

136_Liz_Project_Step5_Log Transformation - 0.8

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```
#install.packages("glmnet")
#install.packages("mlbench")
#install.packages("Boruta")
library(caret)

## Loading required package: lattice

## Loading required package: ggplot2

library(tidyverse)

## -- Attaching packages -----
----- tidyverse 1.2.1 --

## v tibble  1.4.2      v purrr   0.2.5
## v tidyr   0.8.1      v dplyr   0.7.7
## v readr   1.1.1      v stringr 1.3.1
## v tibble  1.4.2      v forcats 0.3.0

## -- Conflicts -----
----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## x purrr::lift()    masks caret::lift()

library(psych)

##
## Attaching package: 'psych'

## The following objects are masked from 'package:ggplot2':
##
##   %+%, alpha

library(glmnet)

## Loading required package: Matrix

##
## Attaching package: 'Matrix'

## The following object is masked from 'package:tidyr':
##
##   expand
```

```
## Loading required package: foreach

##
## Attaching package: 'foreach'

## The following objects are masked from 'package:purrr':
##
##   accumulate, when

## Loaded glmnet 2.0-16

library(mlbench)
library(Boruta)

## Loading required package: ranger

library(MASS) # stepwise regression

##
## Attaching package: 'MASS'

## The following object is masked from 'package:dplyr':
##
##   select

library(leaps) # all subsets regression
library(randomForest)

## randomForest 4.6-14

## Type rfNews() to see new features/changes/bug fixes.

##
## Attaching package: 'randomForest'

## The following object is masked from 'package:ranger':
##
##   importance

## The following object is masked from 'package:psych':
##
##   outlier

## The following object is masked from 'package:dplyr':
##
##   combine

## The following object is masked from 'package:ggplot2':
##
##   margin

library(caret)
library(tidyverse)
library(psych)
```

```

library(glmnet)
library(mlbench)
library(Boruta)
library(MASS) # stepwise regression
library(leaps) # all subsets regression
library(randomForest)
library(MLmetrics)

##
## Attaching package: 'MLmetrics'

## The following object is masked from 'package:psych':
##
##      AUC

## The following objects are masked from 'package:caret':
##
##      MAE, RMSE

## The following object is masked from 'package:base':
##
##      Recall

library(e1071)
library(car) # VIF

## Loading required package: carData

##
## Attaching package: 'car'

## The following object is masked from 'package:psych':
##
##      logit

## The following object is masked from 'package:dplyr':
##
##      recode

## The following object is masked from 'package:purrr':
##
##      some

```

Import Clean Data

```

H_Clean<-read.csv( file = "C:\\Users\\Hyunkyung Kim\\Desktop\\CKME999\\136\\dataset\\all\\H_clean.csv")
H_Clean$MSSubClass<-as.factor(H_Clean$MSSubClass)
Train<-H_Clean[!is.na(H_Clean$SalePrice),]
Test<-H_Clean[is.na(H_Clean$SalePrice),]
actual<-read.csv( file = "C:\\Users\\Hyunkyung Kim\\Desktop\\CKME999\\136\\dataset\\all\\AMES_test.csv")[1461:2919,2] # Test Price

```

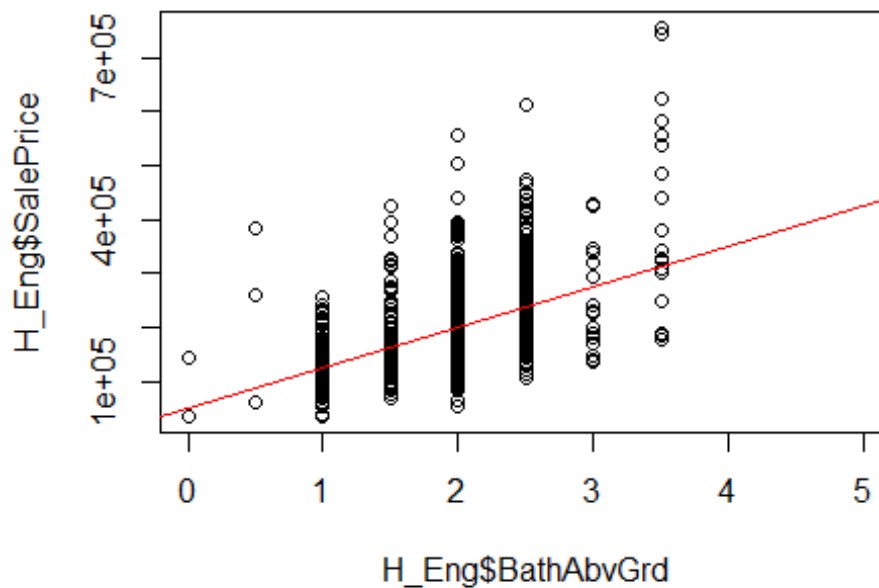
Utilities has extremely low variance where only 1 out 2919 observation is different. This is not really useful and I will drop this.

```
H_Eng<-H_Clean[,names(H_Clean)!="Utilities"]
```

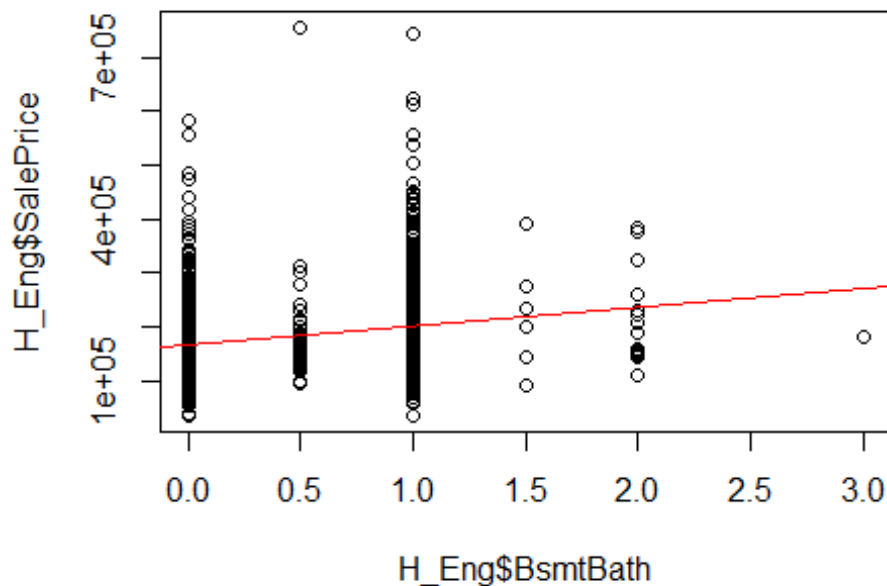
Add Full bath and half baths into one BathAbvGrd and BsmtFullBath+BsmtHalfBath into BsmtBath

```
H_Eng$BathAbvGrd<-H_Eng$FullBath+0.5*H_Eng$HalfBath  
H_Eng$BsmtBath<-H_Eng$BsmtFullBath+0.5*H_Eng$BsmtHalfBath
```

```
plot(H_Eng$BathAbvGrd, H_Eng$SalePrice)  
abline(lm(H_Eng$SalePrice~H_Eng$BathAbvGrd, data=H_Eng), col='red')
```



```
plot(H_Eng$BsmtBath, H_Eng$SalePrice)  
abline(lm(H_Eng$SalePrice~H_Eng$BsmtBath, data=H_Eng), col='red')
```



```
# Remove original columns
#H_Eng<-H_Eng[,!names(H_Clean) %in% c("FullBath", 'HalfBath', 'BathAbvGrd', 'BsmtBath')]
H_Eng<-subset(H_Eng, select=-c(FullBath,HalfBath,BathAbvGrd,BsmtBath))
```

Log Transformation due to Skewness

As Discussed in Step1, we have a few variables including response variable that has high skewness.

```
H_num<-H_Eng[,sapply(H_Eng,is.numeric)] #Numerical
H_num<-H_num[-c(1)]
H_cat<-H_Eng[,sapply(H_Eng[, -1],is.factor)] # Categorical
Check<-describe(H_num)[,c(11,12)]
Check[Check$skew>0.75 & Check$skew<1 ,]

##           skew kurtosis
## ExterQual  0.79      0.06
## BsmtUnfSF  0.92      0.40
## X2ndFlrSF  0.86     -0.43
## TotRmsAbvGrd 0.76      1.16

Check[Check$skew<(-1) ,]

##           skew kurtosis
## LandSlope -4.97     26.51
## BldgType  -2.16      3.16
```

```
## BsmtQual    -1.25    4.04
## BsmtCond    -3.63   17.06
## Functional  -4.96   26.95
## GarageQual  -3.27   10.14
## GarageCond  -3.39   10.62
## PavedDrive  -2.98    7.10
```

Below items all high skewness and may need transformation.- Skew over 1 - Removed ordinal items.

- skew kurtosis

LotFrontage 1.33 9.12 LotArea 12.82 264.31 MasVnrArea 2.61 9.31 ExterCond 1.32 6.27
 TotalBsmtSF 1.16 9.10 X1stFlrSF 1.47 6.94 LowQualFinSF 12.08 174.51 GrLivArea 1.27
 4.11 BsmtHalfBath 3.93 14.82 KitchenAbvGr 4.30 19.73 PavedDrive -2.98 7.10
 WoodDeckSF 1.84 6.72 OpenPorchSF 2.53 10.91 EnclosedPorch 4.00 28.31 X3SsnPorch
 11.37 149.05 ScreenPorch 3.94 17.73 PoolArea 16.89 297.91 PoolQC 15.86 255.66 MiscVal
 21.94 562.72 SalePrice 1.88 6.50

- now add

ExterQual 0.79 0.06 ** not ordinal items. BsmtUnfSF 0.92 0.40 X2ndFlrSF 0.86 -0.43
 TotRmsAbvGrd 0.76 1.16

Subset non-normal variables.

```
H_SkewedVar<-H_Eng[,c(
```

```
'LotFrontage',
'LotArea',
'MasVnrArea',
'TotalBsmtSF',
'X1stFlrSF',
'LowQualFinSF',
'GrLivArea',
'BsmtHalfBath',
'KitchenAbvGr',
'OpenPorchSF',
'EnclosedPorch',
'X3SsnPorch',
'ScreenPorch',
'PoolArea',
'MiscVal',
'SalePrice')]
```

```
H_Lo<-H_Eng
```

Apply Log only to the skewed items

```
H_Lo[,c(
```

```
'LotFrontage',  
'LotArea',  
'MasVnrArea',  
'BsmtFinSF1',  
'BsmtFinSF2',  
'TotalBsmtSF',  
'X1stFlrSF',  
'LowQualFinSF',  
'GrLivArea',  
'BsmtHalfBath',  
'KitchenAbvGr',  
'OpenPorchSF',  
'EnclosedPorch',  
'X3SsnPorch',  
'ScreenPorch',  
'PoolArea',  
'MiscVal',  
'SalePrice',  
  
'BsmtUnfSF',  
'X2ndFlrSF',  
'TotRmsAbvGrd'  
)<-log(1+H_Lo[,c(
```

```
'LotFrontage',  
'LotArea',  
'MasVnrArea',  
'BsmtFinSF1',  
'BsmtFinSF2',  
'TotalBsmtSF',  
'X1stFlrSF',  
'LowQualFinSF',  
'GrLivArea',  
'BsmtHalfBath',  
'KitchenAbvGr',  
'OpenPorchSF',  
'EnclosedPorch',  
'X3SsnPorch',  
'ScreenPorch',  
'PoolArea',  
'MiscVal',  
'SalePrice',  
'BsmtUnfSF',  
'X2ndFlrSF',  
'TotRmsAbvGrd'  
))
```

