

Diamond Price Estimation

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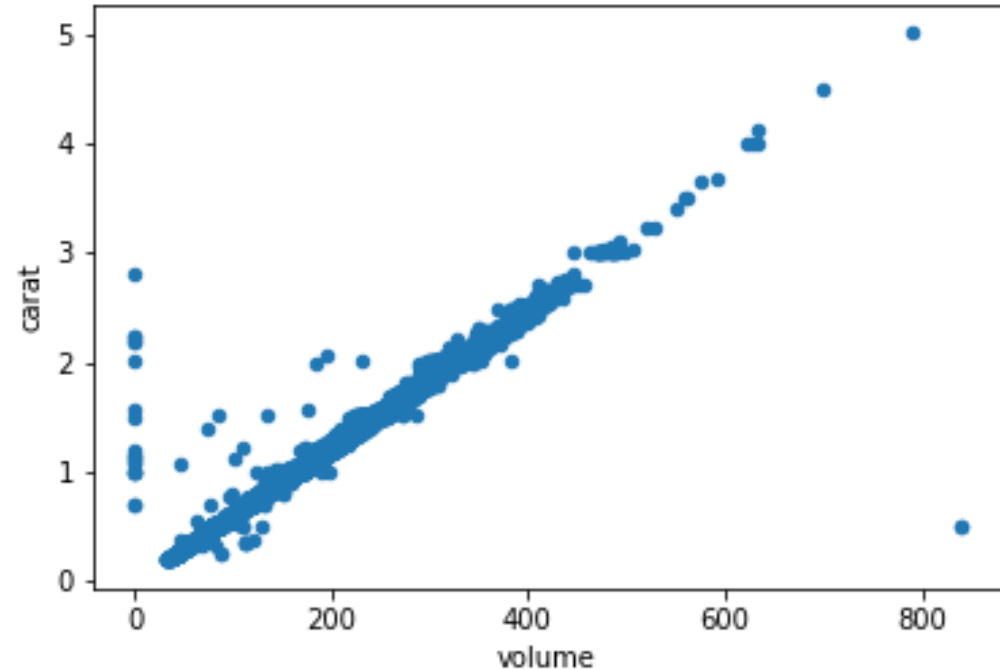
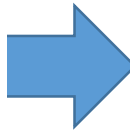
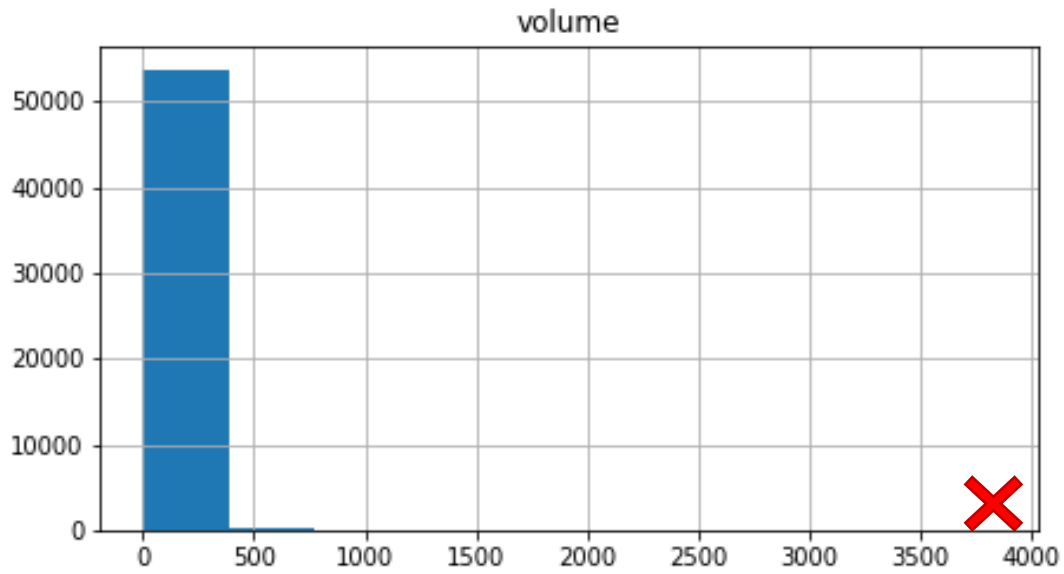
The Dataset

- 1 target
- 10 features
 - Color → D: best, J: Worst
 - Clarity → FL (Flawless), IF, VVS1, VVS2, VS1, VS2, SI1, SI2, I1, I2, I3

carat	cut	color	clarity	depth	table	price	x	y	z
0.50	Premium	E	SI1	62.6	56.0	1314	5.07	5.06	3.17
0.80	Premium	E	VS1	61.7	58.0	3967	5.98	5.95	3.68
0.70	Fair	H	VS1	62.0	73.0	2100	5.65	5.54	3.47
0.32	Premium	G	VVS2	60.8	59.0	730	4.41	4.44	2.69
2.07	Ideal	J	VVS2	62.7	54.0	16617	8.12	8.17	5.11
0.70	Ideal	E	SI1	60.2	57.0	2575	5.78	5.82	3.49
0.38	Ideal	I	SI1	61.6	57.0	626	4.62	4.66	2.86
0.71	Very Good	H	SI1	62.2	56.0	2188	5.72	5.76	3.57
0.59	Very Good	D	SI1	62.8	57.0	1743	5.32	5.38	3.36
0.30	Very Good	E	VS2	62.9	58.0	658	4.22	4.24	2.66

