Linear Regression Predicting Rug Sale Prices

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Intro

Client:

- Design Allure
 - Interior design firm located in Portland, OR
 - The client is interested in learning about potential sale prices for one of their suppliers Rugs.com in order to inform purchasing decisions.

Objective:

• Build a linear regression model to predict sale prices based on rug features. The model with the highest accuracy will be selected.

Goal:

• Predict sale prices of the rugs on Rugs.com to allow the client to identify if it is optimal to purchase the rug at the current price.

The Data:

Product information scraped from Rugs.com using Selenium

Features: Original price, Country Made In, Collection,

Construction (is it handmade?), Indoor/Outdoor Use, Pile Height (height of the tufted loops), Shape, Area, Color, Materials.

Product Details

Designer Sale

Ends in 15:02:17

This rug is currently in stock

Free Shipping & Free Returns

Made in: Turkey

Collection: Tuscany

Pile Material: 100% Polypropylene | Pile Height: 1/2"

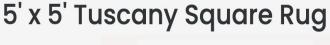
Brand new

The primary color is Black.

Colors include: Black, Green, Ivory, Orange, Red, Yellow, Blue.

Size FT: 5' 0" x 5' 0" | Size CM: 152 cm x 152 cm

SKU #: 6307842 | Currently in stock











Data Cleaning & Eda

- Remove outliers
- Convert categorical data to binary
- Bucket the data to reduce complexity
- Calculate area

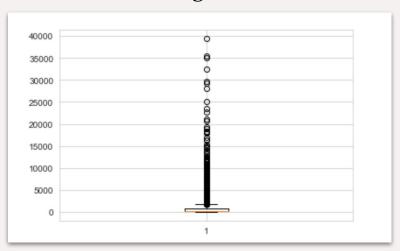


8' x 11' 4 Hand Knotted Nain Persian Rug

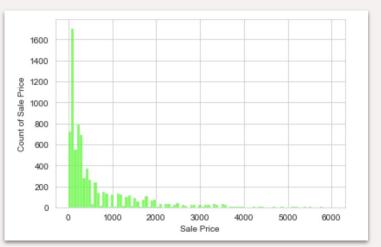
\$294,997 \$10,278 96% off

Data Cleaning & Eda

Box Plot of Rug Sale Price



Distribution of Rug Sale Price



Baseline

Linear Regression



Train:

R^2: 0.8516618639255865

RMSE: 238.7379637310733

MSE: 127.97789835244693

Test:

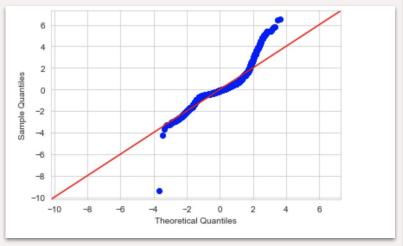
R^2: 0.839964585810739

RMSE: 251.09783772817403

MSE: 132.29850338388272

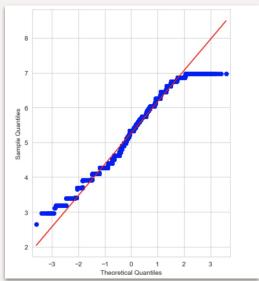
Well that's not what we want:

Diagnostic QQ Plot



After removal of additional outliers and

Box Cox Transformation:



Linear Regression with Box Cox Transformation of Target

Train:

R²: 0.8972325691001115

RMSE: 0.2795328447524849 MSE: 0.2148707656988781

Test:

R^2: 0.8978992083456274

RMSE: 0.27978609992977815

MSE: 0.21388048742505195



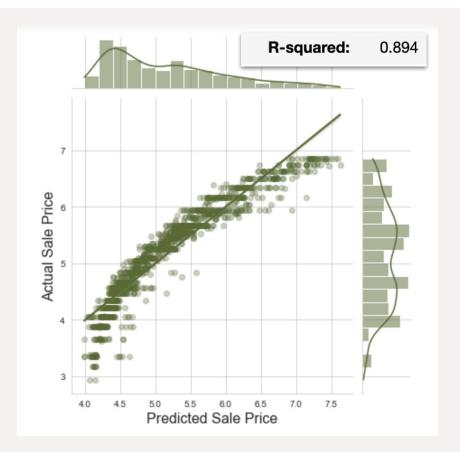
Reduce Model Complexity & Address Multicollinearity

- Remove features that are not adding value
 - Pile Height
 - Indoor/Outdoor Use
- Remove features with high VIF indicating multicollinearity issue



Final Model & Notes





The box cox transformation has the drawback of reduced interpretability.

Requires back transformation.

Future Work