```
head(H_Lo)
```

##	Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape
## 1	1	2SNEW	RL	4.189655	9.042040	Pave	NoAlley	Reg
## 2	2	1SNEW	RL	4.394449	9.169623	Pave	NoAlley	Reg
## 3	3	2SNEW	RL	4.234107	9.328212	Pave	NoAlley	IR1
## 4	4	2SOLD	RL	4.110874	9.164401	Pave	NoAlley	IR1
## 5	5	2SNEW	RL	4.442651	9.565284	Pave	NoAlley	IR1
## 6	6	1NHFin	RL	4.454347	9.555064	Pave	NoAlley	IR1
##	LandContour	LotConfig	LandSlope	Neighborhood	Condition1	Condition2		
## 1	Lvl	Inside	3	CollgCr	Norm	Norm		
## 2	Lvl	FR2	3	Veenker	Feedr	Norm		
## 3	Lvl	Inside	3	CollgCr	Norm	Norm		
## 4	Lvl	Corner	3	Crawfor	Norm	Norm		
## 5	Lvl	FR2	3	NoRidge	Norm	Norm		
## 6	Lvl	Inside	3	Mitchel	Norm	Norm		
##	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	YearRemodAdd		
## 1	5	2Story	7	5	2003	2003		
## 2	5	1Story	6	8	1976	1976		
## 3	5	2Story	7	5	2001	2002		
## 4	5	2Story	7	5	1915	1970		
## 5	5	2Story	8	5	2000	2000		
## 6	5	1.5Fin	5	5	1993	1995		
##	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	MasVnrArea		
## 1	Gable	CompShg	VinylSd	VinylSd	BrkFace	5.283204		
## 2	Gable	CompShg	MetalSd	MetalSd	None	0.000000		
## 3	Gable	CompShg	VinylSd	VinylSd	BrkFace	5.093750		
## 4	Gable	CompShg	Wd Sdng	Wd Shng	None	0.000000		
## 5	Gable	CompShg	VinylSd	VinylSd	BrkFace	5.860786		
## 6	Gable	CompShg	VinylSd	VinylSd	None	0.000000		
##	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	BsmtExposure		
## 1	4	3	PConc	4	3	1		
## 2	3	3	CBlock	4	3	4		
## 3	4	3	PConc	4	3	2		
## 4	3	3	BrkTil	3	4	1		
## 5	4	3	PConc	4	3	3		
## 6	3	3	Wood	4	3	1		
##	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF		
## 1	6	6.561031	1	0	5.017280	6.753438		
## 2	5	6.886532	1	0	5.652489	7.141245		
## 3	6	6.188264	1	0	6.075346	6.825460		
## 4	5	5.379897	1	0	6.293419	6.629363		
## 5	6	6.486161	1	0	6.196444	7.044033		
## 6	6	6.597146	1	0	4.174387	6.680855		
##	Heating	HeatingQC	CentralAir	Electrical	X1stFlrSF	X2ndFlrSF	LowQualFinSF	
## 1	GasA	5	Y	SBrkr	6.753438	6.751101	0	
## 2	GasA	5	Y	SBrkr	7.141245	0.000000	0	
## 3	GasA	5	Y	SBrkr	6.825460	6.765039	0	
## 4	GasA	4	Y	SBrkr	6.869014	6.629363	0	
## 5	GasA	5	Y	SBrkr	7.044033	6.960348	0	
## 6	GasA	5	Y	SBrkr	6.680855	6.340359	0	
##	GrLivArea	BsmtFullBath	BsmtHalfBath	BedroomAbvGr	KitchenAbvGr			


```

## 1 7.444833      1  0.0000000      3  0.6931472
## 2 7.141245      0  0.6931472      3  0.6931472
## 3 7.488294      1  0.0000000      3  0.6931472
## 4 7.448916      1  0.0000000      3  0.6931472
## 5 7.695758      1  0.0000000      4  0.6931472
## 6 7.217443      1  0.0000000      1  0.6931472
##   KitchenQual TotRmsAbvGrd Functional Fireplaces FireplaceQu GarageType
## 1           4      2.197225          7           0           0      Attchd
## 2           3      1.945910          7           1           3      Attchd
## 3           4      1.945910          7           1           3      Attchd
## 4           4      2.079442          7           1           4      Detchd
## 5           4      2.302585          7           1           3      Attchd
## 6           3      1.791759          7           0           0      Attchd
##   GarageYrBlt GarageFinish GarageCars GarageArea GarageQual GarageCond
## 1          2003              2          2          548           3           3
## 2          1976              2          2          460           3           3
## 3          2001              2          2          608           3           3
## 4          1998              1          3          642           3           3
## 5          2000              2          3          836           3           3
## 6          1993              1          2          480           3           3
##   PavedDrive WoodDeckSF OpenPorchSF EnclosedPorch X3SsnPorch ScreenPorch
## 1           2           0      4.127134      0.000000      0.000000           0
## 2           2          298      0.000000      0.000000      0.000000           0
## 3           2           0      3.761200      0.000000      0.000000           0
## 4           2           0      3.583519      5.609472      0.000000           0
## 5           2          192      4.442651      0.000000      0.000000           0
## 6           2          40      3.433987      0.000000      5.771441           0
##   PoolArea PoolQC Fence MiscFeature MiscVal MoSold YrSold SaleType
## 1           0       0 NoFence NoMiscFeature 0.000000      2   2008      WD
## 2           0       0 NoFence NoMiscFeature 0.000000      5   2007      WD
## 3           0       0 NoFence NoMiscFeature 0.000000      9   2008      WD
## 4           0       0 NoFence NoMiscFeature 0.000000      2   2006      WD
## 5           0       0 NoFence NoMiscFeature 0.000000     12   2008      WD
## 6           0       0 MnPrv      Shed 6.552508     10   2009      WD
##   SaleCondition SalePrice
## 1          Normal  12.24770
## 2          Normal  12.10902
## 3          Normal  12.31717
## 4         Abnorml  11.84940
## 5          Normal  12.42922
## 6          Normal  11.87061

```

`head(H_Eng)`

```

##   Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape
## 1  1      2SNEW      RL           65    8450  Pave NoAlley      Reg
## 2  2      1SNEW      RL           80    9600  Pave NoAlley      Reg
## 3  3      2SNEW      RL           68   11250  Pave NoAlley      IR1
## 4  4      2SOLD      RL           60    9550  Pave NoAlley      IR1
## 5  5      2SNEW      RL           84   14260  Pave NoAlley      IR1

```

## 6	6	1NHFin	RL	85	14115	Pave NoAlley	IR1
##	LandContour	LotConfig	LandSlope	Neighborhood	Condition1	Condition2	
## 1	Lvl	Inside	3	CollgCr	Norm	Norm	
## 2	Lvl	FR2	3	Veenker	Feedr	Norm	
## 3	Lvl	Inside	3	CollgCr	Norm	Norm	
## 4	Lvl	Corner	3	Crawfor	Norm	Norm	
## 5	Lvl	FR2	3	NoRidge	Norm	Norm	
## 6	Lvl	Inside	3	Mitchel	Norm	Norm	
##	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt	YearRemodAdd	
## 1	5	2Story	7	5	2003	2003	
## 2	5	1Story	6	8	1976	1976	
## 3	5	2Story	7	5	2001	2002	
## 4	5	2Story	7	5	1915	1970	
## 5	5	2Story	8	5	2000	2000	
## 6	5	1.5Fin	5	5	1993	1995	
##	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType	MasVnrArea	
## 1	Gable	CompShg	VinylSd	VinylSd	BrkFace	196	
## 2	Gable	CompShg	MetalSd	MetalSd	None	0	
## 3	Gable	CompShg	VinylSd	VinylSd	BrkFace	162	
## 4	Gable	CompShg	Wd Sdng	Wd Shng	None	0	
## 5	Gable	CompShg	VinylSd	VinylSd	BrkFace	350	
## 6	Gable	CompShg	VinylSd	VinylSd	None	0	
##	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	BsmtExposure	
## 1	4	3	PConc	4	3	1	
## 2	3	3	CBlock	4	3	4	
## 3	4	3	PConc	4	3	2	
## 4	3	3	BrkTil	3	4	1	
## 5	4	3	PConc	4	3	3	
## 6	3	3	Wood	4	3	1	
##	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF	
## 1	6	706	1	0	150	856	
## 2	5	978	1	0	284	1262	
## 3	6	486	1	0	434	920	
## 4	5	216	1	0	540	756	
## 5	6	655	1	0	490	1145	
## 6	6	732	1	0	64	796	
##	Heating	HeatingQC	CentralAir	Electrical	X1stFlrSF	X2ndFlrSF	LowQualFinSF
## 1	GasA	5	Y	SBrkr	856	854	0
## 2	GasA	5	Y	SBrkr	1262	0	0
## 3	GasA	5	Y	SBrkr	920	866	0
## 4	GasA	4	Y	SBrkr	961	756	0
## 5	GasA	5	Y	SBrkr	1145	1053	0
## 6	GasA	5	Y	SBrkr	796	566	0
##	GrLivArea	BsmtFullBath	BsmtHalfBath	BedroomAbvGr	KitchenAbvGr		
## 1	1710	1	0	3	1		
## 2	1262	0	1	3	1		
## 3	1786	1	0	3	1		
## 4	1717	1	0	3	1		
## 5	2198	1	0	4	1		
## 6	1362	1	0	1	1		

	KitchenQual	TotRmsAbvGrd	Functional	Fireplaces	FireplaceQu	GarageType
## 1	4	8	7	0	0	Attchd
## 2	3	6	7	1	3	Attchd
## 3	4	6	7	1	3	Attchd
## 4	4	7	7	1	4	Detchd
## 5	4	9	7	1	3	Attchd
## 6	3	5	7	0	0	Attchd

	GarageYrBlt	GarageFinish	GarageCars	GarageArea	GarageQual	GarageCond
## 1	2003	2	2	548	3	3
## 2	1976	2	2	460	3	3
## 3	2001	2	2	608	3	3
## 4	1998	1	3	642	3	3
## 5	2000	2	3	836	3	3
## 6	1993	1	2	480	3	3

	PavedDrive	WoodDeckSF	OpenPorchSF	EnclosedPorch	X3SsnPorch	ScreenPorch
## 1	2	0	61	0	0	0
## 2	2	298	0	0	0	0
## 3	2	0	42	0	0	0
## 4	2	0	35	272	0	0
## 5	2	192	84	0	0	0
## 6	2	40	30	0	320	0

	PoolArea	PoolQC	Fence	MiscFeature	MiscVal	MoSold	YrSold	SaleType
## 1	0	0	NoFence	NoMiscFeature	0	2	2008	WD
## 2	0	0	NoFence	NoMiscFeature	0	5	2007	WD
## 3	0	0	NoFence	NoMiscFeature	0	9	2008	WD
## 4	0	0	NoFence	NoMiscFeature	0	2	2006	WD
## 5	0	0	NoFence	NoMiscFeature	0	12	2008	WD
## 6	0	0	MnPrv	Shed	700	10	2009	WD

	SaleCondition	SalePrice
## 1	Normal	208500
## 2	Normal	181500
## 3	Normal	223500
## 4	Abnorml	140000
## 5	Normal	250000
## 6	Normal	143000

Feature Selection done on the engineered training set.

