Question 1

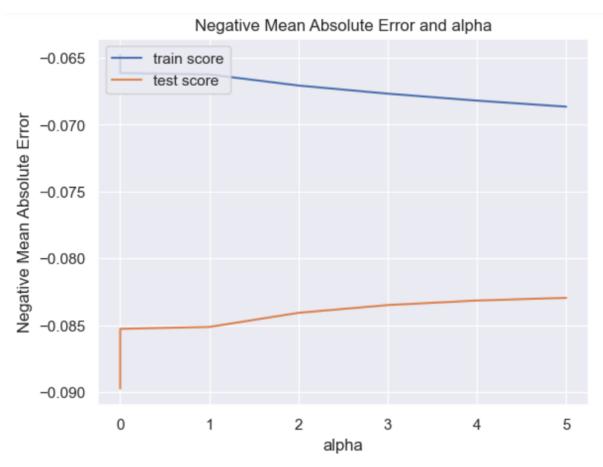
What is the optimal value of alpha for ridge and lasso regression? What will be the changes in the model if you choose double the value of alpha for both ridge and lasso? What will be the most important predictor variables after the change is implemented?

Answer:

For ridge regression: When I plot the curve between negative mean absolute error and alpha I can see that as the value of alpha increase from 0 the error term decrease and the train error is showing increasing trend when value of alpha increases .when the value of alpha is 2 the test error is minimum so I decided to go with value of alpha equal to 2 for our ridge regression.

For lasso regression: I have decided to keep very small value that is 0.01, when I increase the value of alpha the model try to penalize more and try to make most of the coefficient value zero. Initially it came as 0.4 in negative mean absolute error and alpha.

When I double the value of alpha for ridge regression now will take the value of alpha equal to 4 the model will apply more penalty on the curve and try to make the model more generalized that is making model more simpler and no thinking to fit every data of the data set .from the graph I can see that when alpha is 5 I get more error for both test and train. Similarly when I increase the value of alpha for lasso I try to penalize more our model and more coefficient of the variable will reduced to zero, when I increase the value of our r2 square also decreases.



The most important variable after the changes has been implemented for ridge regression are as follows:-

- 1. MSZoning_FV
- 2. MSZoning_RL
- 3. Neighborhood_Crawfor
- 4. MSZoning_RH
- 5. MSZoning RM

The most important variable after the changes has been implemented for lasso regression are as follows:

- 1. GrLivArea
- 2. OverallQual
- 3. OverallCond
- 4. TotalBsmtSF
- 5. BsmtFinSF1

Question 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?

Answer:

The optimal value of lasso is .01. R2 score is better for ridge regression but, it is better to use Lasso, since it brings and assigns a zero value to insignificant features, enabling us to choose the predictive variables. It is always advisable to use simple yet robust model.

Question 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?

Answer:

Those 5 most important predictor variables that will be excluded are :-

- 1. GrLivArea
- 2. OverallQual
- 3. OverallCond
- 4. TotalBsmtSF
- 5. GarageArea

Question 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?

Answer:

The model should be as simple as possible, though its accuracy will decrease but it will be more robust and generalisable. It can be also understood using the Bias-Variance trade-off. The simpler the model the more the bias but less variance and more generalizable. Its implication in terms of accuracy is that a robust and generalisable model will perform equally well on both training and test data

i.e. the accuracy does not change much for training and test data.