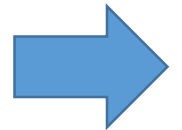
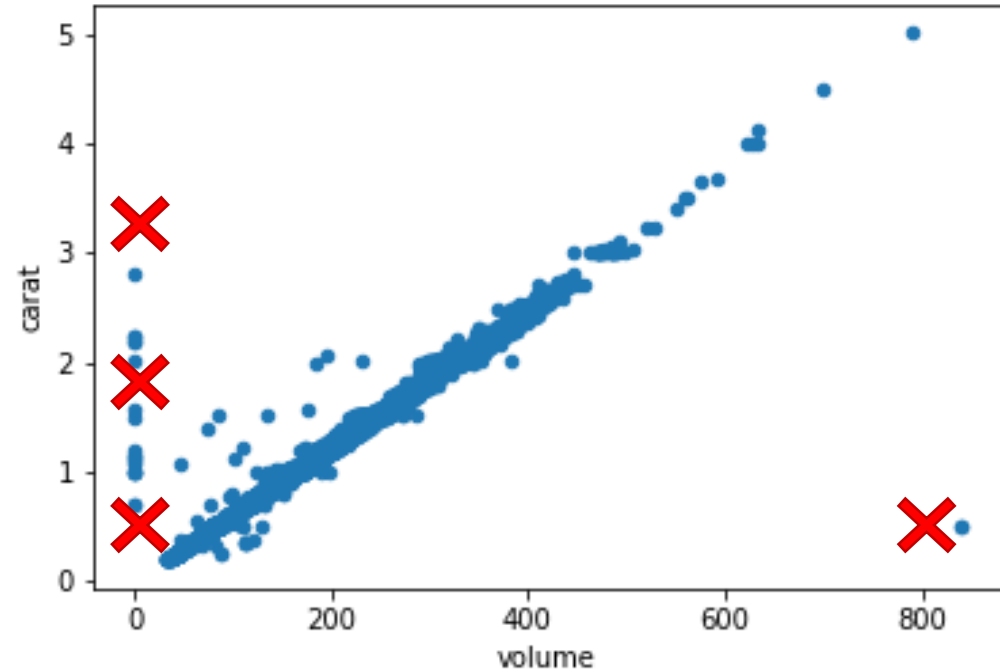
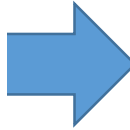
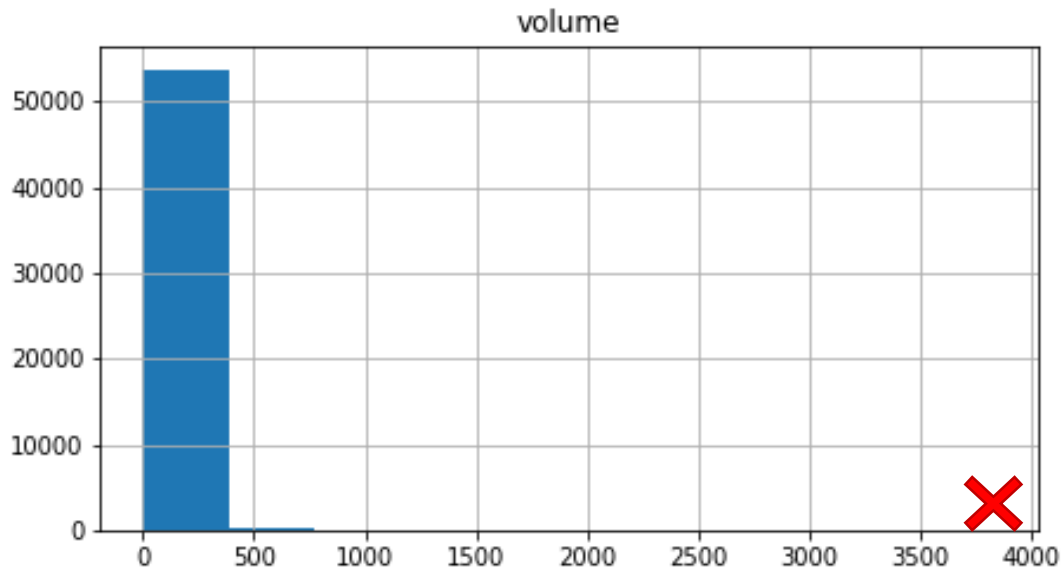
Feature Engineering