```

nulllog<-lm(SalePrice~1,data=H_Lo[1:1460,])
fulllog<-lm(SalePrice~.,data=H_Lo[1:1460,])

StepF_Log<-stepAIC(nulllog, scope=list(lower=nulllog, upper=fulllog), direction='forward', trace=F)
StepB_Log<-stepAIC(fulllog, direction='backward', trace=F)
StepS_Log<-stepAIC(fulllog,direction='both', trace=F)

summary(StepF_Log)

##
## Call:
## lm(formula = SalePrice ~ OverallQual + GrLivArea + Neighborhood +

```

```

##      MSSubClass + BsmtFinSF1 + OverallCond + GarageCars + YearBuilt +
##      RoofMatl + LotArea + MSZoning + BsmtExposure + Condition2 +
##      Functional + SaleCondition + KitchenQual + Condition1 + Exterior1st +
##      BsmtFullBath + Fireplaces + BsmtQual + HeatingQC + PoolQC +
##      ScreenPorch + Heating + LotConfig + GarageArea + Foundation +
##      WoodDeckSF + CentralAir + SaleType + X1stFlrSF + Street +
##      YearRemodAdd + MoSold + KitchenAbvGr + MasVnrType, data = H_Lo[1:1460,

##    ])
##
## Residuals:
##      Min        1Q    Median        3Q        Max
## -0.67331 -0.04917  0.00068  0.05551  0.50000
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.097e-01  7.983e-01  -1.014  0.310655
## OverallQual     4.659e-02  4.274e-03  10.899 < 2e-16 ***
## GrLivArea      4.392e-01  2.944e-02  14.917 < 2e-16 ***
## NeighborhoodBlueste  1.907e-02  8.603e-02   0.222  0.824572
## NeighborhoodBrDale  2.763e-02  4.965e-02   0.557  0.577914
## NeighborhoodBrkSide  4.474e-02  4.046e-02   1.106  0.269035
## NeighborhoodClearCr  1.648e-02  4.001e-02   0.412  0.680473
## NeighborhoodCollgCr -3.839e-02  3.212e-02  -1.195  0.232136
## NeighborhoodCrawfor  9.331e-02  3.713e-02   2.513  0.012088 *
## NeighborhoodEdwards -7.222e-02  3.542e-02  -2.039  0.041683 *
## NeighborhoodGilbert -3.391e-02  3.417e-02  -0.992  0.321296
## NeighborhoodIDOTRR  -2.144e-02  4.656e-02  -0.460  0.645295
## NeighborhoodMeadowV -1.054e-01  5.285e-02  -1.994  0.046365 *
## NeighborhoodMitchel -6.177e-02  3.588e-02  -1.722  0.085390 .
## NeighborhoodNames  -3.797e-02  3.422e-02  -1.110  0.267317
## NeighborhoodNoRidge  7.158e-02  3.605e-02   1.986  0.047255 *
## NeighborhoodNPkVill  6.739e-02  4.930e-02   1.367  0.171905
## NeighborhoodNridgHt  8.237e-02  3.240e-02   2.542  0.011134 *
## NeighborhoodNWAmes  -5.001e-02  3.531e-02  -1.416  0.156903
## NeighborhoodOldTown -2.880e-02  4.152e-02  -0.694  0.488056
## NeighborhoodSawyer  -2.287e-02  3.586e-02  -0.638  0.523797
## NeighborhoodSawyerW -2.553e-02  3.478e-02  -0.734  0.463138
## NeighborhoodSomerst -4.250e-03  4.020e-02  -0.106  0.915819
## NeighborhoodStoneBr  1.277e-01  3.636e-02   3.511  0.000461 ***
## NeighborhoodSWISU   -4.765e-03  4.143e-02  -0.115  0.908452
## NeighborhoodTimber  -2.777e-02  3.608e-02  -0.770  0.441593
## NeighborhoodVeenker  3.504e-03  4.655e-02   0.075  0.940019
## MSSubClass1NHFin    5.831e-02  2.910e-02   2.004  0.045286 *
## MSSubClass1NHUnf    1.128e-01  4.492e-02   2.511  0.012154 *
## MSSubClass1SFA      1.994e-02  6.314e-02   0.316  0.752153
## MSSubClass1SNEW     9.145e-02  2.938e-02   3.113  0.001894 **
## MSSubClass1SOLD     6.271e-02  3.563e-02   1.760  0.078599 .
## MSSubClass1SPUD     6.010e-02  2.887e-02   2.082  0.037573 *
## MSSubClass2FCNV     2.927e-02  3.801e-02   0.770  0.441454

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## MSSubClass2NHS	4.378e-02	4.227e-02	1.036	0.300471	
## MSSubClass2SNEW	5.799e-02	2.454e-02	2.363	0.018271	*
## MSSubClass2SOLD	6.631e-02	3.303e-02	2.007	0.044912	*
## MSSubClassDUPL	4.895e-02	3.879e-02	1.262	0.207186	
## MSSubClassMLPUD	7.054e-02	4.559e-02	1.547	0.122017	
## MSSubClassSPL	5.637e-02	3.063e-02	1.840	0.065918	.
## MSSubClassSPLF	8.082e-02	3.706e-02	2.181	0.029377	*
## BsmtFinSF1	9.266e-03	1.332e-03	6.959	5.36e-12	***
## OverallCond	3.774e-02	3.560e-03	10.602	< 2e-16	***
## GarageCars	2.515e-02	9.481e-03	2.653	0.008085	**
## YearBuilt	2.036e-03	3.299e-04	6.173	8.90e-10	***
## RoofMatlCompShg	1.701e+00	1.242e-01	13.695	< 2e-16	***
## RoofMatlMembran	1.818e+00	1.666e-01	10.914	< 2e-16	***
## RoofMatlMetal	1.777e+00	1.667e-01	10.661	< 2e-16	***
## RoofMatlRoll	1.674e+00	1.678e-01	9.974	< 2e-16	***
## RoofMatlTar&Grv	1.666e+00	1.282e-01	12.995	< 2e-16	***
## RoofMatlWdShake	1.721e+00	1.354e-01	12.708	< 2e-16	***
## RoofMatlWdShngl	1.877e+00	1.318e-01	14.240	< 2e-16	***
## LotArea	8.194e-02	1.066e-02	7.684	2.99e-14	***
## MSZoningFV	4.389e-01	5.244e-02	8.370	< 2e-16	***
## MSZoningRH	3.956e-01	5.214e-02	7.587	6.13e-14	***
## MSZoningRL	3.879e-01	4.401e-02	8.813	< 2e-16	***
## MSZoningRM	3.598e-01	4.124e-02	8.725	< 2e-16	***
## BsmtExposure	1.671e-02	3.550e-03	4.707	2.78e-06	***
## Condition2Feedr	2.491e-02	9.715e-02	0.256	0.797699	
## Condition2Norm	2.507e-02	8.226e-02	0.305	0.760631	
## Condition2PosA	3.921e-01	1.383e-01	2.836	0.004637	**
## Condition2PosN	-6.227e-01	1.170e-01	-5.324	1.19e-07	***
## Condition2RR Ae	-2.724e-02	1.375e-01	-0.198	0.843007	
## Condition2RR An	-1.320e-02	1.358e-01	-0.097	0.922583	
## Condition2RR Nn	8.330e-02	1.139e-01	0.731	0.464740	
## Functional	3.697e-02	4.750e-03	7.783	1.42e-14	***
## SaleConditionAdjLand	1.300e-01	5.993e-02	2.170	0.030219	*
## SaleConditionAlloca	5.613e-02	3.653e-02	1.537	0.124581	
## SaleConditionFamily	8.511e-03	2.712e-02	0.314	0.753670	
## SaleConditionNormal	6.695e-02	1.256e-02	5.330	1.15e-07	***
## SaleConditionPartial	-5.873e-02	6.579e-02	-0.893	0.372186	
## KitchenQual	2.334e-02	6.917e-03	3.374	0.000762	***
## Condition1Feedr	3.796e-02	2.171e-02	1.749	0.080518	.
## Condition1Norm	8.490e-02	1.797e-02	4.723	2.57e-06	***
## Condition1PosA	4.559e-02	4.345e-02	1.049	0.294167	
## Condition1PosN	8.633e-02	3.237e-02	2.667	0.007744	**
## Condition1RR Ae	-4.985e-02	3.849e-02	-1.295	0.195502	
## Condition1RR An	5.332e-02	2.979e-02	1.790	0.073692	.
## Condition1RR Ne	1.441e-02	7.893e-02	0.183	0.855129	
## Condition1RR Nn	1.215e-01	5.511e-02	2.205	0.027601	*
## Exterior1stAsphShn	-4.854e-02	1.124e-01	-0.432	0.666044	
## Exterior1stBrkComm	-1.753e-01	8.393e-02	-2.089	0.036904	*
## Exterior1stBrkFace	9.787e-02	3.152e-02	3.105	0.001940	**
## Exterior1stCBlock	-1.103e-02	1.120e-01	-0.099	0.921546	

## Exterior1stCemntBd	5.570e-02	3.256e-02	1.710	0.087414	.
## Exterior1stHdBoard	6.282e-03	2.849e-02	0.220	0.825528	
## Exterior1stImStucc	-1.561e-02	1.104e-01	-0.141	0.887640	
## Exterior1stMetalSd	4.073e-02	2.764e-02	1.473	0.140888	
## Exterior1stPlywood	4.895e-03	2.990e-02	0.164	0.869985	
## Exterior1stStone	7.646e-04	8.570e-02	0.009	0.992882	
## Exterior1stStucco	4.125e-02	3.513e-02	1.174	0.240559	
## Exterior1stVinylSd	2.591e-02	2.790e-02	0.929	0.353310	
## Exterior1stWd Sdng	4.984e-03	2.754e-02	0.181	0.856409	
## Exterior1stWdShing	1.146e-02	3.453e-02	0.332	0.739923	
## BsmtFullBath	2.922e-02	7.338e-03	3.982	7.21e-05	***
## Fireplaces	2.217e-02	5.849e-03	3.790	0.000158	***
## BsmtQual	2.217e-02	6.444e-03	3.439	0.000601	***
## HeatingQC	1.575e-02	4.197e-03	3.753	0.000182	***
## PoolQC	4.825e-02	1.189e-02	4.059	5.21e-05	***
## ScreenPorch	8.223e-03	2.101e-03	3.914	9.52e-05	***
## HeatingGasA	6.869e-02	1.106e-01	0.621	0.534675	
## HeatingGasW	1.462e-01	1.134e-01	1.289	0.197547	
## HeatingGrav	-1.699e-02	1.175e-01	-0.145	0.885050	
## HeatingOthW	-1.253e-02	1.365e-01	-0.092	0.926908	
## HeatingWall	1.384e-01	1.262e-01	1.096	0.273064	
## LotConfigCulDSac	1.710e-02	1.382e-02	1.238	0.215995	
## LotConfigFR2	-3.745e-02	1.766e-02	-2.120	0.034157	*
## LotConfigFR3	-9.599e-02	5.644e-02	-1.701	0.089213	.
## LotConfigInside	-1.167e-02	7.656e-03	-1.524	0.127847	
## GarageArea	9.522e-05	3.216e-05	2.961	0.003123	**
## FoundationCBlock	2.099e-02	1.368e-02	1.534	0.125311	
## FoundationPConc	3.769e-02	1.498e-02	2.516	0.011971	*
## FoundationSlab	2.639e-02	3.517e-02	0.750	0.453174	
## FoundationStone	1.113e-01	4.586e-02	2.426	0.015384	*
## FoundationWood	-1.032e-01	6.399e-02	-1.613	0.107055	
## WoodDeckSF	6.905e-05	2.544e-05	2.714	0.006727	**
## CentralAirY	3.686e-02	1.592e-02	2.316	0.020726	*
## SaleTypeCon	6.035e-02	7.946e-02	0.759	0.447694	
## SaleTypeConLD	1.290e-01	4.255e-02	3.033	0.002467	**
## SaleTypeConLI	1.109e-03	5.139e-02	0.022	0.982780	
## SaleTypeConLw	-1.158e-04	5.269e-02	-0.002	0.998246	
## SaleTypeCWD	4.248e-02	5.742e-02	0.740	0.459520	
## SaleTypeNew	1.737e-01	6.816e-02	2.548	0.010935	*
## SaleTypeOth	7.019e-02	6.508e-02	1.079	0.281003	
## SaleTypeWD	-6.954e-03	1.841e-02	-0.378	0.705701	
## X1stFlrSF	5.117e-02	2.742e-02	1.866	0.062244	.
## StreetPave	9.488e-02	4.896e-02	1.938	0.052824	.
## YearRemodAdd	4.240e-04	2.317e-04	1.830	0.067513	.
## MoSold	-1.787e-03	1.065e-03	-1.677	0.093813	.
## KitchenAbvGr	-9.023e-02	5.928e-02	-1.522	0.128227	
## MasVnrTypeBrkFace	5.423e-02	2.892e-02	1.875	0.060993	.
## MasVnrTypeNone	4.872e-02	2.850e-02	1.709	0.087639	.
## MasVnrTypeStone	6.871e-02	3.077e-02	2.233	0.025703	*
## ---					

