

# DSI Project II

**DSIR-1116 - Registration Challenge**

**Test Automation Response**

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# Introduction

**The following study was based on Kagel House Dataset used to predict a specific parameter such as SalePrice estimates the importance of Test Automation in performing structured tasks**

# Methodology

- ✳ Perform a Data Cleaning on all Columns of Data Set relevant to predictions.
- ✳ Create a list with all the Features to be explored.
- ✳ Create a Var to be predicted by our Features.
- ✳ Call Method `project_two_lr(list, var)`
- ✳ Method will create `X_train`, `X_test`, `y_train`, `y_test` parameters.
- ✳ Method will Initiate Linear Regression Model
- ✳ Method will Fit Model using `X,y` data
- ✳ Method will calculate Coeffs for `M` and `B` or `B0` and `B1`.
- ✳ Method will calculate Score for Feature.
- ✳ Method will calculate Predictions on Test Data.
- ✳ From there we can select the Highest Score Features - To be implemented in the future.
- ✳ The idea is to generate Score for all the Features and select the ones with Highest Score.
- ✳ The second step is to find the combinations of Features that Maximizes the  $R^2$  Score.

# Methodology

```
# features = ['Lot Area','Overall Qual','Pool Area','Yr Sold','MS SubClass']
```

```
# y = df['SalePrice']
```

```
def project_two_lr(feature_list, predict_var):
```

```
    # Input Parameters
```

```
    X = df_train[feature_list]
```

```
    y = df_train[predict_var]
```

```
    # Split for Train and Test Model
```

```
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 2020)
```

```
    # Liner Model Instantiation
```

```
    lr = LinearRegression()
```

```
    # Model Fit
```

```
    lr.fit(X_train,y_train)
```

# Methodology

```
# Coeffs for Slope of M
```

```
lr.coef_
```

```
print(f'### COEFFS M #### {lr.coef_}')
```

```
# Coeffs for Slope of B
```

```
lr.intercept_
```

```
print(f'### COEFFS B #### {lr.intercept_}')
```

```
# Score Calculation
```

```
score = lr.score(X_test,y_test)
```

```
print(f'### SCORE #### {lr.score(X_test,y_test)}')
```

```
return score
```

```
print(project_two_lr(['Lot Area','Overall Qual','Pool Area','Yr Sold','MS SubClass'], 'SalePrice'))
```

# Findings

```
### COEFFS M #### [ 2.08336256e+00  4.34217305e+04 -5.96901459e+00  1.60450202e+02  
-1.31650182e+02]  
### COEFFS B #### -419709.43981279276  
### SCORE #### 0.7109823041075626  
0.7109823041075626
```