- `df['volume'] = df['x'] * df['y'] * df['z']`



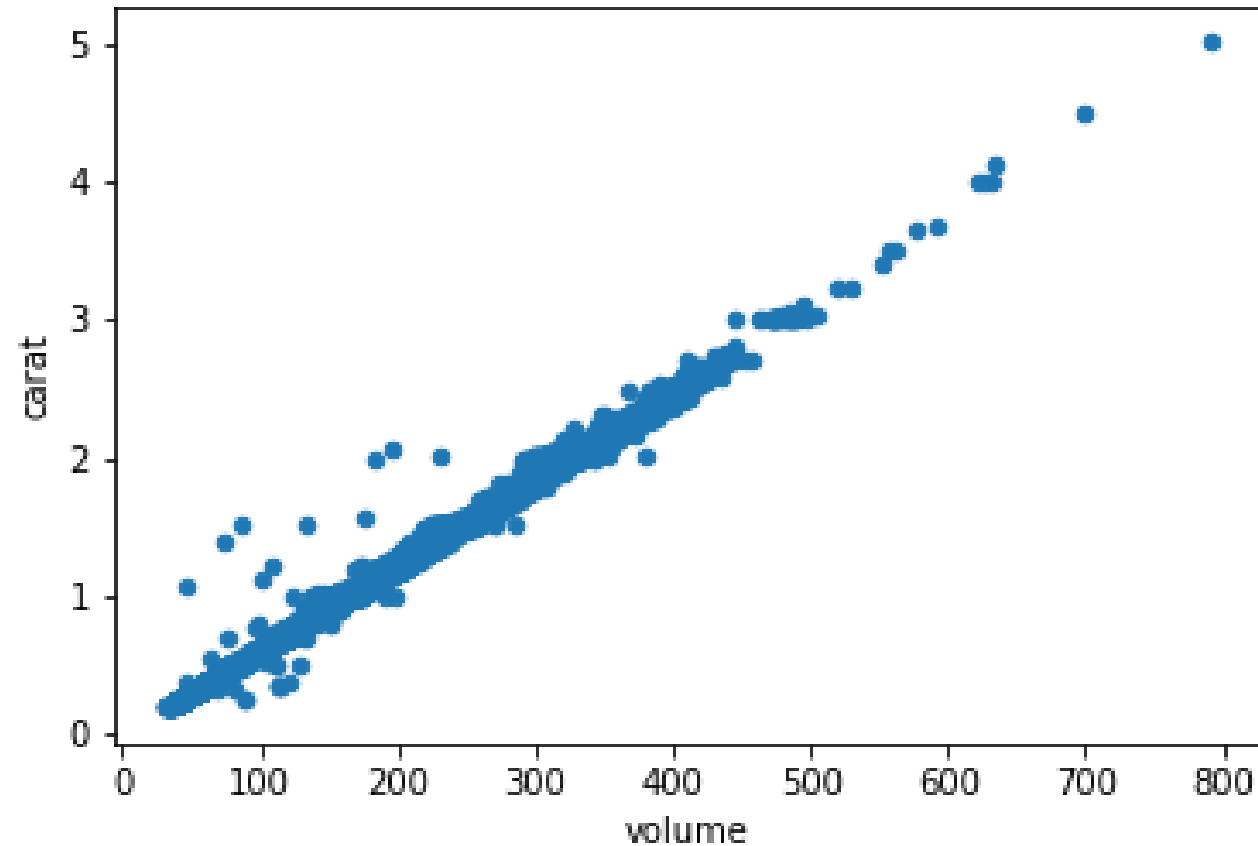
Feature Engineering

- `df['volume'] = df['x'] * df['y'] * df['z']`



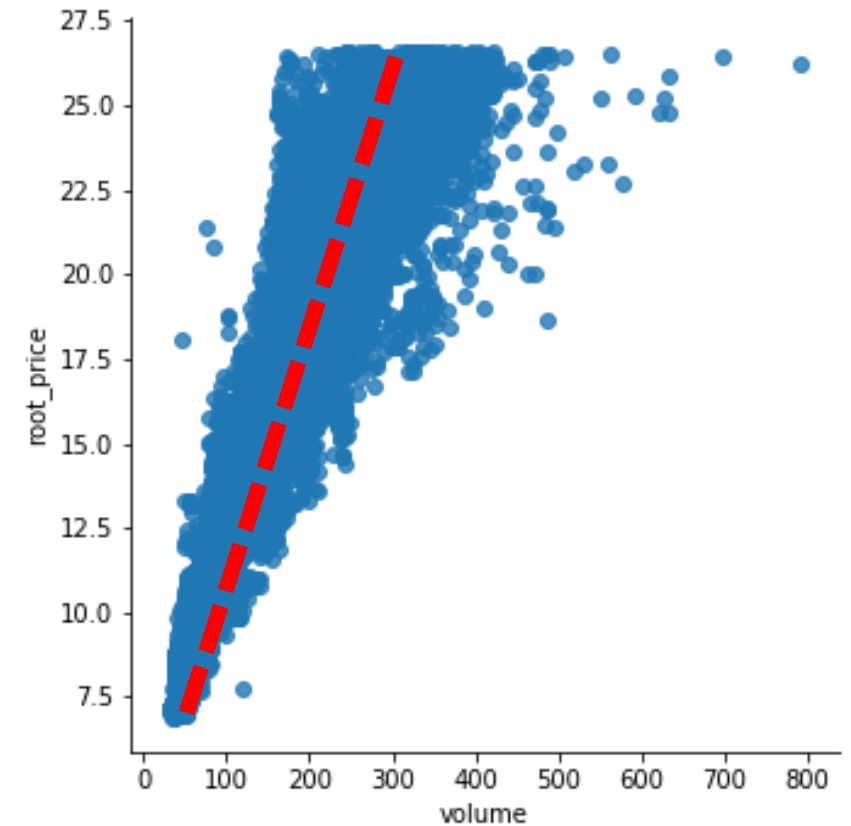
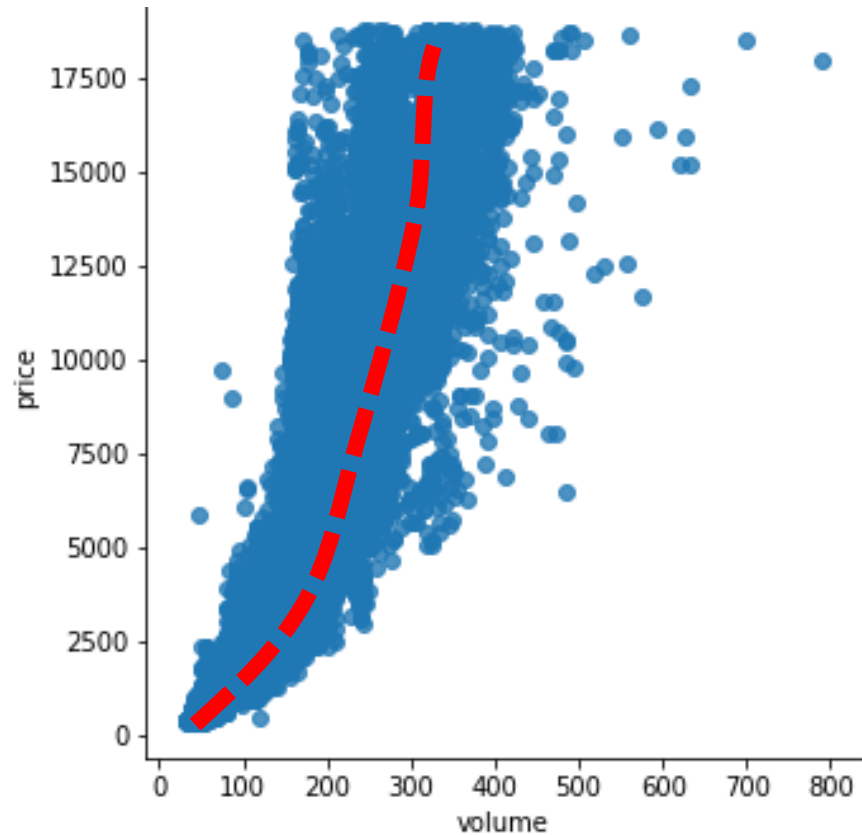
Feature Engineering