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1047 on 1327 degrees of freedom
## Multiple R-squared:  0.9376, Adjusted R-squared:  0.9313
## F-statistic: 150.9 on 132 and 1327 DF,  p-value: < 2.2e-16
```

```
summary(StepB_Log)
```

```
##
## Call:
## lm(formula = SalePrice ~ MSZoning + LotArea + Street + LotConfig +
##      Neighborhood + Condition1 + Condition2 + BldgType + OverallQual +
##      OverallCond + YearBuilt + YearRemodAdd + RoofMatl + Exterior1st +
##      Foundation + BsmtQual + BsmtExposure + BsmtFinSF1 + Heating +
##      HeatingQC + CentralAir + X1stFlrSF + GrLivArea + BsmtFullBath +
##      KitchenAbvGr + KitchenQual + Functional + Fireplaces + GarageType +
##      GarageFinish + GarageCars + GarageArea + GarageQual + WoodDeckSF +
##      ScreenPorch + PoolQC + MoSold + SaleType + SaleCondition,
##      data = H_Lo[1:1460, ])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.67962 -0.04954  0.00030  0.05688  0.49948
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.418e-01  7.374e-01  -1.142  0.253804
## MSZoningFV      4.154e-01  5.146e-02   8.072  1.53e-15 ***
## MSZoningRH      4.039e-01  5.176e-02   7.803  1.21e-14 ***
## MSZoningRL      3.807e-01  4.367e-02   8.717  < 2e-16 ***
## MSZoningRM      3.597e-01  4.108e-02   8.757  < 2e-16 ***
## LotArea        8.103e-02  1.024e-02   7.915  5.16e-15 ***
## StreetPave      7.749e-02  4.950e-02   1.566  0.117695
## LotConfigCulDSac 1.535e-02  1.376e-02   1.115  0.264923
## LotConfigFR2    -3.713e-02  1.759e-02  -2.111  0.034975 *
## LotConfigFR3    -9.979e-02  5.625e-02  -1.774  0.076277 .
## LotConfigInside -1.194e-02  7.679e-03  -1.555  0.120272
## NeighborhoodBlueste -9.840e-03  8.292e-02  -0.119  0.905557
## NeighborhoodBrDale  7.673e-03  4.470e-02   0.172  0.863722
## NeighborhoodBrkSide  2.544e-02  3.878e-02   0.656  0.511902
## NeighborhoodClearCr  7.317e-03  3.915e-02   0.187  0.851789
## NeighborhoodCollgCr -3.904e-02  3.117e-02  -1.252  0.210681
## NeighborhoodCrawfor  9.905e-02  3.590e-02   2.759  0.005882 **
## NeighborhoodEdwards -7.581e-02  3.387e-02  -2.239  0.025346 *
## NeighborhoodGilbert -4.219e-02  3.280e-02  -1.286  0.198591
## NeighborhoodIDOTRR -3.832e-02  4.470e-02  -0.857  0.391425
## NeighborhoodMeadowV -1.066e-01  4.701e-02  -2.267  0.023568 *
## NeighborhoodMitchel -6.324e-02  3.463e-02  -1.826  0.068031 .
## NeighborhoodNames  -3.963e-02  3.279e-02  -1.209  0.227046
## NeighborhoodNoRidge  6.553e-02  3.505e-02   1.870  0.061751 .
```

## NeighborhoodNPkVill	5.743e-02	4.748e-02	1.210	0.226613	
## NeighborhoodNridgHt	8.593e-02	3.117e-02	2.757	0.005910	**
## NeighborhoodNWAmes	-5.136e-02	3.421e-02	-1.502	0.133463	
## NeighborhoodOldTown	-4.040e-02	4.006e-02	-1.008	0.313425	
## NeighborhoodSawyer	-2.729e-02	3.449e-02	-0.791	0.428845	
## NeighborhoodSawyerW	-2.980e-02	3.377e-02	-0.882	0.377705	
## NeighborhoodSomerst	3.134e-03	3.821e-02	0.082	0.934647	
## NeighborhoodStoneBr	1.209e-01	3.549e-02	3.407	0.000675	***
## NeighborhoodSWISU	-1.386e-02	4.019e-02	-0.345	0.730344	
## NeighborhoodTimber	-2.921e-02	3.487e-02	-0.838	0.402276	
## NeighborhoodVeenker	-6.048e-04	4.569e-02	-0.013	0.989441	
## Condition1Feedr	4.201e-02	2.147e-02	1.957	0.050578	.
## Condition1Norm	8.757e-02	1.762e-02	4.969	7.61e-07	***
## Condition1PosA	4.867e-02	4.301e-02	1.132	0.257983	
## Condition1PosN	8.905e-02	3.183e-02	2.798	0.005217	**
## Condition1RR Ae	-4.265e-02	3.844e-02	-1.110	0.267408	
## Condition1RR An	5.609e-02	2.946e-02	1.904	0.057171	.
## Condition1RR Ne	1.901e-02	7.871e-02	0.242	0.809184	
## Condition1RR Nn	1.326e-01	5.540e-02	2.393	0.016867	*
## Condition2Feedr	1.235e-02	9.649e-02	0.128	0.898160	
## Condition2Norm	1.392e-02	8.227e-02	0.169	0.865629	
## Condition2PosA	3.476e-01	1.361e-01	2.555	0.010740	*
## Condition2PosN	-6.445e-01	1.170e-01	-5.510	4.30e-08	***
## Condition2RR Ae	-2.485e-02	1.356e-01	-0.183	0.854615	
## Condition2RR An	-4.150e-02	1.351e-01	-0.307	0.758813	
## Condition2RR Nn	7.438e-02	1.138e-01	0.654	0.513530	
## BldgType	1.241e-02	4.665e-03	2.660	0.007909	**
## OverallQual	4.741e-02	4.237e-03	11.191	< 2e-16	***
## OverallCond	3.683e-02	3.550e-03	10.374	< 2e-16	***
## YearBuilt	2.032e-03	2.917e-04	6.964	5.17e-12	***
## YearRemodAdd	4.225e-04	2.307e-04	1.832	0.067236	.
## RoofMatlCompShg	1.690e+00	1.228e-01	13.757	< 2e-16	***
## RoofMatlMembran	1.834e+00	1.651e-01	11.106	< 2e-16	***
## RoofMatlMetal	1.754e+00	1.647e-01	10.646	< 2e-16	***
## RoofMatlRoll	1.688e+00	1.647e-01	10.246	< 2e-16	***
## RoofMatlTar&Grv	1.657e+00	1.267e-01	13.079	< 2e-16	***
## RoofMatlWdShake	1.691e+00	1.332e-01	12.689	< 2e-16	***
## RoofMatlWdShngl	1.853e+00	1.302e-01	14.231	< 2e-16	***
## Exterior1stAsphShn	-6.924e-02	1.119e-01	-0.619	0.536325	
## Exterior1stBrkComm	-1.830e-01	8.421e-02	-2.174	0.029902	*
## Exterior1stBrkFace	9.149e-02	3.123e-02	2.930	0.003453	**
## Exterior1stCBlock	-2.764e-02	1.128e-01	-0.245	0.806408	
## Exterior1stCemntBd	4.938e-02	3.242e-02	1.523	0.128046	
## Exterior1stHdBoard	8.119e-04	2.839e-02	0.029	0.977191	
## Exterior1stImStucc	-2.129e-02	1.101e-01	-0.193	0.846737	
## Exterior1stMetalSd	3.305e-02	2.761e-02	1.197	0.231491	
## Exterior1stPlywood	-1.392e-03	2.979e-02	-0.047	0.962749	
## Exterior1stStone	-7.851e-03	8.523e-02	-0.092	0.926623	
## Exterior1stStucco	2.964e-02	3.478e-02	0.852	0.394151	
## Exterior1stVinylSd	1.973e-02	2.778e-02	0.710	0.477668	

## Exterior1stWd Sdng	-1.649e-03	2.754e-02	-0.060	0.952274	
## Exterior1stWdShing	-1.151e-03	3.434e-02	-0.034	0.973275	
## FoundationCBlock	2.648e-02	1.350e-02	1.961	0.050077	.
## FoundationPConc	4.013e-02	1.494e-02	2.686	0.007314	**
## FoundationSlab	4.113e-02	3.506e-02	1.173	0.240986	
## FoundationStone	1.240e-01	4.557e-02	2.721	0.006587	**
## FoundationWood	-9.518e-02	6.423e-02	-1.482	0.138653	
## BsmtQual	2.283e-02	6.412e-03	3.561	0.000383	***
## BsmtExposure	1.552e-02	3.330e-03	4.661	3.46e-06	***
## BsmtFinSF1	9.064e-03	1.327e-03	6.829	1.29e-11	***
## HeatingGasA	5.030e-02	1.101e-01	0.457	0.648003	
## HeatingGasW	1.320e-01	1.130e-01	1.168	0.242828	
## HeatingGrav	-3.260e-02	1.170e-01	-0.279	0.780578	
## HeatingOthW	-1.976e-02	1.363e-01	-0.145	0.884753	
## HeatingWall	9.832e-02	1.252e-01	0.785	0.432585	
## HeatingQC	1.603e-02	4.181e-03	3.834	0.000132	***
## CentralAirY	3.858e-02	1.564e-02	2.467	0.013731	*
## X1stFlrSF	9.998e-02	1.491e-02	6.705	2.97e-11	***
## GrLivArea	3.930e-01	1.549e-02	25.374	< 2e-16	***
## BsmtFullBath	2.993e-02	7.300e-03	4.099	4.39e-05	***
## KitchenAbvGr	-9.581e-02	4.387e-02	-2.184	0.029127	*
## KitchenQual	2.508e-02	6.870e-03	3.651	0.000272	***
## Functional	3.854e-02	4.633e-03	8.319	< 2e-16	***
## Fireplaces	2.054e-02	5.798e-03	3.543	0.000409	***
## GarageTypeAttchd	8.810e-02	4.718e-02	1.867	0.062058	.
## GarageTypeBasement	6.644e-02	5.352e-02	1.241	0.214698	
## GarageTypeBuiltIn	9.228e-02	4.895e-02	1.885	0.059602	.
## GarageTypeCarPort	5.106e-02	5.953e-02	0.858	0.391177	
## GarageTypeDetchd	1.003e-01	4.684e-02	2.141	0.032456	*
## GarageTypeNoGarage	1.674e-01	6.140e-02	2.726	0.006490	**
## GarageFinish	7.124e-03	5.180e-03	1.375	0.169245	
## GarageCars	2.887e-02	9.821e-03	2.940	0.003344	**
## GarageArea	8.932e-05	3.300e-05	2.706	0.006889	**
## GarageQual	1.954e-02	1.283e-02	1.523	0.128096	
## WoodDeckSF	6.922e-05	2.533e-05	2.733	0.006368	**
## ScreenPorch	8.744e-03	2.087e-03	4.189	2.99e-05	***
## PoolQC	4.495e-02	1.171e-02	3.840	0.000129	***
## MoSold	-1.671e-03	1.061e-03	-1.575	0.115476	
## SaleTypeCon	6.443e-02	7.932e-02	0.812	0.416777	
## SaleTypeConLD	1.076e-01	4.186e-02	2.570	0.010275	*
## SaleTypeConLI	-1.200e-02	5.158e-02	-0.233	0.816077	
## SaleTypeConLw	-2.313e-03	5.238e-02	-0.044	0.964782	
## SaleTypeCWD	3.390e-02	5.725e-02	0.592	0.553935	
## SaleTypeNew	1.561e-01	6.795e-02	2.297	0.021780	*
## SaleTypeOth	6.692e-02	6.507e-02	1.028	0.303923	
## SaleTypeWD	-1.314e-02	1.845e-02	-0.712	0.476450	
## SaleConditionAdjLand	1.241e-01	5.794e-02	2.142	0.032336	*
## SaleConditionAlloca	5.460e-02	3.637e-02	1.501	0.133527	
## SaleConditionFamily	1.594e-02	2.706e-02	0.589	0.555899	
## SaleConditionNormal	6.918e-02	1.253e-02	5.522	4.02e-08	***

