

NBA Salary Linear Regression

Kevin Joo - DA Flex Student

Introduction

Objective: Analyze NBA player stat data to predict salary and pinpoint which stats are desirable for a higher salary

Methodology

Data:

- 2 years worth of salary data for every player**

Tools:

- NumPy and Pandas for data manipulation**
- Matplotlib and Seaborn for data visualizations**
- BeautifulSoup for web-scraping**
- Sklearn for linear modeling**

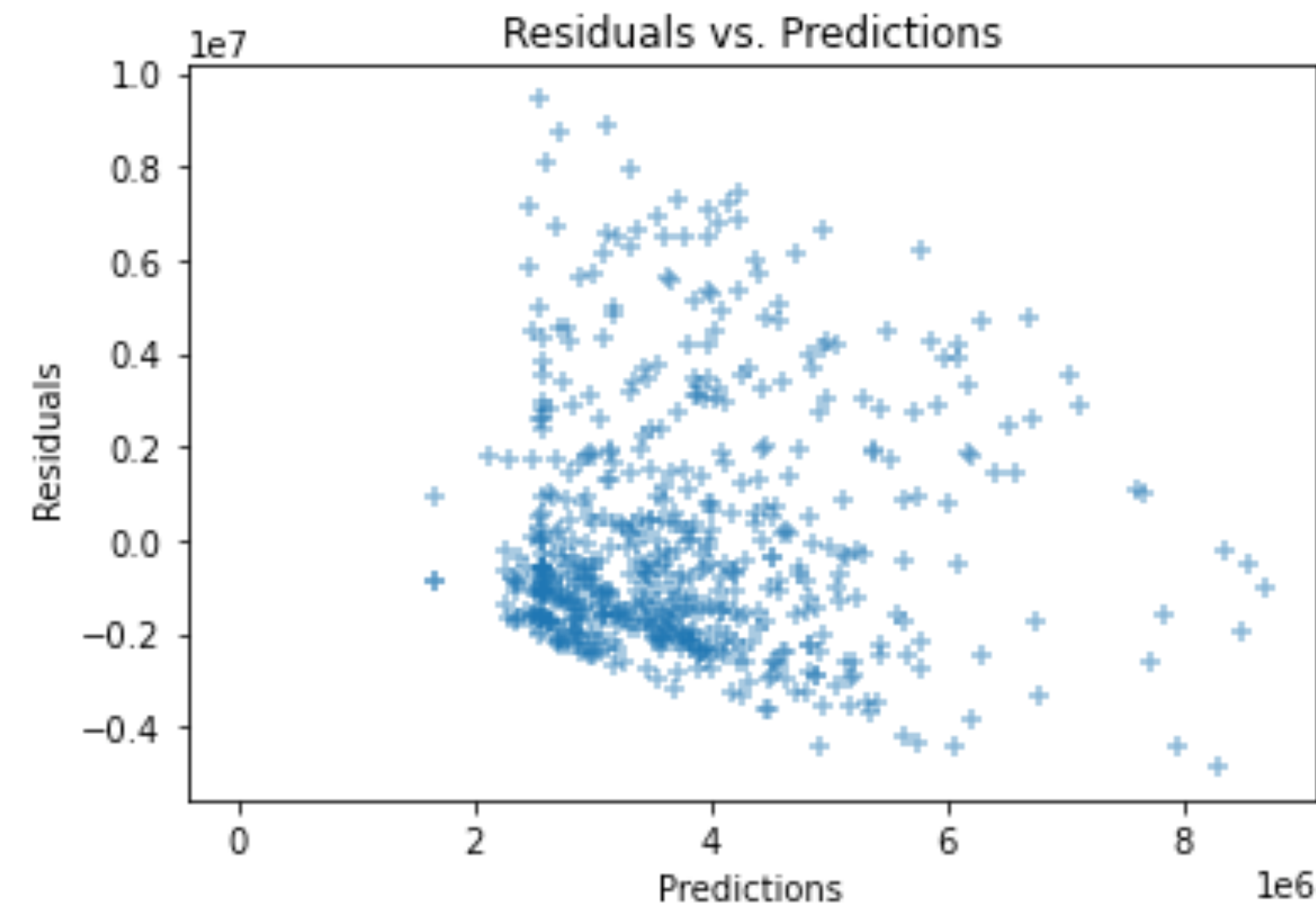
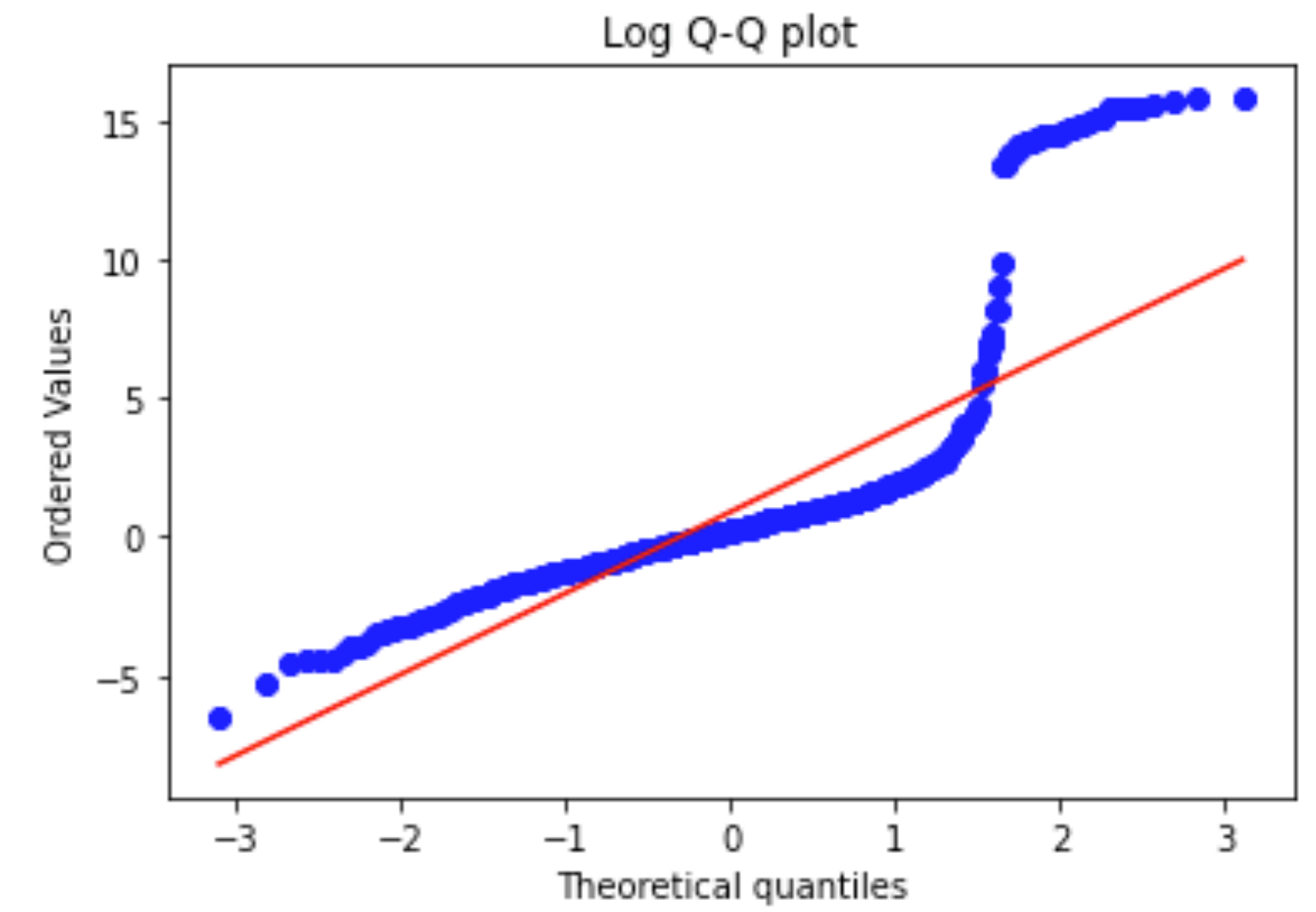
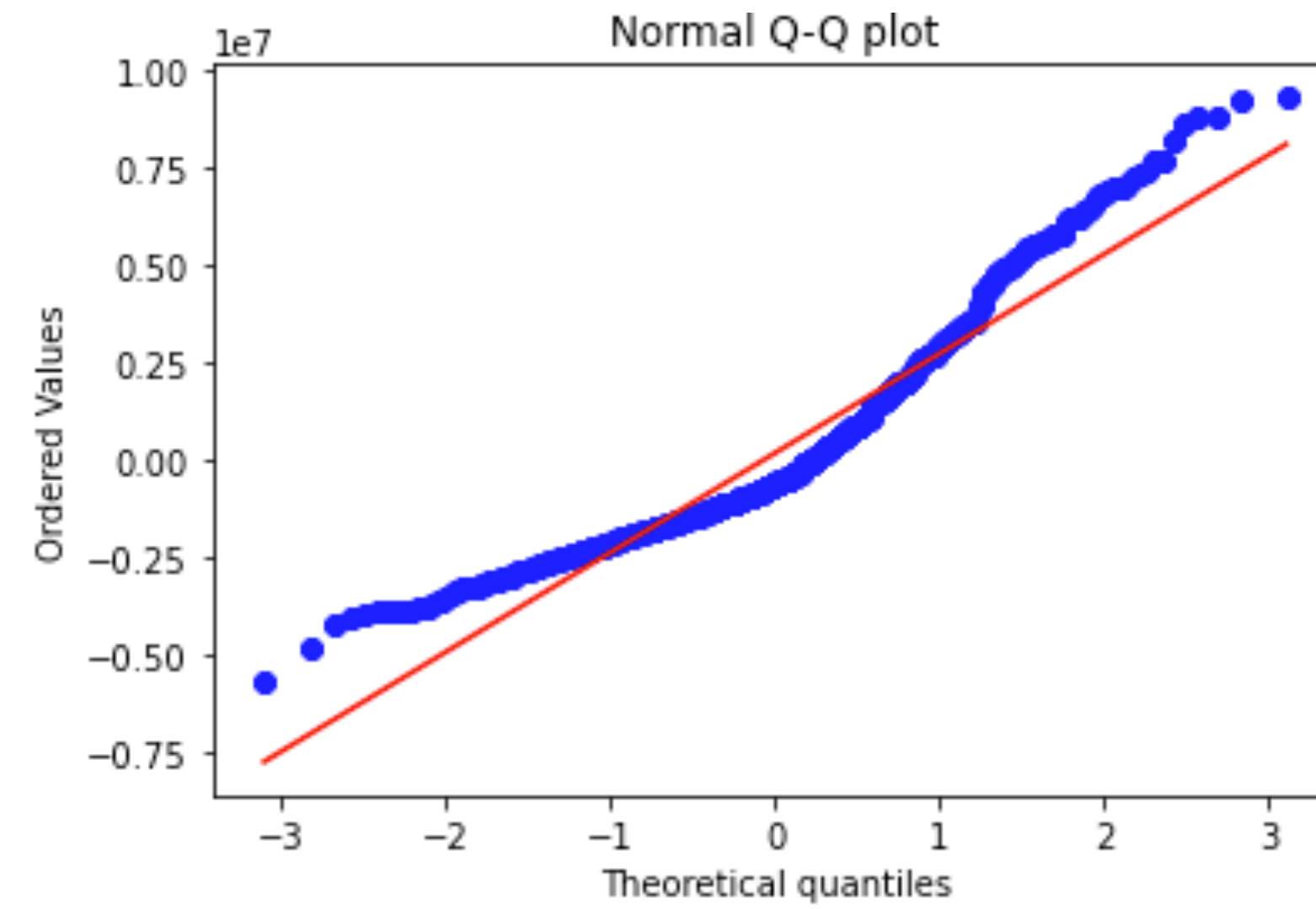
Methodology