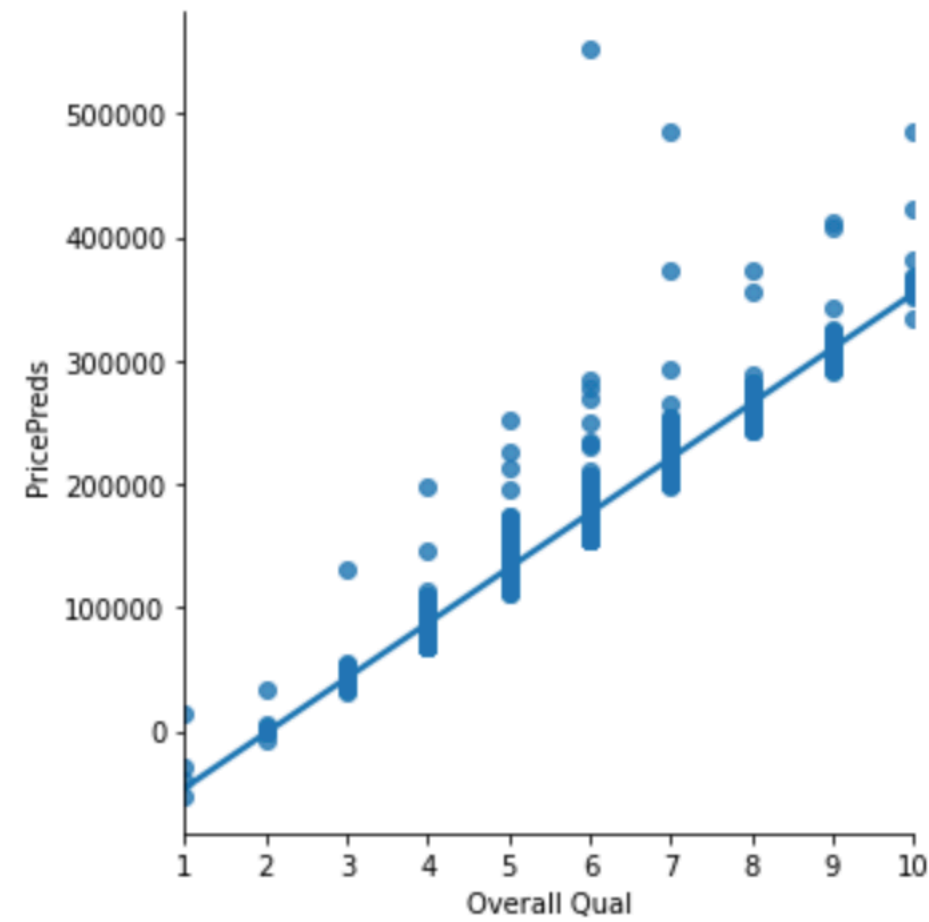
```
# Generating Predictions  
df_test['SalePrice'] = lr.predict(X_test)  
predictions = lr.predict(X_test)
```

```
In [117]: 1 for i in range(len(predictions)):
          2     print(f'### Index {X_test.index[i]} - Prediction {predictions[i]}')
          3     print("-----")

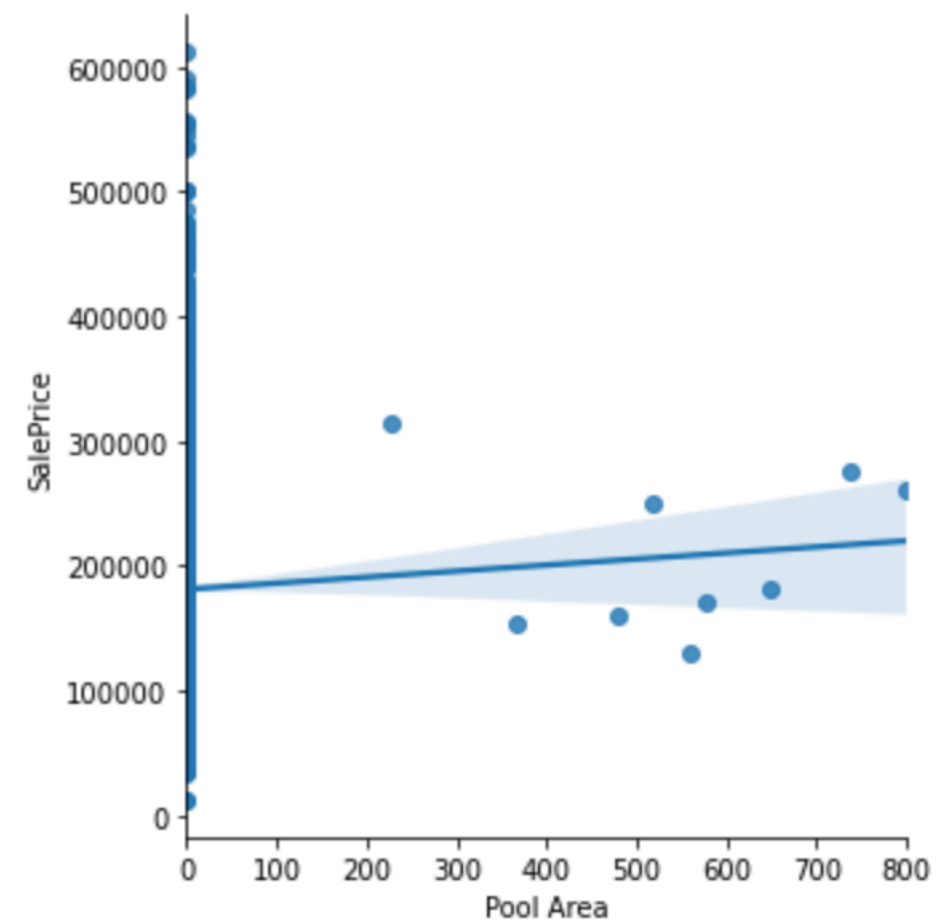
### Index 1970 - Prediction 261598.42585134646
-----
### Index 1739 - Prediction 238850.25571054418
-----
### Index 1263 - Prediction 273440.65414306347
-----
### Index 1942 - Prediction 127700.62415766803
-----
### Index 1258 - Prediction 134417.3294409581
-----
### Index 1916 - Prediction 131033.21368690455
-----
### Index 772 - Prediction 79269.23756095534
-----
### Index 963 - Prediction 148136.17829744978
-----
### Index 196 - Prediction 219026.7628140936
-----
### Index 874 - Prediction 168483.42890468927
```