```
## SaleConditionPartial -4.354e-02  6.558e-02  -0.664 0.506854
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1046 on 1335 degrees of freedom
## Multiple R-squared:  0.9373, Adjusted R-squared:  0.9314
## F-statistic: 160.8 on 124 and 1335 DF,  p-value: < 2.2e-16
```

summary(StepS_Log)

```
##
## Call:
## lm(formula = SalePrice ~ MSZoning + LotArea + Street + LotConfig +
##      Neighborhood + Condition1 + Condition2 + BldgType + OverallQual +
##      OverallCond + YearBuilt + YearRemodAdd + RoofMatl + Exterior1st +
##      Foundation + BsmtQual + BsmtExposure + BsmtFinSF1 + Heating +
##      HeatingQC + CentralAir + X1stFlrSF + GrLivArea + BsmtFullBath +
##      KitchenAbvGr + KitchenQual + Functional + Fireplaces + GarageType +
##      GarageFinish + GarageCars + GarageArea + GarageQual + WoodDeckSF +
##      ScreenPorch + PoolQC + MoSold + SaleType + SaleCondition,
##      data = H_Lo[1:1460, ])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.67962 -0.04954  0.00030  0.05688  0.49948
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.418e-01  7.374e-01  -1.142 0.253804
## MSZoningFV      4.154e-01  5.146e-02   8.072 1.53e-15 ***
## MSZoningRH      4.039e-01  5.176e-02   7.803 1.21e-14 ***
## MSZoningRL      3.807e-01  4.367e-02   8.717 < 2e-16 ***
## MSZoningRM      3.597e-01  4.108e-02   8.757 < 2e-16 ***
## LotArea        8.103e-02  1.024e-02   7.915 5.16e-15 ***
## StreetPave      7.749e-02  4.950e-02   1.566 0.117695
## LotConfigCulDSac 1.535e-02  1.376e-02   1.115 0.264923
## LotConfigFR2    -3.713e-02  1.759e-02  -2.111 0.034975 *
## LotConfigFR3    -9.979e-02  5.625e-02  -1.774 0.076277 .
## LotConfigInside -1.194e-02  7.679e-03  -1.555 0.120272
## NeighborhoodBlueste -9.840e-03  8.292e-02  -0.119 0.905557
## NeighborhoodBrDale  7.673e-03  4.470e-02   0.172 0.863722
## NeighborhoodBrkSide  2.544e-02  3.878e-02   0.656 0.511902
## NeighborhoodClearCr  7.317e-03  3.915e-02   0.187 0.851789
## NeighborhoodCollgCr -3.904e-02  3.117e-02  -1.252 0.210681
## NeighborhoodCrawfor  9.905e-02  3.590e-02   2.759 0.005882 **
## NeighborhoodEdwards -7.581e-02  3.387e-02  -2.239 0.025346 *
## NeighborhoodGilbert -4.219e-02  3.280e-02  -1.286 0.198591
## NeighborhoodIDOTRR -3.832e-02  4.470e-02  -0.857 0.391425
## NeighborhoodMeadowV -1.066e-01  4.701e-02  -2.267 0.023568 *
## NeighborhoodMitchel -6.324e-02  3.463e-02  -1.826 0.068031 .
```

## NeighborhoodNames	-3.963e-02	3.279e-02	-1.209	0.227046	
## NeighborhoodNoRidge	6.553e-02	3.505e-02	1.870	0.061751	.
## NeighborhoodNPkVill	5.743e-02	4.748e-02	1.210	0.226613	
## NeighborhoodNridgHt	8.593e-02	3.117e-02	2.757	0.005910	**
## NeighborhoodNWAmes	-5.136e-02	3.421e-02	-1.502	0.133463	
## NeighborhoodOldTown	-4.040e-02	4.006e-02	-1.008	0.313425	
## NeighborhoodSawyer	-2.729e-02	3.449e-02	-0.791	0.428845	
## NeighborhoodSawyerW	-2.980e-02	3.377e-02	-0.882	0.377705	
## NeighborhoodSomerst	3.134e-03	3.821e-02	0.082	0.934647	
## NeighborhoodStoneBr	1.209e-01	3.549e-02	3.407	0.000675	***
## NeighborhoodSWISU	-1.386e-02	4.019e-02	-0.345	0.730344	
## NeighborhoodTimber	-2.921e-02	3.487e-02	-0.838	0.402276	
## NeighborhoodVeenker	-6.048e-04	4.569e-02	-0.013	0.989441	
## Condition1Feedr	4.201e-02	2.147e-02	1.957	0.050578	.
## Condition1Norm	8.757e-02	1.762e-02	4.969	7.61e-07	***
## Condition1PosA	4.867e-02	4.301e-02	1.132	0.257983	
## Condition1PosN	8.905e-02	3.183e-02	2.798	0.005217	**
## Condition1RRAE	-4.265e-02	3.844e-02	-1.110	0.267408	
## Condition1RRAn	5.609e-02	2.946e-02	1.904	0.057171	.
## Condition1RRNe	1.901e-02	7.871e-02	0.242	0.809184	
## Condition1RRNn	1.326e-01	5.540e-02	2.393	0.016867	*
## Condition2Feedr	1.235e-02	9.649e-02	0.128	0.898160	
## Condition2Norm	1.392e-02	8.227e-02	0.169	0.865629	
## Condition2PosA	3.476e-01	1.361e-01	2.555	0.010740	*
## Condition2PosN	-6.445e-01	1.170e-01	-5.510	4.30e-08	***
## Condition2RRAE	-2.485e-02	1.356e-01	-0.183	0.854615	
## Condition2RRAn	-4.150e-02	1.351e-01	-0.307	0.758813	
## Condition2RRNn	7.438e-02	1.138e-01	0.654	0.513530	
## BldgType	1.241e-02	4.665e-03	2.660	0.007909	**
## OverallQual	4.741e-02	4.237e-03	11.191	< 2e-16	***
## OverallCond	3.683e-02	3.550e-03	10.374	< 2e-16	***
## YearBuilt	2.032e-03	2.917e-04	6.964	5.17e-12	***
## YearRemodAdd	4.225e-04	2.307e-04	1.832	0.067236	.
## RoofMatlCompShg	1.690e+00	1.228e-01	13.757	< 2e-16	***
## RoofMatlMembran	1.834e+00	1.651e-01	11.106	< 2e-16	***
## RoofMatlMetal	1.754e+00	1.647e-01	10.646	< 2e-16	***
## RoofMatlRoll	1.688e+00	1.647e-01	10.246	< 2e-16	***
## RoofMatlTar&Grv	1.657e+00	1.267e-01	13.079	< 2e-16	***
## RoofMatlWdShake	1.691e+00	1.332e-01	12.689	< 2e-16	***
## RoofMatlWdShngl	1.853e+00	1.302e-01	14.231	< 2e-16	***
## Exterior1stAsphShn	-6.924e-02	1.119e-01	-0.619	0.536325	
## Exterior1stBrkComm	-1.830e-01	8.421e-02	-2.174	0.029902	*
## Exterior1stBrkFace	9.149e-02	3.123e-02	2.930	0.003453	**
## Exterior1stCBlock	-2.764e-02	1.128e-01	-0.245	0.806408	
## Exterior1stCemntBd	4.938e-02	3.242e-02	1.523	0.128046	
## Exterior1stHdBoard	8.119e-04	2.839e-02	0.029	0.977191	
## Exterior1stImStucc	-2.129e-02	1.101e-01	-0.193	0.846737	
## Exterior1stMetalSd	3.305e-02	2.761e-02	1.197	0.231491	
## Exterior1stPlywood	-1.392e-03	2.979e-02	-0.047	0.962749	
## Exterior1stStone	-7.851e-03	8.523e-02	-0.092	0.926623	

## Exterior1stStucco	2.964e-02	3.478e-02	0.852	0.394151	
## Exterior1stVinylSd	1.973e-02	2.778e-02	0.710	0.477668	
## Exterior1stWd Sdng	-1.649e-03	2.754e-02	-0.060	0.952274	
## Exterior1stWdShing	-1.151e-03	3.434e-02	-0.034	0.973275	
## FoundationCBlock	2.648e-02	1.350e-02	1.961	0.050077	.
## FoundationPConc	4.013e-02	1.494e-02	2.686	0.007314	**
## FoundationSlab	4.113e-02	3.506e-02	1.173	0.240986	
## FoundationStone	1.240e-01	4.557e-02	2.721	0.006587	**
## FoundationWood	-9.518e-02	6.423e-02	-1.482	0.138653	
## BsmtQual	2.283e-02	6.412e-03	3.561	0.000383	***
## BsmtExposure	1.552e-02	3.330e-03	4.661	3.46e-06	***
## BsmtFinSF1	9.064e-03	1.327e-03	6.829	1.29e-11	***
## HeatingGasA	5.030e-02	1.101e-01	0.457	0.648003	
## HeatingGasW	1.320e-01	1.130e-01	1.168	0.242828	
## HeatingGrav	-3.260e-02	1.170e-01	-0.279	0.780578	
## HeatingOthW	-1.976e-02	1.363e-01	-0.145	0.884753	
## HeatingWall	9.832e-02	1.252e-01	0.785	0.432585	
## HeatingQC	1.603e-02	4.181e-03	3.834	0.000132	***
## CentralAirY	3.858e-02	1.564e-02	2.467	0.013731	*
## X1stFlrSF	9.998e-02	1.491e-02	6.705	2.97e-11	***
## GrLivArea	3.930e-01	1.549e-02	25.374	< 2e-16	***
## BsmtFullBath	2.993e-02	7.300e-03	4.099	4.39e-05	***
## KitchenAbvGr	-9.581e-02	4.387e-02	-2.184	0.029127	*
## KitchenQual	2.508e-02	6.870e-03	3.651	0.000272	***
## Functional	3.854e-02	4.633e-03	8.319	< 2e-16	***
## Fireplaces	2.054e-02	5.798e-03	3.543	0.000409	***
## GarageTypeAttchd	8.810e-02	4.718e-02	1.867	0.062058	.
## GarageTypeBasment	6.644e-02	5.352e-02	1.241	0.214698	
## GarageTypeBuiltIn	9.228e-02	4.895e-02	1.885	0.059602	.
## GarageTypeCarPort	5.106e-02	5.953e-02	0.858	0.391177	
## GarageTypeDetchd	1.003e-01	4.684e-02	2.141	0.032456	*
## GarageTypeNoGarage	1.674e-01	6.140e-02	2.726	0.006490	**
## GarageFinish	7.124e-03	5.180e-03	1.375	0.169245	
## GarageCars	2.887e-02	9.821e-03	2.940	0.003344	**
## GarageArea	8.932e-05	3.300e-05	2.706	0.006889	**
## GarageQual	1.954e-02	1.283e-02	1.523	0.128096	
## WoodDeckSF	6.922e-05	2.533e-05	2.733	0.006368	**
## ScreenPorch	8.744e-03	2.087e-03	4.189	2.99e-05	***
## PoolQC	4.495e-02	1.171e-02	3.840	0.000129	***
## MoSold	-1.671e-03	1.061e-03	-1.575	0.115476	
## SaleTypeCon	6.443e-02	7.932e-02	0.812	0.416777	
## SaleTypeConLD	1.076e-01	4.186e-02	2.570	0.010275	*
## SaleTypeConLI	-1.200e-02	5.158e-02	-0.233	0.816077	
## SaleTypeConLw	-2.313e-03	5.238e-02	-0.044	0.964782	
## SaleTypeCWD	3.390e-02	5.725e-02	0.592	0.553935	
## SaleTypeNew	1.561e-01	6.795e-02	2.297	0.021780	*
## SaleTypeOth	6.692e-02	6.507e-02	1.028	0.303923	
## SaleTypeWD	-1.314e-02	1.845e-02	-0.712	0.476450	
## SaleConditionAdjLand	1.241e-01	5.794e-02	2.142	0.032336	*
## SaleConditionAlloca	5.460e-02	3.637e-02	1.501	0.133527	