- `df['volume'] = df['x'] * df['y'] * df['z']`



Feature Engineering

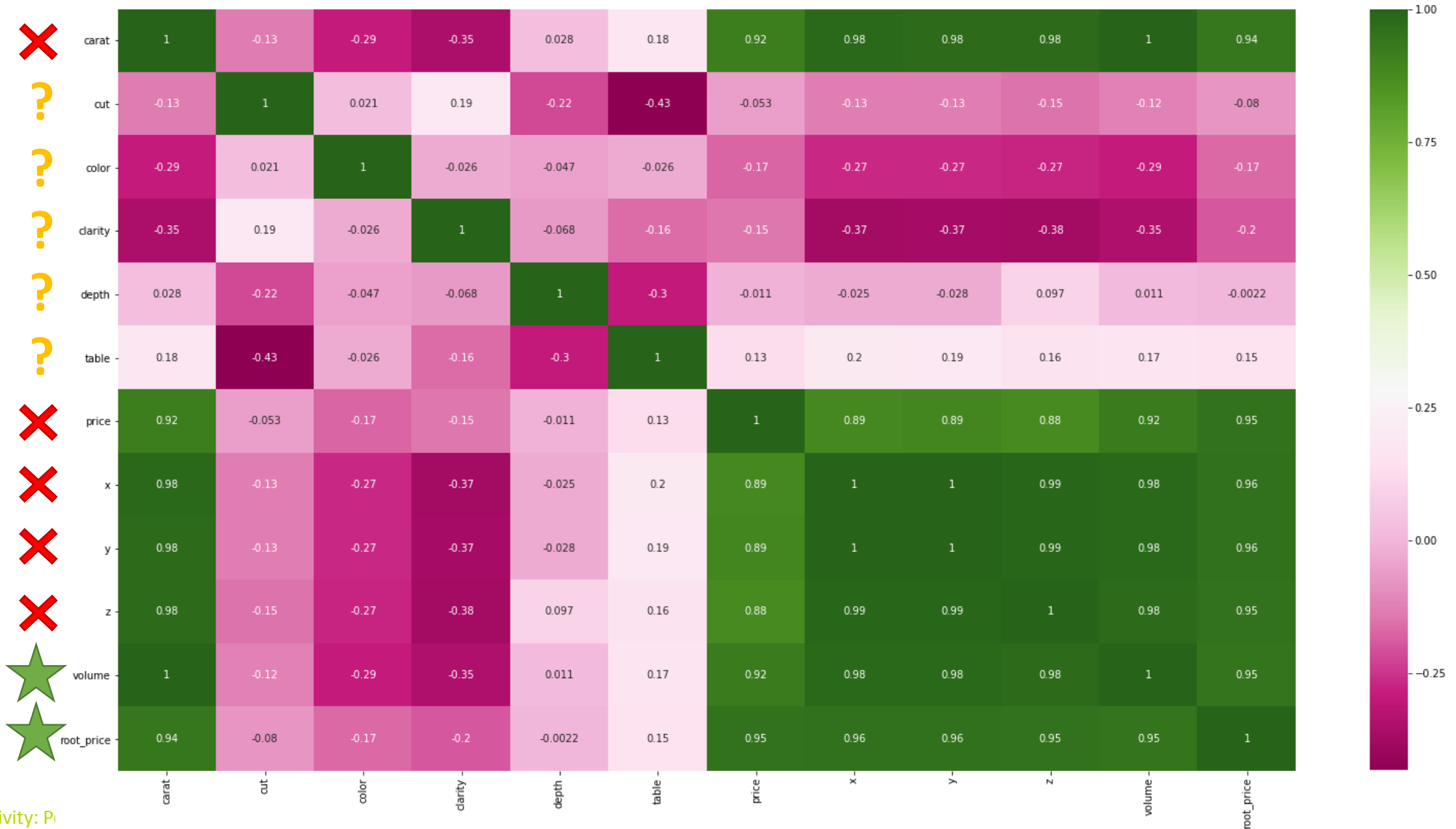
- `df['root_price'] = np.cbrt(df['price'])`



