



#### **Problem Statement - Part II**

**Assignment Part-II** 

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1. What is the optimal value of alpha for ridge and lasso regression? What will be the changes in the model if you choose double the value of alpha for both ridge and lasso? What will be the most important predictor variables after the change is implemented?

Optimal Alpha value for Ridge: {'alpha': 1.6}

Optimal Alpha value for Lasso: {'alpha': 0.0001}

Double the Optimal Alpha value for Ridge: {'alpha': 3.2}

Double the Optimal Alpha value for Lasso: {'alpha': 0.0002}

	Metric	Linear Regression	Ridge Regression	Lasso Regression		Metric	Linear Regression	Ridge Regression	Lasso Regression
0	R2 Score (Train)	9.536295e-01	0.947998	0.943149	0	R2 Score (Train)	9.536295e-01	0.943438	0.934957
1	R2 Score (Test)	8.834611e-01	0.898155	0.905573	1	R2 Score (Test)	8.834611e-01	0.898562	0.904065
2	RSS (Train)	2.709449e+11	0.585965	0.640602	2	RSS (Train)	2.709449e+11	0.637352	0.732914
3	RSS (Test)	2.696234e+11	0.454402	0.421307	3	RSS (Test)	2.696234e+11	0.452586	0.428036
4	RMSE (Train)	1.661058e+04	0.024428	0.025541	4	RMSE (Train)	1.661058e+04	0.025476	0.027319
5	RMSE (Test)	2.527683e+04	0.032814	0.031597	5	RMSE (Test)	2.527683e+04	0.032749	0.031848

By theoretically, After doubling the optimum alpha, model bias will increase. Here In out, R2\_train score has decreased





1. What will be the most important predictor variables after the change is implemented?

#### **Optimal Alpha value**



#### **Double Optimal Alpha value**

1stFlrSF	0.110734		
BsmtFinSF1	0.081349		
OverallQual V Excel	0.074142		
2ndFlrSF	0.068898		
RoofMatl WdShngl	0.056522		
OverallQual Excellent	0.052298		
TotRmsAbvGrd	0.049884 0.043679 0.039525 0.038788 0.037957		
MasVnrArea			
BsmtUnfSF			
LotArea			
Neighborhood StoneBr			
GarageArea	0.034623		
Neighborhood NoRidge	0.034594		
FullBath	0.031834		
OverallQual V good	0.028395		
SaleType New	0.028138		
Exterior1st_BrkFace	0.027078		
BsmtExposure Gd	0.026258		
OverallCond_Excellent beta_coef2['Lasso'].sor	0.026179 		
beta_coef2['Lasso'].sor	t_values(ascending=False)		
beta_coef2['Lasso'].sor	t_values(ascending=False)		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel	t_values(ascending=False) 0.234709 0.132964		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel 2ndFlrSF	t_values(ascending=False) 0.234709 0.132964 0.114980		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent	t_values(ascending=False) 0.234709 0.132964 0.114980 0.087676		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1	t_values(ascending=False) 0.234709 0.132964 0.114980 0.087676 0.072872		
beta_coef2['Lasso'].sor lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_wdShngl	t_values(ascending=False) 0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820		
beta_coef2['Lasso'].sor 1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328		
beta_coef2['Lasso'].sor  1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good	t_values(ascending=False) 0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150		
beta_coef2['Lasso'].sor  1stF1rSF OverallQual_V_Excel 2ndF1rSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854		
beta_coef2['Lasso'].sor  1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_wdShngl MasVnrArea OverallQual_V_good GarageArea LotArea	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803		
beta_coef2['Lasso'].sor  1stFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977		
beta_coef2['Lasso'].sor  lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr Neighborhood_NridgHt	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543		
beta_coef2['Lasso'].sor  lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_NridgHt Neighborhood_NoRidge	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543 0.023295		
beta_coef2['Lasso'].sor  lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr Neighborhood_NridgHt Neighborhood_NoRidge BsmtExposure_Gd	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543 0.023295 0.023043		
beta_coef2['Lasso'].sor  lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSFl SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr Neighborhood_NridgHt Neighborhood_NoRidge BsmtExposure_Gd ExteriorIst_BrkFace	nt_values(ascending=False) 0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543 0.023295 0.023043 0.022476		
beta_coef2['Lasso'].sor  lstFlrSF OverallQual_V_Excel 2ndFlrSF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_WdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr Neighborhood_NridgHt Neighborhood_NoRidge BsmtExposure_Gd Exteriorlst_BrkFace Functional_Typ	0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543 0.023295 0.023043 0.022476 0.022476		
beta_coef2['Lasso'].sor  lstFlrsF OverallQual_V_Excel 2ndFlrsF OverallQual_Excellent BsmtFinSF1 SaleType_New RoofMatl_wdShngl MasVnrArea OverallQual_V_good GarageArea LotArea Neighborhood_StoneBr Neighborhood_NridgHt Neighborhood_NRidge BsmtExposure_Gd Exteriorlst_BrkFace	nt_values(ascending=False) 0.234709 0.132964 0.114980 0.087676 0.072872 0.044355 0.041820 0.038328 0.037150 0.034854 0.029803 0.026977 0.025543 0.023295 0.023043 0.022476		

## The Important predictor variables after change is implemented are

```
'1stFlrSF',
'OverallQual V Excel',
'2ndFlrSF',
'OverallQual Excellent',
'BsmtFinSF1',
'SaleType New',
'RoofMatl WdShnql',
'MasVnrArea',
'OverallOual V good',
'GarageArea',
'LotArea',
'Neighborhood StoneBr',
'Neighborhood NridgHt',
'Neighborhood NoRidge',
'BsmtExposure Gd',
'Exterior1st BrkFace',
'Functional Typ',
'Neighborhood Crawfor',
'GarageCars'
```





2. You have determined the optimal value of lambda for ridge and lasso regression during the assignment. Now, which one will you choose to apply and why?

From ridge & lasso regression, I'll choose Lasso Regression model because of the high R2\_Score(Test) - 0.90 and it is close to its R2\_Score(Train) - 0.94

Lasso Regression model uses very less number of predictor variables when compared to the Ridge regression model

RMSE value is also very less in Lasso

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3. After building the model, you realised that the five most important predictor variables in the lasso model are not available in the incoming data. You will now have to create another model excluding the five most important predictor variables. Which are the five most important predictor variables now?

After removing the top 5 significant variable, the next 5 significant variable will be

```
'SaleType_New',
'RoofMatl_WdShngl',
'MasVnrArea',
'OverallQual_V_good',
'GarageArea',
'LotArea',
```





4. How can you make sure that a model is robust and generalisable? What are the implications of the same for the accuracy of the model and why?

Model which learns the generic pattern of the data from seen data & able to predict the output of target variable in unseen data is robust & generalisable

r2\_score is the one of measure of linear regression model along with RMSE & MSE.

r2\_score represents the amount of variability explained by the model

#### For ex:

- a. If the r2\_score of test data is too low when compare to it's train data, it means the model has memorised the noise along with pattern and it is a clear sign of overfitting
- b. If the r2\_score of train data itself is too low, then the model was not able to find any pattern and it is underfitting





# Thank you

Arunkumar