```
## SaleConditionFamily    1.594e-02  2.706e-02    0.589 0.555899
## SaleConditionNormal    6.918e-02  1.253e-02    5.522 4.02e-08 ***
## SaleConditionPartial  -4.354e-02  6.558e-02   -0.664 0.506854
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1046 on 1335 degrees of freedom
## Multiple R-squared:  0.9373, Adjusted R-squared:  0.9314
## F-statistic: 160.8 on 124 and 1335 DF,  p-value: < 2.2e-16
```

```
Train<-H_Clean[!is.na(H_Clean$SalePrice),-1]
```

```
StepF_Log$call
```

```
## lm(formula = SalePrice ~ OverallQual + GrLivArea + Neighborhood +
##     MSSubClass + BsmtFinSF1 + OverallCond + GarageCars + YearBuilt +
##     RoofMatl + LotArea + MSZoning + BsmtExposure + Condition2 +
##     Functional + SaleCondition + KitchenQual + Condition1 + Exterior1st +
##     BsmtFullBath + Fireplaces + BsmtQual + HeatingQC + PoolQC +
##     ScreenPorch + Heating + LotConfig + GarageArea + Foundation +
##     WoodDeckSF + CentralAir + SaleType + X1stFlrSF + Street +
##     YearRemodAdd + MoSold + KitchenAbvGr + MasVnrType, data = H_Lo[1:1460,
##
##     ])
```

```
StepB_Log$call
```

```
## lm(formula = SalePrice ~ MSZoning + LotArea + Street + LotConfig +
##     Neighborhood + Condition1 + Condition2 + BldgType + OverallQual +
##     OverallCond + YearBuilt + YearRemodAdd + RoofMatl + Exterior1st +
##     Foundation + BsmtQual + BsmtExposure + BsmtFinSF1 + Heating +
##     HeatingQC + CentralAir + X1stFlrSF + GrLivArea + BsmtFullBath +
##     KitchenAbvGr + KitchenQual + Functional + Fireplaces + GarageType +
##     GarageFinish + GarageCars + GarageArea + GarageQual + WoodDeckSF +
##     ScreenPorch + PoolQC + MoSold + SaleType + SaleCondition,
##     data = H_Lo[1:1460, ])
```

```
StepS_Log$call
```

```
## lm(formula = SalePrice ~ MSZoning + LotArea + Street + LotConfig +
##     Neighborhood + Condition1 + Condition2 + BldgType + OverallQual +
##     OverallCond + YearBuilt + YearRemodAdd + RoofMatl + Exterior1st +
##     Foundation + BsmtQual + BsmtExposure + BsmtFinSF1 + Heating +
##     HeatingQC + CentralAir + X1stFlrSF + GrLivArea + BsmtFullBath +
##     KitchenAbvGr + KitchenQual + Functional + Fireplaces + GarageType +
##     GarageFinish + GarageCars + GarageArea + GarageQual + WoodDeckSF +
##     ScreenPorch + PoolQC + MoSold + SaleType + SaleCondition,
##     data = H_Lo[1:1460, ])
```

```

#summary(StepB_Log)
#summary(StepS_Log)
identical(StepB_Log$call, StepS_Log$call) # Shows they are same.

## [1] TRUE

identical(StepF_Log$call, StepS_Log$call)

## [1] FALSE

TR<-1:1460
TR_RmvOut<-TR[-c(524,1299)]

a<-lm(SalePrice~.,data=H_Lo[TR,-1])) # Removing ID
b<-lm(StepF_Log$call,data=H_Lo[TR,-1])
c<-lm(StepS_Log$call,data=H_Lo[TR,-1])
par(mfrow=c(2,2))
plot(a, sub='ALL Var')

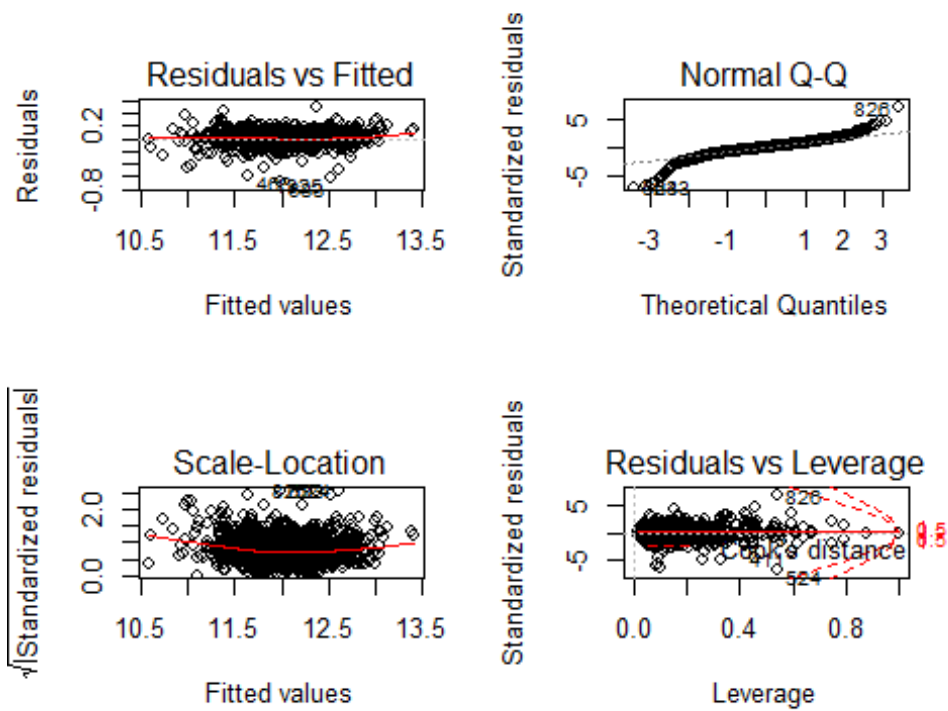
## Warning: not plotting observations with leverage one:
## 121, 272, 347, 584, 596, 1004, 1012, 1188, 1231, 1271, 1276, 1299, 1322,
1371, 1387

## Warning: not plotting observations with leverage one:
## 121, 272, 347, 584, 596, 1004, 1012, 1188, 1231, 1271, 1276, 1299, 1322,
1371, 1387

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

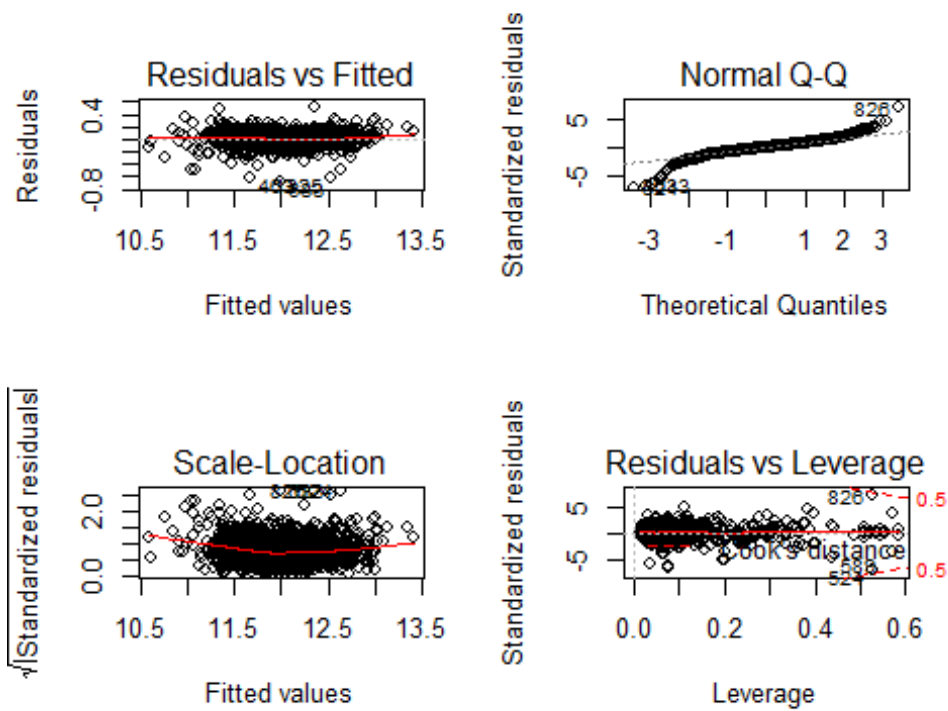
```



```
plot(b, sub='Foward Sel')
```

```
## Warning: not plotting observations with leverage one:
## 121, 272, 584, 1004, 1012, 1188, 1231, 1276, 1299, 1322, 1371