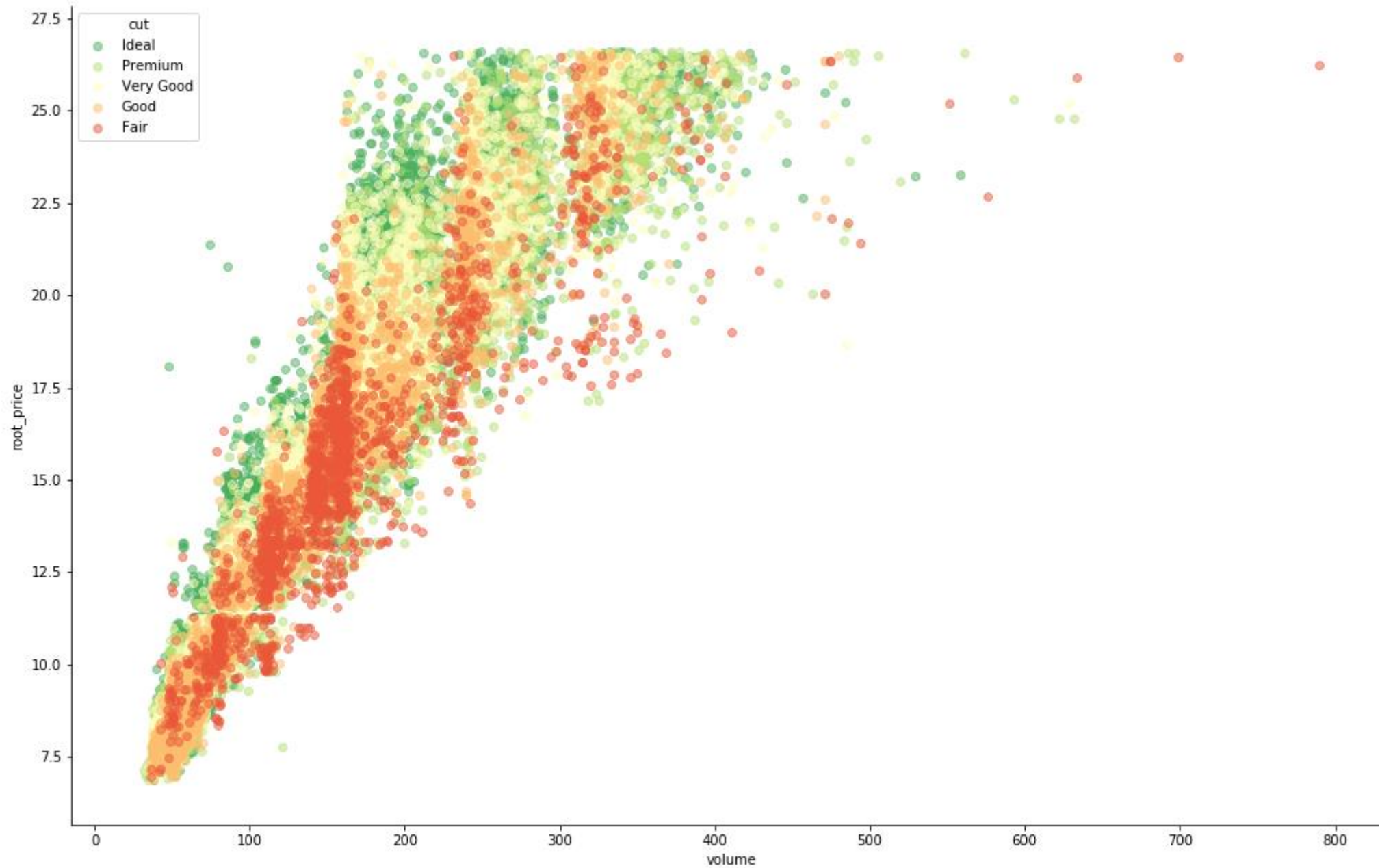
Exploratory Analysis



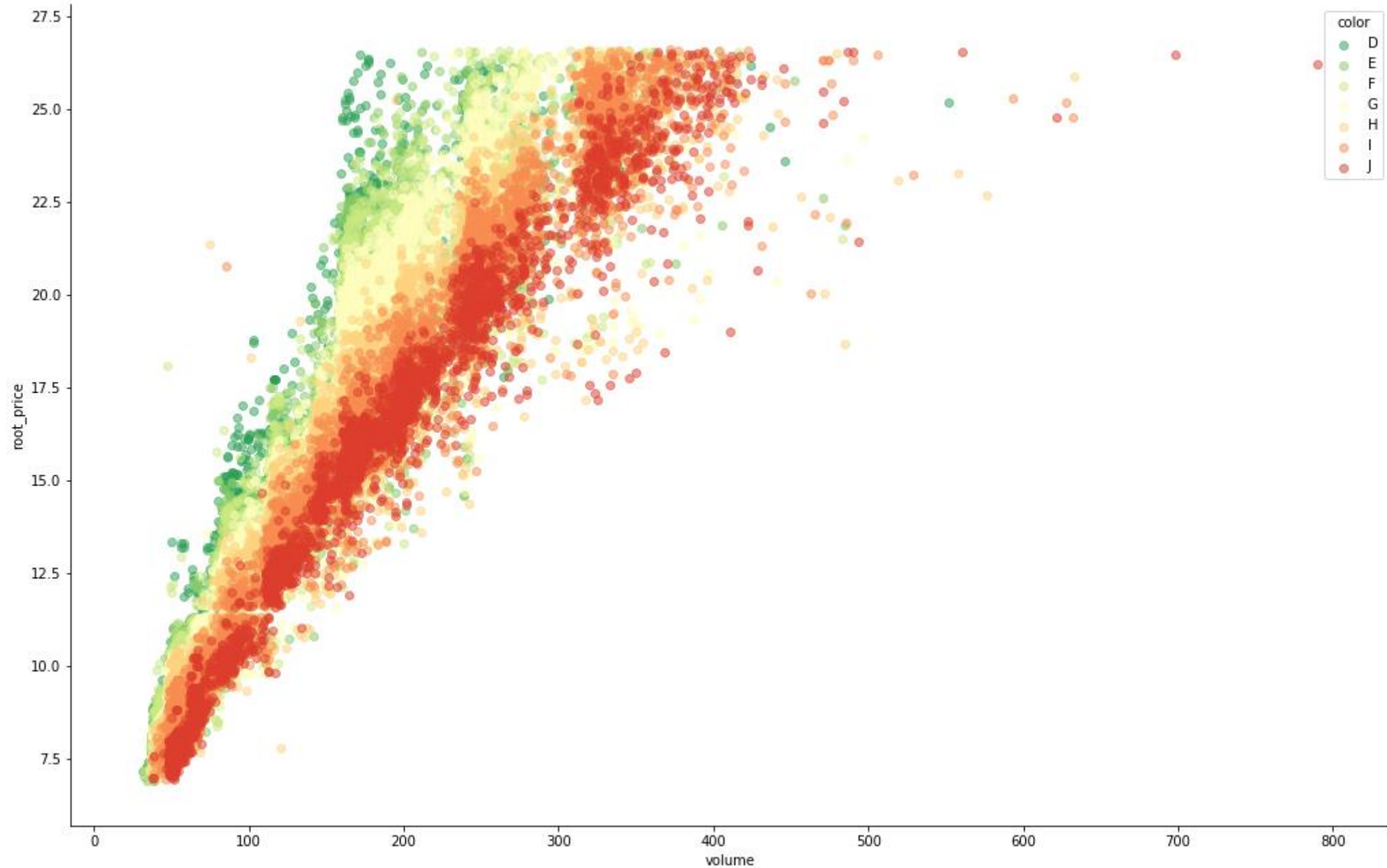
