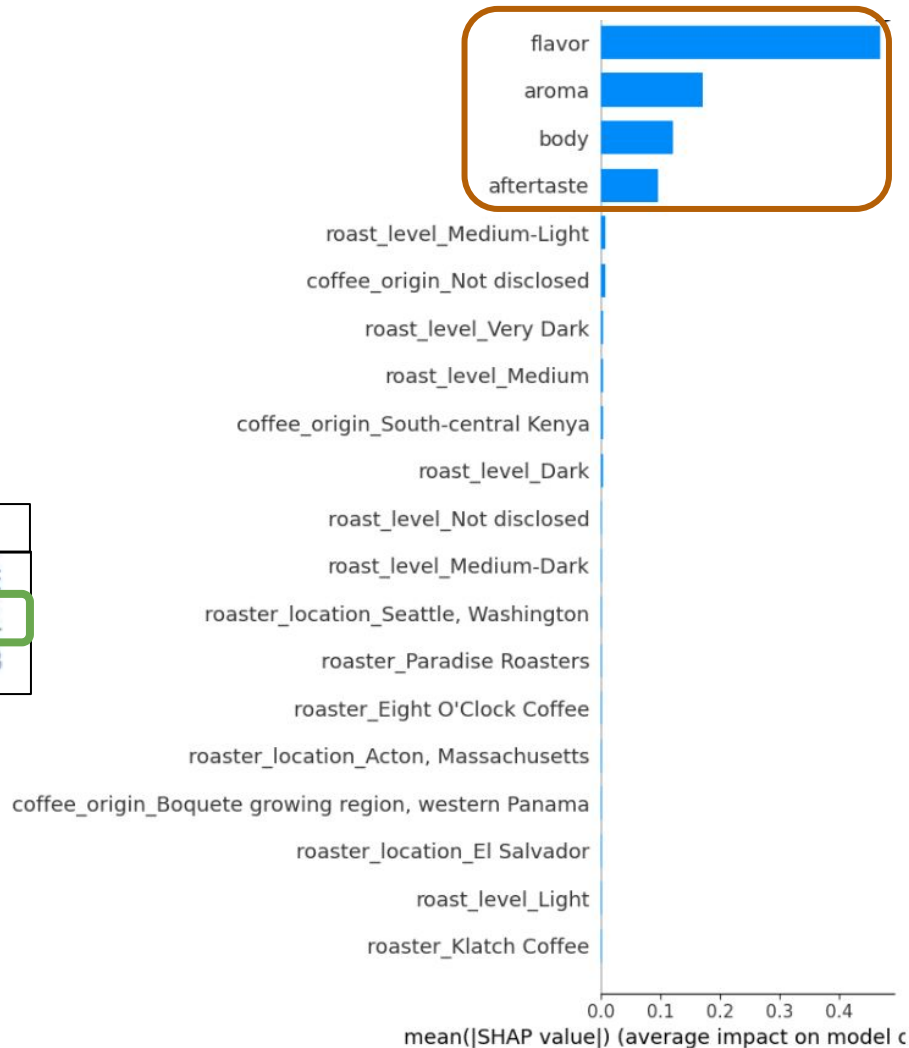
	Model with GridSearchCV	MAE	MSE	RMSE
0	Linear Regression	0.208727	0.095691	0.309340
1	Random Forest	0.184572	0.082398	0.287050
2	XGBoosting	0.184580	0.080327	0.283421
3	SVR	0.196410	0.084748	0.291114
4	Elastic Net	0.200381	0.090394	0.300656