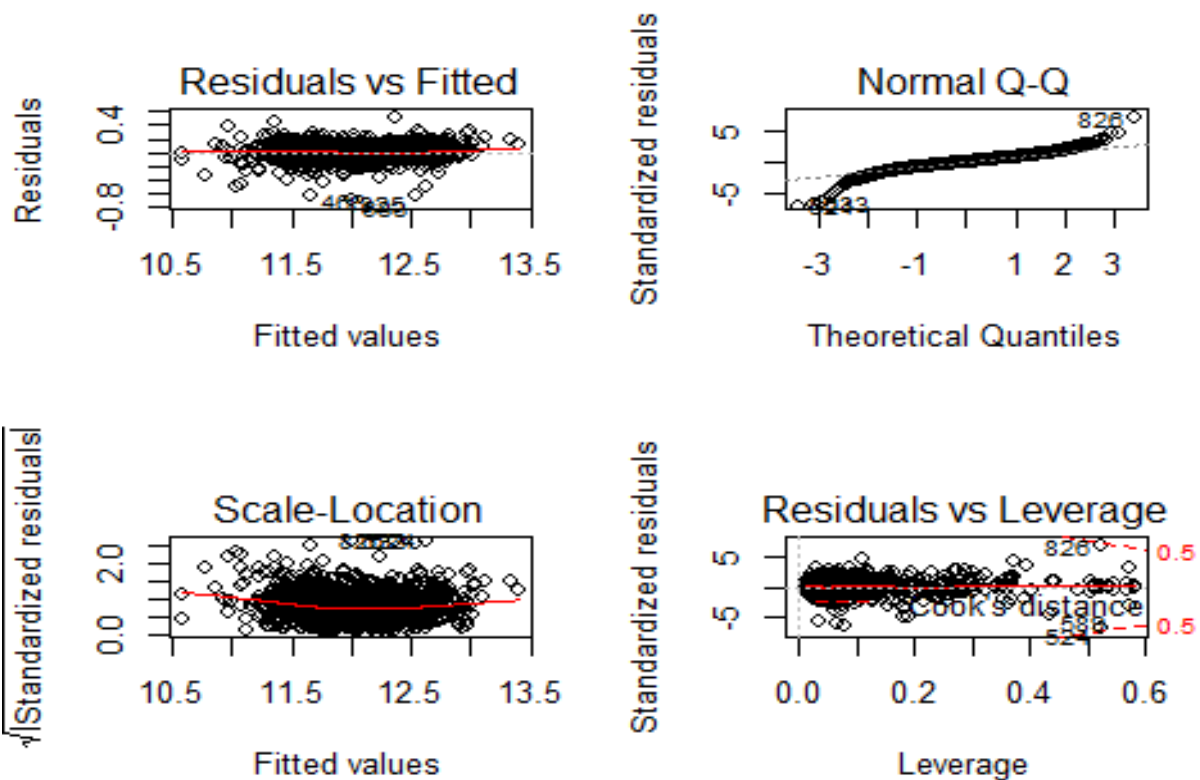
## Warning: not plotting observations with leverage one:
## 121, 272, 584, 1004, 1012, 1188, 1231, 1276, 1299, 1322, 1371
```



```
plot(c, sub='Backwards E1')
```

```
## Warning: not plotting observations with leverage one:
## 121, 272, 584, 1004, 1012, 1188, 1231, 1276, 1299, 1322, 1371

## Warning: not plotting observations with leverage one:
## 121, 272, 584, 1004, 1012, 1188, 1231, 1276, 1299, 1322, 1371
```

```
plot(a,which=c(4))
plot(b,which=c(4))

#pdtrain<-predict(a,newdata=H_Lo[!is.na(H_Lo$SalePrice),])
pd_Log_LM_All<-predict(a,newdata=H_Lo[is.na(H_Lo$SalePrice),])

## Warning in predict.lm(a, newdata = H_Lo[is.na(H_Lo$SalePrice), ]):
## prediction from a rank-deficient fit may be misleading

pd_Log_LM_f2<-predict(b,newdata=H_Lo[is.na(H_Lo$SalePrice),])
pd_Log_LM_f3<-predict(c,newdata=H_Lo[is.na(H_Lo$SalePrice),])
#trainSP<-H_Lo$SalePrice[!is.na(H_Lo$SalePrice)]

#plot(exp(trainSP),exp(pdtrain), col='blue', ylab='Prediction')
#points(actual,exp(pdtest), col='red')
#abline(b=1,a=0)
#RMSE(exp(trainSP),exp(pdtrain))
#RMSE(actual,exp(pdtest))

print("RMSLE for All Var - Log")

## [1] "RMSLE for All Var - Log"
```

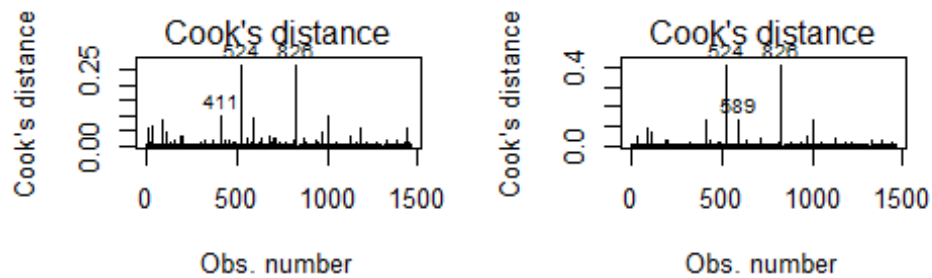
```

RMSE(pd_Log_LM_All,log(actual))
## [1] 0.1368791

RMSE(pd_Log_LM_f2,log(actual))
## [1] 0.1338353

RMSE(pd_Log_LM_f3,log(actual))
## [1] 0.1334131

```



Try without

outliers

```

summary(a)