Exploratory Analysis



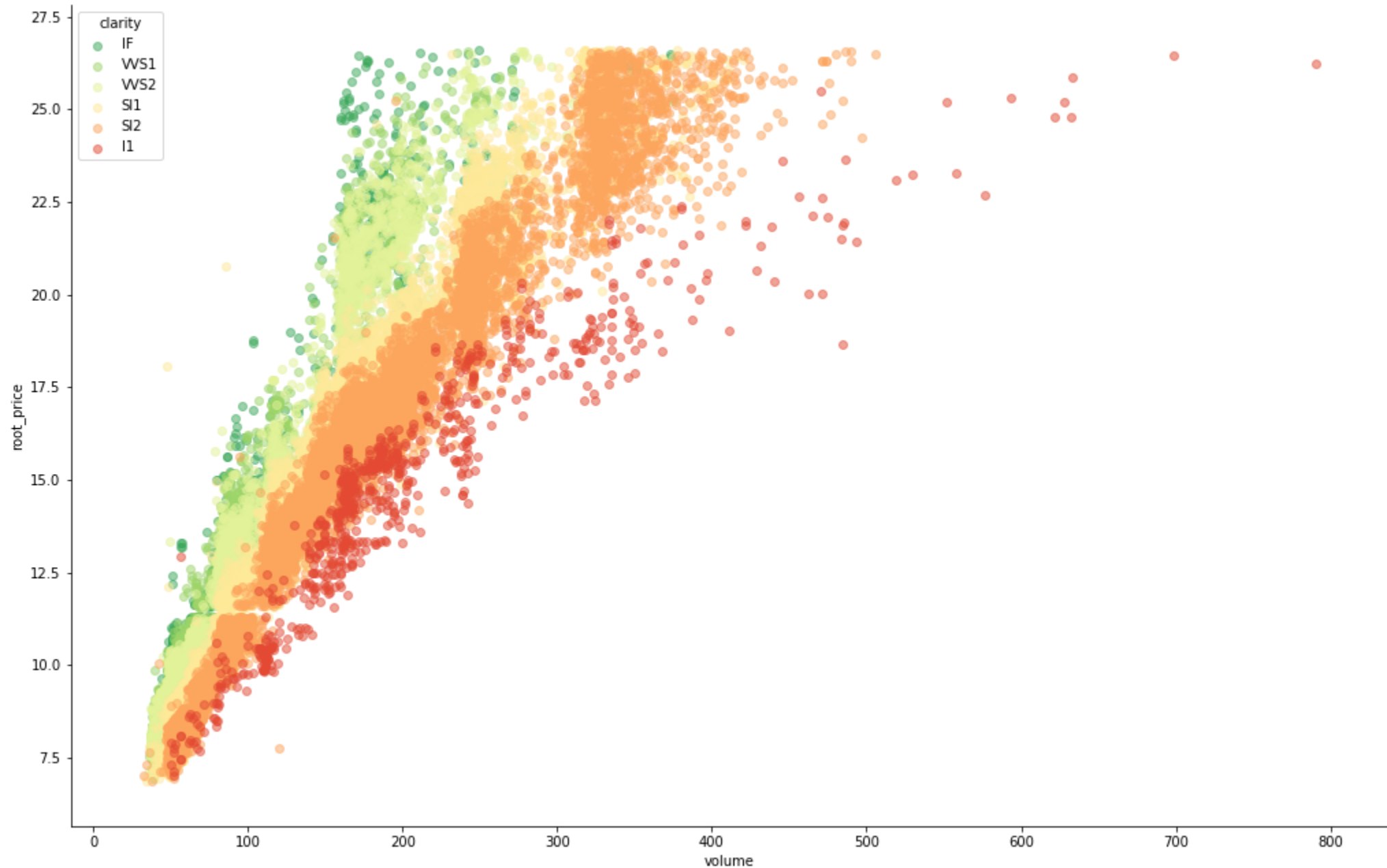
Exploratory Analysis - Cut