# Findings

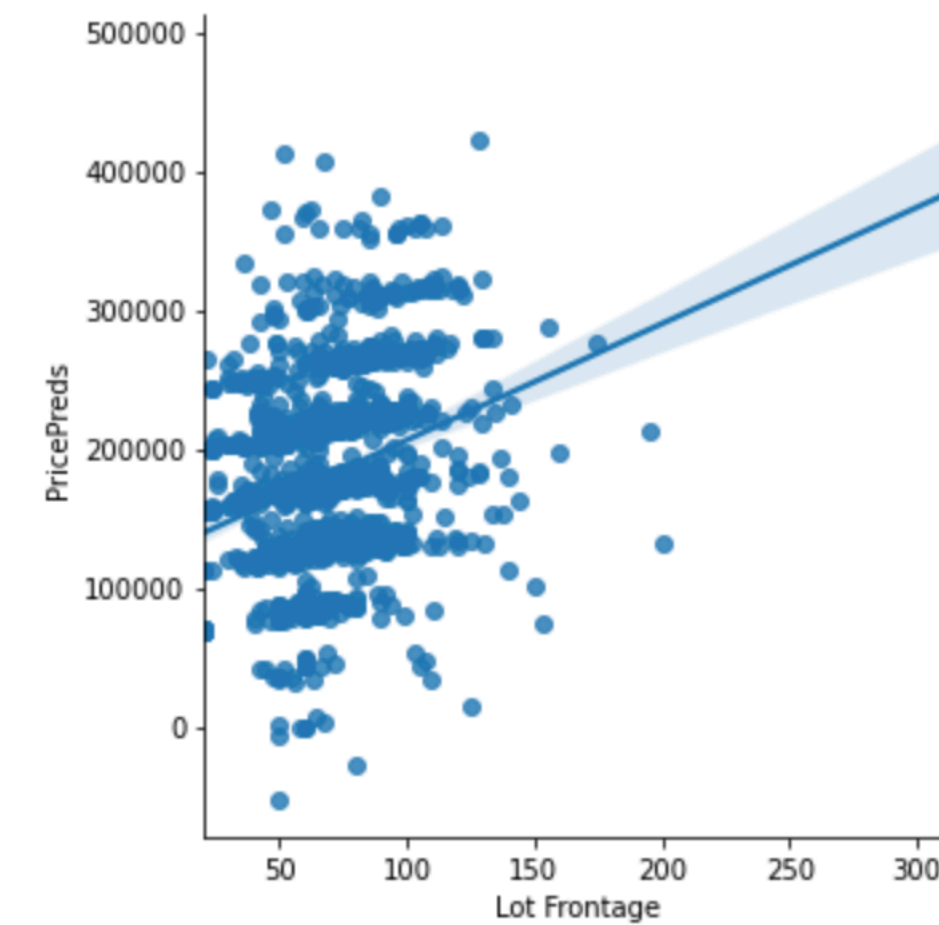
```
1 sns.lmplot(x='Overall Qual', y='PricePreds', data=df_train)
2 plt.show()
```



```
1 sns.lmplot(x='Pool Area', y='SalePrice', data=df_train)
2 plt.show()
```



```
1 sns.lmplot(x='Lot Frontage', y='PricePreds', data=df_train)
2 plt.show()
```



# Conclusion

*A project like this will enormously benefit from Test Automation. Stages such as Cleaning the Data. Estimating the best set of Features - Evaluating Individuals plus Combinations of Parameters to Find the Highest Score for specific Models. Calculate Predictions and Populate resulting Data Frame.*

```
print(project_two_lr(['Lot Area','Overall Qual','Pool Area','Yr Sold','MS SubClass'], 'SalePrice'))
```



QA

***Thank You!***