Visualizations: Learning Curve



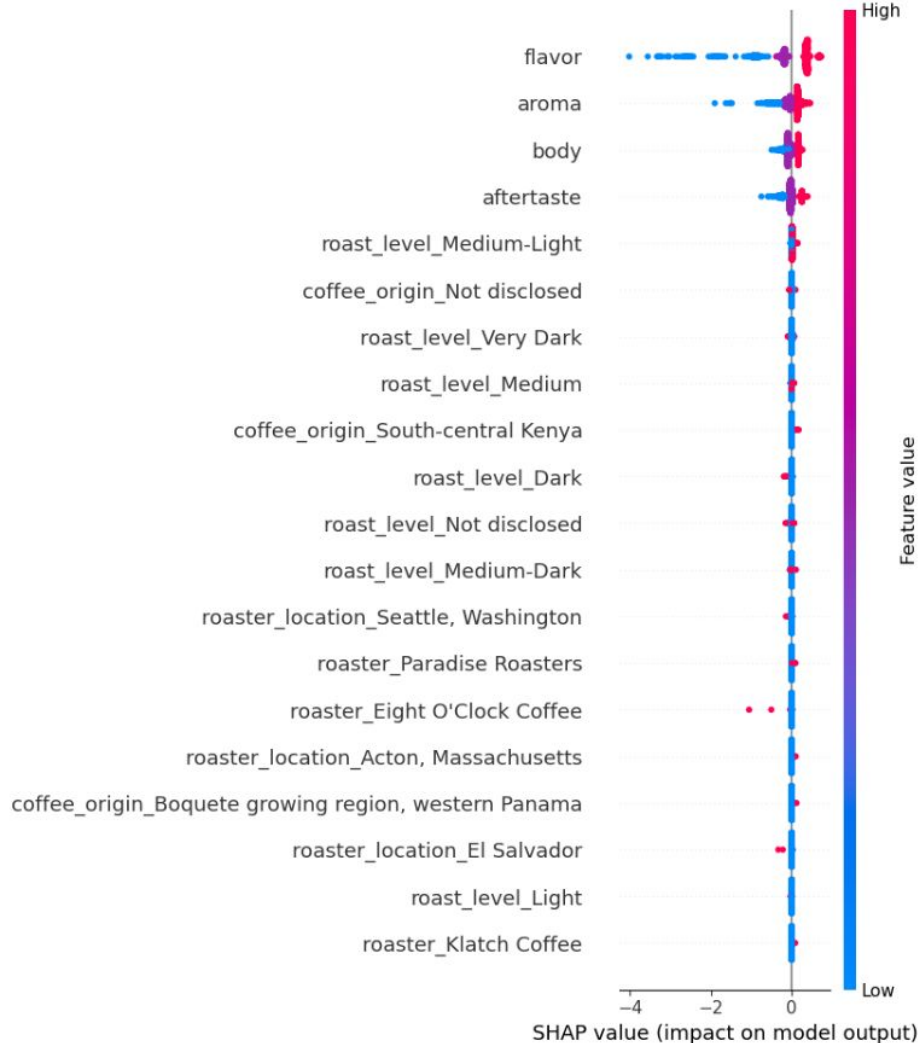
Visualization: SHAP Summary Bar Plot

Original Data vs. Focused DataMetrics:				
	XGBoost Model	MAE	MSE	RMSE
0	Original Dataset	0.184580	0.080327	0.283421
1	Focused Dataset	0.187503	0.083615	0.289163

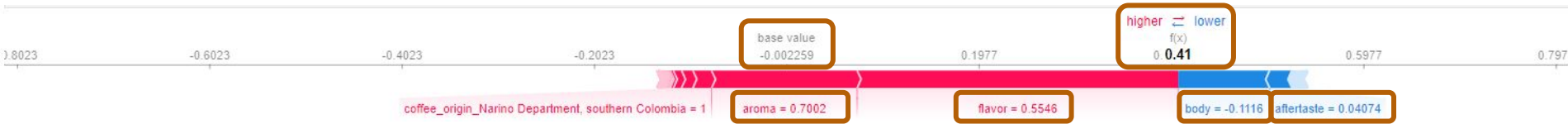


Visualizations: SHAP Summary Plot

Higher SHAP values have high positive impact on the rating for flavor, aroma, body, and aftertaste



Visualizations: SHAP Force Plot for Initial Review



- base value represents the predicted rating for all observations in the data set
- projected value represents the predicted rating for the initial review
- flavor and aroma are the two greatest components in the rating, both with a positive impact
- body and aftertaste are the next two greatest components in the rating, both with a negative impact

Next Steps:

Further Exploration of Key Features

Price Exploration

Top 4 features affecting rating:
flavor, aroma, body, aftertaste

- descriptive info needed
- parse commentary features
- new data set recommended

Price Exploration

- est_price: \$18.00/12 ounces
(2.2% of feature)
 - extensive data cleaning
 - new data set recommended
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