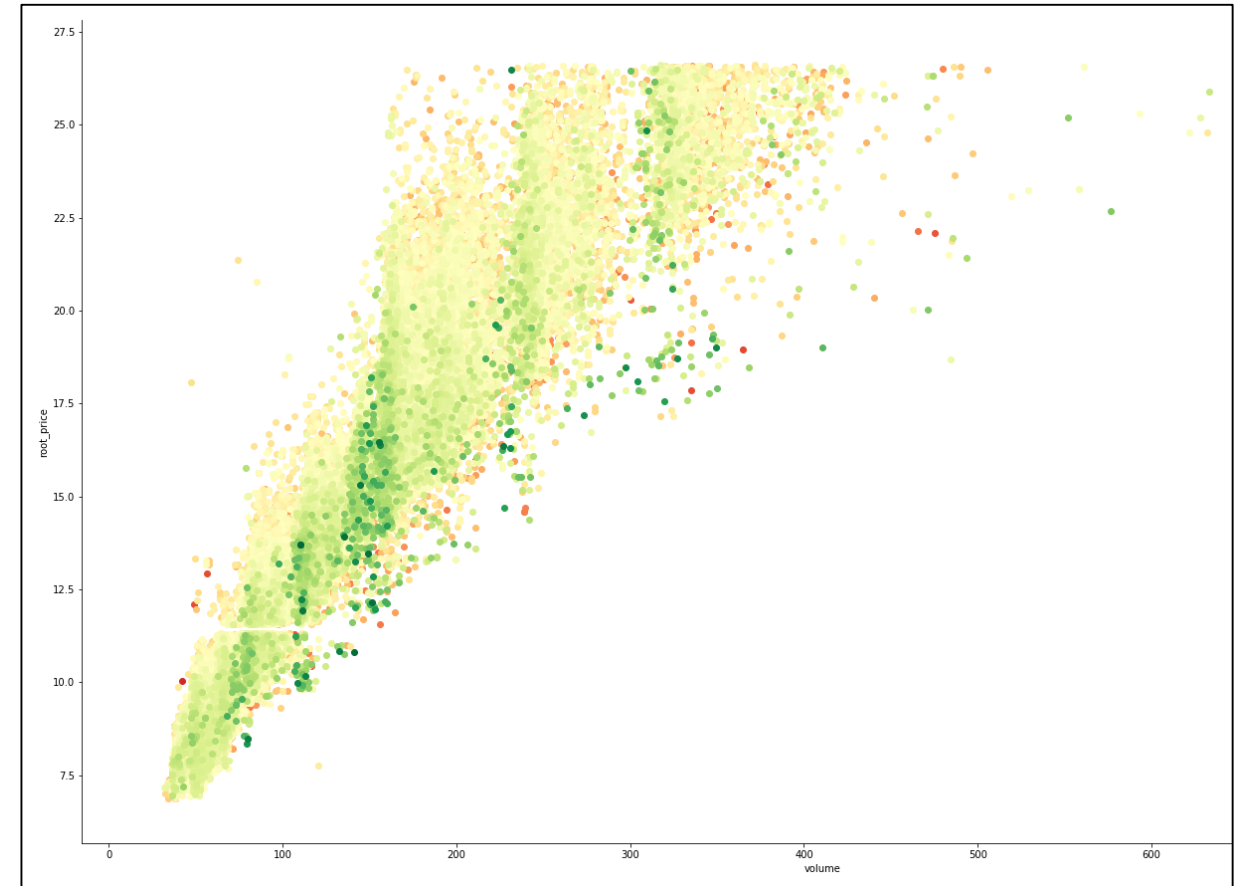
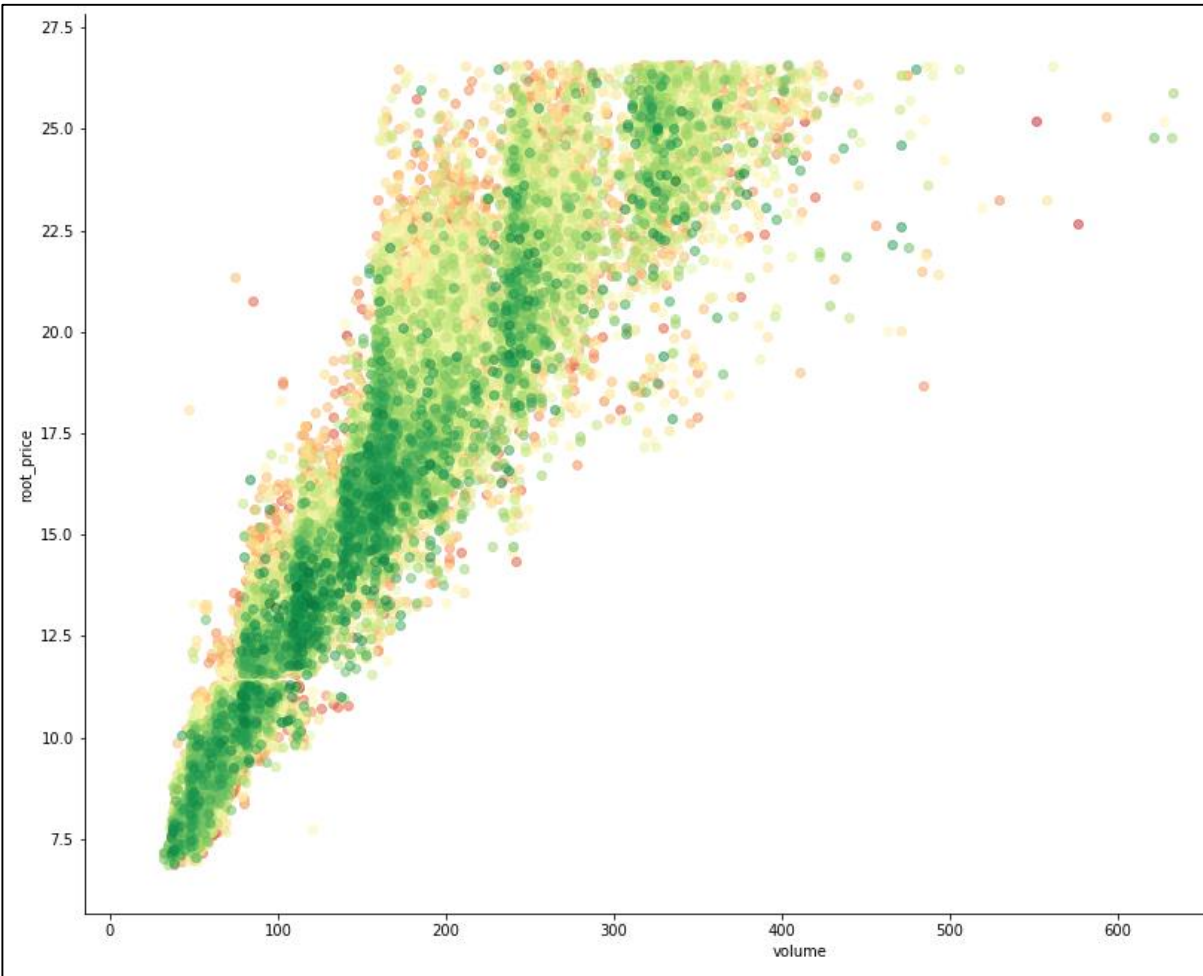
Exploratory Analysis - Color