Predictor Values: “3P%”, “2P%”, “FT%”, “ORB”, “DRB”, “AST”, “STL”, “BLK”, “TOV”, “PF”

Predicting Value: “Salary”

Results

	variables	vif
0	3P%	8.379234
1	2P%	15.329776
2	FT%	19.967706
3	ORB	8.442780
4	DRB	15.073159
5	AST	12.357243
6	STL	9.077994
7	BLK	5.455680
8	TOV	17.927051
9	PF	15.679048



Results

Simple Linear Regression Model

Split the data: 20/80

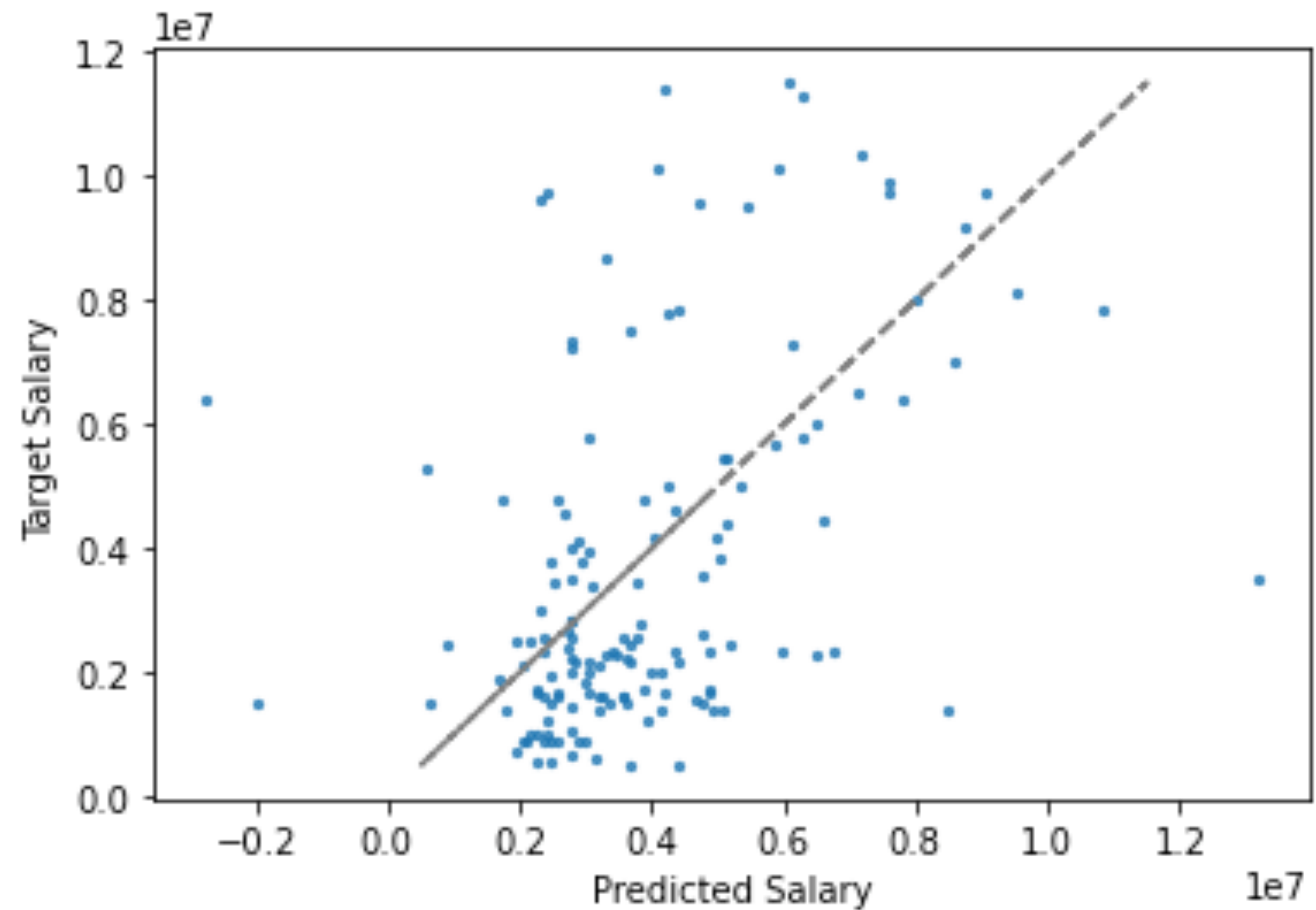
MAE: \$1,996,622

R^2 : 0.211

Added Polynomial Feature:

MAE: \$1,925,832

R^2 : 0.213



Results

Lasso Model

Split the data: 20/80

R^2 Train: 0.160

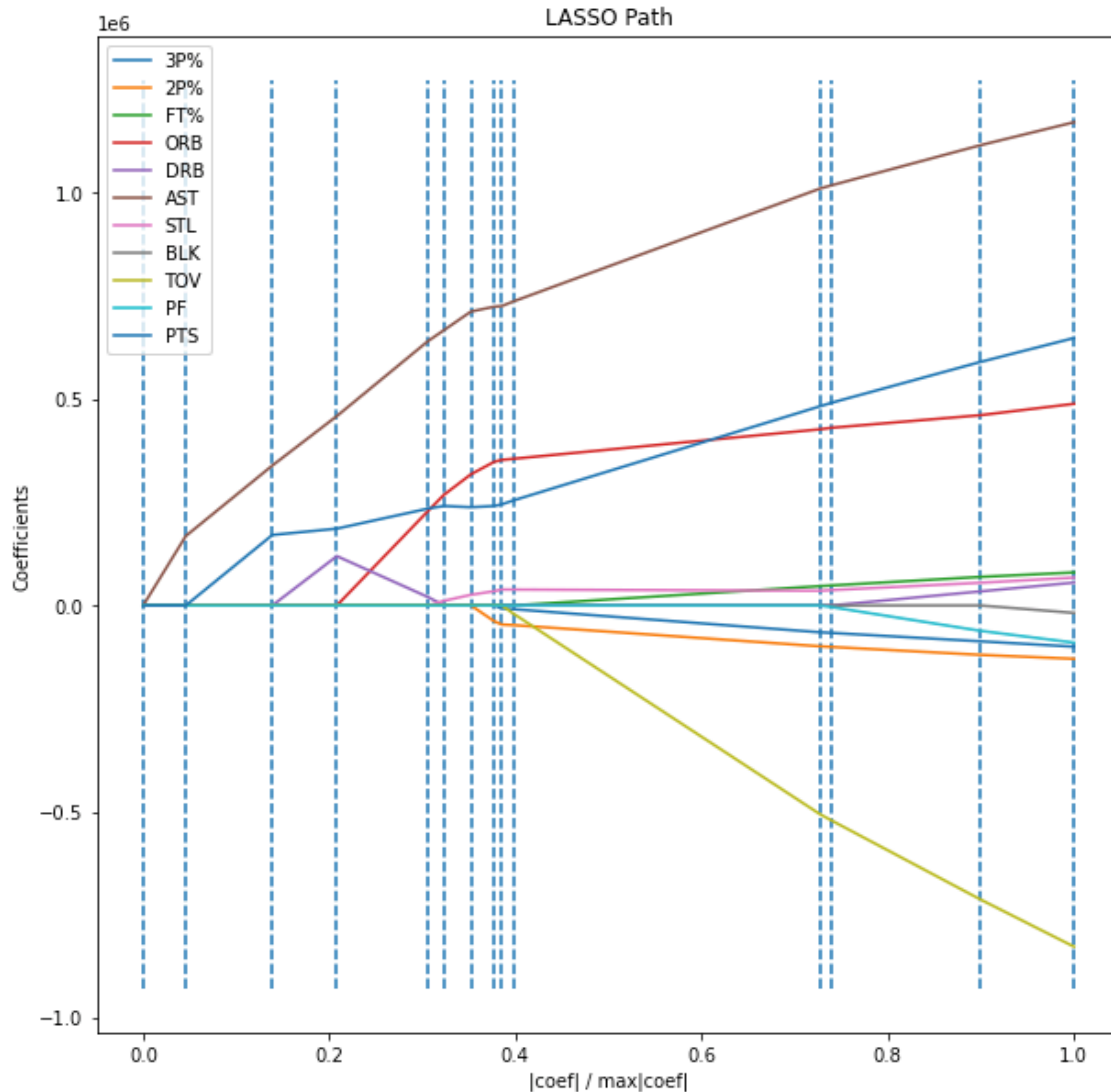
R^2 Test: 0.210

Entry Variable:

AST: 1st

3P%: 2nd

DRB: 3rd



Future Work

If I had more time:

- Investigate further into the data to improve upon the R^2**
- Lowering the heteroskedasticity**
- Incorporating interaction terms to help with the model**

End