Exploratory Analysis - Clarity



Exploratory Analysis - Table & Depth



Exploratory Analysis - Table & Depth

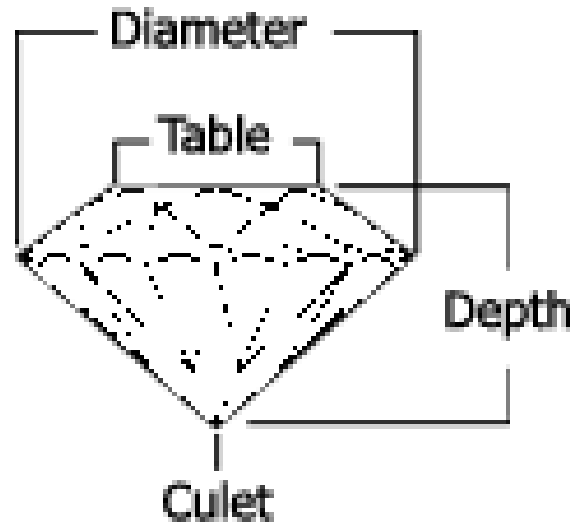


Table % = Table + Diameter
Depth % = Depth + Diameter

	Depth %	Table %
Excellent	59.0% - 61.0%	53% - 60%
Very Good	58.0% - 62.0%	61% - 62%
Good	56% - 64%	62% - 64%
Fair	64% - 70%	64% - 66%
Poor	over 70%	over 66% or under 53%

Source: <https://www.torresjewelco.com.au/diamonds/education/depth-table-percentage.html>

Label Encoding

- `df['cut']`

`{'Fair': 1, 'Good': 2, 'Very Good': 3, 'Premium': 4, 'Ideal':5}`

- `df['color']`

`{'J': 1, 'I': 2, 'H': 3, 'G': 4, 'F':5, 'E':6, 'D':7}`

- `df['clarity']`

`{'I3': 1, 'I2': 2, 'I1': 3, 'SI2': 4, 'SI1':5,
'VS2':6, 'VS1':7, 'VVS2': 8, 'VVS1':9, 'IF':10, 'FL':11}`

Train/Validation/Test Split

- Separation
 - Train: %50
 - Validation: %25
 - Test: %25
- Stratification
 - 100 bins on the target using `np. digitize`, as `root_price` is continuous
- Shuffling
 - Dataset is ordered on price when imported

Model Selection – Decision Tree

- Default Decision Tree regressor on training data

```
DecisionTreeRegressor(criterion='mse', max_depth=None, max_features=None,  
                      max_leaf_nodes=None, min_impurity_decrease=0.0,  
                      min_impurity_split=None, min_samples_leaf=1,  
                      min_samples_split=2, min_weight_fraction_leaf=0.0,  
                      presort=False, random_state=22, splitter='best')
```

Decision Tree Performance on Training Data:

r2: 0.9997

Mean Absolute Error: 0.0251

Mean Squared Log Error: 0.0001

Decision Tree Hyperparameter Tuning

- GridSearchCV is used
- `grid={"max_depth":[1,2,3,4,5,6,7,8,9,10,11,12]`
 `"min_samples_split":[10,20,50,100],`
 `"min_samples_leaf":[1,5,10,20,50,100]}`

Tuned Hyperparameters:

```
{'max_depth': 11, 'min_samples_leaf': 5, 'min_samples_split': 20}
```

Model Selection – Decision Tree

- Decision Tree regressor on validation data, w/ or w/o tuning

Decision Tree Performance on Validation Data, NO Hyperparameter Tuning:

r2: 0.9784

Mean Absolute Error: 0.4814

Mean Squared Log Error: 0.0021

Decision Tree Performance on Validation Data, WITH Hyperparameter Tuning:

r2: 0.9859

Mean Absolute Error: 0.4141

Mean Squared Log Error: 0.0014

Model Selection – Linear Regression

- Linear Regression with the following (rounded) formula:

$$\text{Price} = -4.46 + (\text{Volume} * 0.07) + (\text{Clarity} * 0.5) + (\text{Color} * 0.4) + (\text{Cut} * 0.09) + (\text{Table} * 0.03) + (\text{Depth} * 0.05)$$

Linear Regression Performance on Training Data:

r2: 0.9317

Mean Absolute Error: 0.8894

Mean Squared Log Error: 0.0055

Linear Regression Performance on Validation Data:

r2: 0.9306

Mean Absolute Error: 0.8918

Mean Squared Log Error: 0.0057

Model Selection – Final Performance

Decision Tree Performance on Test Data

r^2 : 0.9860

Mean Absolute Error: 0.4179

Mean Squared Log Error: 0.0014

Linear Regression Performance on Test Data:

r^2 : 0.9316

Mean Absolute Error: 0.8952

Mean Squared Log Error: 0.0056