##
## Call:
## lm(formula = SalePrice ~ ., data = H_Lo[TR, -1])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.67396 -0.04631  0.00222  0.05248  0.48404
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.955e+00  4.708e+00   1.265 0.206185
## MSSubClass1NHFin    7.060e-02  7.237e-02   0.975 0.329515
## MSSubClass1NHUnf   -1.140e-01  1.212e-01  -0.941 0.346844

```

## MSSubClass1SFA	-4.537e-02	1.084e-01	-0.418	0.675682	
## MSSubClass1SNEW	8.489e-02	6.916e-02	1.227	0.219907	
## MSSubClass1SOLD	5.203e-02	7.313e-02	0.711	0.476934	
## MSSubClass1SPUD	7.871e-02	4.299e-02	1.831	0.067330	.
## MSSubClass2FCONV	1.934e-02	6.038e-02	0.320	0.748794	
## MSSubClass2NHS	-4.257e-02	8.848e-02	-0.481	0.630547	
## MSSubClass2SNEW	3.810e-02	6.047e-02	0.630	0.528790	
## MSSubClass2SOLD	6.222e-02	6.481e-02	0.960	0.337167	
## MSSubClassDUPL	6.346e-02	5.649e-02	1.123	0.261472	
## MSSubClassMLPUD	1.219e-01	6.248e-02	1.951	0.051230	.
## MSSubClassSPL	1.022e-01	8.181e-02	1.249	0.211797	
## MSSubClassSPLF	8.269e-02	8.303e-02	0.996	0.319520	
## MSZoningFV	4.714e-01	5.514e-02	8.550	< 2e-16	***
## MSZoningRH	4.260e-01	5.486e-02	7.765	1.70e-14	***
## MSZoningRL	4.245e-01	4.688e-02	9.056	< 2e-16	***
## MSZoningRM	4.015e-01	4.426e-02	9.073	< 2e-16	***
## LotFrontage	2.112e-02	1.651e-02	1.279	0.201086	
## LotArea	8.364e-02	1.253e-02	6.674	3.72e-11	***
## StreetPave	6.663e-02	5.275e-02	1.263	0.206823	
## AlleyNoAlley	4.428e-03	1.925e-02	0.230	0.818144	
## AlleyPave	4.931e-02	2.787e-02	1.769	0.077124	.
## LotShapeIR2	2.871e-02	1.910e-02	1.503	0.133106	
## LotShapeIR3	2.952e-02	3.918e-02	0.754	0.451288	
## LotShapeReg	1.296e-02	7.352e-03	1.763	0.078229	.
## LandContourHLS	2.931e-02	2.340e-02	1.253	0.210501	
## LandContourLow	-1.155e-02	2.863e-02	-0.404	0.686601	
## LandContourLvl	2.385e-02	1.683e-02	1.417	0.156687	
## LotConfigCulDSac	2.925e-02	1.688e-02	1.733	0.083348	.
## LotConfigFR2	-3.597e-02	1.848e-02	-1.946	0.051862	.
## LotConfigFR3	-1.143e-01	5.762e-02	-1.984	0.047490	*
## LotConfigInside	-9.662e-03	8.228e-03	-1.174	0.240497	
## LandSlope	-1.500e-03	1.524e-02	-0.098	0.921600	
## NeighborhoodBlueste	6.923e-02	9.074e-02	0.763	0.445615	
## NeighborhoodBrDale	2.939e-02	5.373e-02	0.547	0.584419	
## NeighborhoodBrkSide	2.225e-02	4.399e-02	0.506	0.613051	
## NeighborhoodClearCr	1.264e-02	4.353e-02	0.290	0.771668	
## NeighborhoodCollgCr	-3.464e-02	3.480e-02	-0.996	0.319660	
## NeighborhoodCrawfor	1.004e-01	4.070e-02	2.466	0.013790	*
## NeighborhoodEdwards	-8.106e-02	3.807e-02	-2.129	0.033424	*
## NeighborhoodGilbert	-3.408e-02	3.674e-02	-0.928	0.353764	
## NeighborhoodIDOTRR	-3.391e-02	4.999e-02	-0.678	0.497675	
## NeighborhoodMeadowV	-9.887e-02	5.692e-02	-1.737	0.082633	.
## NeighborhoodMitchel	-5.559e-02	3.881e-02	-1.432	0.152302	
## NeighborhoodNames	-4.261e-02	3.709e-02	-1.149	0.250884	
## NeighborhoodNoRidge	6.784e-02	3.941e-02	1.721	0.085414	.
## NeighborhoodNPkVill	4.150e-02	6.405e-02	0.648	0.517191	
## NeighborhoodNridgHt	8.428e-02	3.530e-02	2.388	0.017096	*
## NeighborhoodNWAmes	-4.947e-02	3.821e-02	-1.295	0.195699	
## NeighborhoodOldTown	-3.949e-02	4.506e-02	-0.876	0.380963	
## NeighborhoodSawyer	-2.619e-02	3.835e-02	-0.683	0.494824	

## NeighborhoodSawyerW	-2.121e-02	3.716e-02	-0.571	0.568261	
## NeighborhoodSomerst	-5.873e-03	4.241e-02	-0.138	0.889888	
## NeighborhoodStoneBr	1.417e-01	3.895e-02	3.639	0.000285	***
## NeighborhoodSWISU	-3.261e-03	4.527e-02	-0.072	0.942587	
## NeighborhoodTimber	-2.480e-02	3.878e-02	-0.639	0.522669	
## NeighborhoodVeenker	1.256e-02	4.912e-02	0.256	0.798157	
## Condition1Feedr	3.450e-02	2.270e-02	1.520	0.128831	
## Condition1Norm	8.692e-02	1.880e-02	4.624	4.15e-06	***
## Condition1PosA	3.877e-02	4.537e-02	0.855	0.392913	
## Condition1PosN	8.596e-02	3.377e-02	2.545	0.011036	*
## Condition1RR Ae	-5.332e-02	3.972e-02	-1.342	0.179704	
## Condition1RR An	6.036e-02	3.148e-02	1.917	0.055439	.
## Condition1RR Ne	1.501e-02	8.057e-02	0.186	0.852215	
## Condition1RR Nn	1.030e-01	5.856e-02	1.758	0.078981	.
## Condition2Feedr	1.395e-01	1.151e-01	1.212	0.225759	
## Condition2Norm	9.989e-02	1.017e-01	0.982	0.326193	
## Condition2PosA	4.314e-01	1.562e-01	2.761	0.005840	**
## Condition2PosN	-5.375e-01	1.318e-01	-4.079	4.82e-05	***
## Condition2RR Ae	-2.115e-01	2.444e-01	-0.865	0.387045	
## Condition2RR An	4.383e-02	1.498e-01	0.293	0.769924	
## Condition2RR Nn	1.709e-01	1.303e-01	1.311	0.190070	
## BldgType	6.955e-03	1.618e-02	0.430	0.667388	
## HouseStyle1.5Unf	2.464e-01	1.049e-01	2.350	0.018933	*
## HouseStyle1Story	2.785e-02	4.523e-02	0.616	0.538078	
## HouseStyle2.5Fin	9.312e-02	7.479e-02	1.245	0.213379	
## HouseStyle2.5Unf	1.266e-01	7.148e-02	1.771	0.076861	.
## HouseStyle2Story	2.201e-02	3.702e-02	0.595	0.552255	
## HouseStyleSFoyer	5.783e-03	5.824e-02	0.099	0.920912	
## HouseStyleSLvl	-2.327e-02	6.062e-02	-0.384	0.701174	
## OverallQual	4.113e-02	4.645e-03	8.855	< 2e-16	***
## OverallCond	3.761e-02	3.935e-03	9.558	< 2e-16	***
## YearBuilt	2.108e-03	3.831e-04	5.501	4.57e-08	***
## YearRemodAdd	4.751e-04	2.471e-04	1.923	0.054768	.
## RoofStyleGable	3.689e-03	8.257e-02	0.045	0.964375	
## RoofStyleGambrel	-4.374e-02	9.040e-02	-0.484	0.628582	
## RoofStyleHip	1.553e-02	8.277e-02	0.188	0.851164	
## RoofStyleMansard	2.878e-02	9.601e-02	0.300	0.764430	
## RoofStyleShed	2.363e-01	1.702e-01	1.389	0.165220	
## RoofMatlCompShg	1.718e+00	1.361e-01	12.628	< 2e-16	***
## RoofMatlMembran	1.896e+00	1.951e-01	9.720	< 2e-16	***
## RoofMatlMetal	1.782e+00	1.985e-01	8.975	< 2e-16	***
## RoofMatlRoll	1.639e+00	1.780e-01	9.207	< 2e-16	***
## RoofMatlTar&Grv	1.697e+00	1.597e-01	10.628	< 2e-16	***
## RoofMatlWdShake	1.689e+00	1.520e-01	11.111	< 2e-16	***
## RoofMatlWdShngl	1.908e+00	1.445e-01	13.209	< 2e-16	***
## Exterior1stAsphShn	-9.239e-02	1.513e-01	-0.611	0.541588	
## Exterior1stBrkComm	-3.157e-01	1.270e-01	-2.486	0.013042	*
## Exterior1stBrkFace	6.916e-02	5.816e-02	1.189	0.234623	
## Exterior1stCBlock	-4.370e-02	1.163e-01	-0.376	0.707230	
## Exterior1stCemntBd	-5.485e-02	8.587e-02	-0.639	0.523052	

## Exterior1stHdBoard	-2.918e-02	5.863e-02	-0.498	0.618761	
## Exterior1stImStucc	-8.538e-02	1.281e-01	-0.666	0.505294	
## Exterior1stMetalSd	1.632e-02	6.675e-02	0.245	0.806851	
## Exterior1stPlywood	-2.283e-02	5.789e-02	-0.394	0.693406	
## Exterior1stStone	-4.263e-02	1.082e-01	-0.394	0.693650	
## Exterior1stStucco	2.101e-02	6.371e-02	0.330	0.741642	
## Exterior1stVinylSd	-9.268e-03	6.143e-02	-0.151	0.880104	
## Exterior1stWd Sdng	-6.144e-02	5.643e-02	-1.089	0.276406	
## Exterior1stWdShing	-1.169e-02	6.107e-02	-0.191	0.848175	
## Exterior2ndAsphShn	7.348e-02	1.015e-01	0.724	0.469075	
## Exterior2ndBrk Cmn	1.191e-01	9.365e-02	1.272	0.203566	
## Exterior2ndBrkFace	-6.928e-04	5.936e-02	-0.012	0.990691	
## Exterior2ndCBlock	NA	NA	NA	NA	
## Exterior2ndCmentBd	1.088e-01	8.417e-02	1.293	0.196299	
## Exterior2ndHdBoard	3.634e-02	5.570e-02	0.652	0.514217	
## Exterior2ndImStucc	6.716e-02	6.475e-02	1.037	0.299864	
## Exterior2ndMetalSd	2.138e-02	6.473e-02	0.330	0.741251	
## Exterior2ndOther	-1.003e-01	1.246e-01	-0.806	0.420606	
## Exterior2ndPlywood	2.673e-02	5.410e-02	0.494	0.621380	
## Exterior2ndStone	2.620e-02	7.768e-02	0.337	0.735923	
## Exterior2ndStucco	2.160e-04	6.173e-02	0.003	0.997209	
## Exterior2ndVinylSd	3.331e-02	5.883e-02	0.566	0.571415	
## Exterior2ndWd Sdng	7.197e-02	5.378e-02	1.338	0.181012	
## Exterior2ndWd Shng	2.253e-02	5.653e-02	0.398	0.690333	
## MasVnrTypeBrkFace	4.441e-02	3.024e-02	1.469	0.142194	
## MasVnrTypeNone	5.469e-02	4.075e-02	1.342	0.179743	
## MasVnrTypeStone	6.418e-02	3.209e-02	2.000	0.045685	*
## MasVnrArea	3.533e-03	5.384e-03	0.656	0.511827	
## ExterQual	2.236e-03	9.565e-03	0.234	0.815212	
## ExterCond	-1.183e-02	9.597e-03	-1.233	0.217862	
## FoundationCBlock	2.309e-02	1.451e-02	1.591	0.111811	
## FoundationPConc	4.063e-02	1.560e-02	2.605	0.009310	**
## FoundationSlab	6.109e-02	4.511e-02	1.354	0.175858	
## FoundationStone	1.128e-01	4.994e-02	2.259	0.024083	*
## FoundationWood	-1.070e-01	6.747e-02	-1.586	0.113036	
## BsmtQual	1.244e-02	8.206e-03	1.516	0.129666	
## BsmtCond	-6.147e-03	1.126e-02	-0.546	0.585197	
## BsmtExposure	1.617e-02	3.858e-03	4.192	2.96e-05	***
## BsmtFinType1	1.773e-03	3.390e-03	0.523	0.601140	
## BsmtFinSF1	7.132e-03	2.514e-03	2.836	0.004637	**
## BsmtFinType2	2.107e-03	7.842e-03	0.269	0.788264	
## BsmtFinSF2	-2.813e-03	3.842e-03	-0.732	0.464181	
## BsmtUnfSF	-4.949e-03	2.700e-03	-1.833	0.067016	.
## TotalBsmtSF	1.701e-02	7.978e-03	2.133	0.033151	*
## HeatingGasA	1.219e-02	1.164e-01	0.105	0.916605	
## HeatingGasW	9.816e-02	1.197e-01	0.820	0.412255	
## HeatingGrav	-9.541e-02	1.247e-01	-0.765	0.444262	
## HeatingOthW	-6.390e-02	1.429e-01	-0.447	0.654914	
## HeatingWall	6.492e-02	1.348e-01	0.482	0.630107	
## HeatingQC	1.468e-02	4.361e-03	3.366	0.000786	***

## CentralAirY	4.418e-02	1.745e-02	2.531	0.011485	*
## ElectricalFuseF	-1.448e-02	2.618e-02	-0.553	0.580140	
## ElectricalFuseP	-8.806e-03	7.717e-02	-0.114	0.909163	
## ElectricalMix	-2.982e-02	1.187e-01	-0.251	0.801710	
## ElectricalSBrkr	-1.736e-02	1.352e-02	-1.284	0.199550	
## X1stFlrSF	3.637e-02	4.532e-02	0.802	0.422420	
## X2ndFlrSF	2.659e-03	5.814e-03	0.457	0.647451	
## LowQualFinSF	-6.090e-03	6.204e-03	-0.982	0.326513	
## GrLivArea	4.224e-01	5.346e-02	7.903	5.97e-15	***
## BsmtFullBath	2.946e-02	8.304e-03	3.547	0.000404	***
## BsmtHalfBath	5.431e-03	1.973e-02	0.275	0.783140	
## BedroomAbvGr	-6.480e-03	6.196e-03	-1.046	0.295901	
## KitchenAbvGr	-9.608e-02	6.832e-02	-1.406	0.159874	
## KitchenQual	2.518e-02	7.415e-03	3.396	0.000706	***
## TotRmsAbvGrd	4.133e-02	3.197e-02	1.293	0.196244	
## Functional	3.419e-02	5.133e-03	6.662	4.04e-11	***
## Fireplaces	2.188e-02	9.918e-03	2.206	0.027544	*
## FireplaceQu	-1.909e-04	3.618e-03	-0.053	0.957917	
## GarageTypeAttchd	1.042e-01	5.012e-02	2.080	0.037751	*
## GarageTypeBasement	9.805e-02	5.801e-02	1.690	0.091235	.
## GarageTypeBuiltIn	1.072e-01	5.215e-02	2.055	0.040048	*
## GarageTypeCarPort	1.012e-01	6.536e-02	1.549	0.121702	
## GarageTypeDetchd	1.219e-01	4.998e-02	2.439	0.014871	*
## GarageTypeNoGarage	2.345e-01	7.061e-02	3.321	0.000923	***
## GarageYrBlt	-8.767e-05	2.709e-04	-0.324	0.746242	
## GarageFinish	6.436e-03	5.461e-03	1.179	0.238796	
## GarageCars	2.747e-02	1.019e-02	2.695	0.007142	**
## GarageArea	9.212e-05	3.576e-05	2.576	0.010108	*
## GarageQual	2.422e-02	1.747e-02	1.386	0.165910	
## GarageCond	8.692e-03	1.815e-02	0.479	0.632040	
## PavedDrive	5.508e-03	7.564e-03	0.728	0.466598	
## WoodDeckSF	8.456e-05	2.678e-05	3.158	0.001627	**
## OpenPorchSF	1.776e-03	1.723e-03	1.031	0.302890	
## EnclosedPorch	2.963e-03	2.052e-03	1.444	0.148940	
## X3SsnPorch	4.407e-03	4.604e-03	0.957	0.338580	
## ScreenPorch	9.393e-03	2.193e-03	4.284	1.98e-05	***
## PoolArea	-1.627e-02	2.863e-02	-0.568	0.569815	
## PoolQC	7.997e-02	4.534e-02	1.764	0.078041	.
## FenceGdWo	-2.448e-02	2.221e-02	-1.102	0.270539	
## FenceMnPrv	1.368e-02	1.815e-02	0.753	0.451404	
## FenceMnWw	5.371e-03	3.750e-02	0.143	0.886134	
## FenceNoFence	1.354e-02	1.652e-02	0.819	0.412747	
## MiscFeatureNoMiscFeature	-1.975e-01	1.691e-01	-1.168	0.243062	
## MiscFeatureOthr	-1.279e-01	1.587e-01	-0.806	0.420367	
## MiscFeatureShed	-9.447e-02	1.175e-01	-0.804	0.421646	
## MiscFeatureTenC	-1.398e-01	1.847e-01	-0.757	0.449250	
## MiscVal	-1.662e-02	1.355e-02	-1.226	0.220316	
## MoSold	-1.514e-03	1.107e-03	-1.367	0.171914	
## YrSold	-3.445e-03	2.310e-03	-1.491	0.136089	
## SaleTypeCon	8.367e-02	8.066e-02	1.037	0.299810	

```
## SaleTypeConLD          1.387e-01  4.449e-02   3.117 0.001868 **
## SaleTypeConLI         -6.487e-04  5.276e-02  -0.012 0.990192
## SaleTypeConLw        -2.245e-03  5.436e-02  -0.041 0.967064
## SaleTypeCWD           6.375e-02  5.870e-02   1.086 0.277688
## SaleTypeNew           1.621e-01  6.953e-02   2.331 0.019890 *
## SaleTypeOth           1.761e-02  7.340e-02   0.240 0.810449
## SaleTypeWD           -5.866e-03  1.903e-02  -0.308 0.757913
## SaleConditionAdjLand   1.331e-01  6.689e-02   1.990 0.046862 *
## SaleConditionAlloca    4.850e-02  3.944e-02   1.230 0.219008
## SaleConditionFamily    1.559e-03  2.790e-02   0.056 0.955453
## SaleConditionNormal    6.190e-02  1.312e-02   4.716 2.67e-06 ***
## SaleConditionPartial  -5.966e-02  6.681e-02  -0.893 0.372070
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1046 on 1246 degrees of freedom
## Multiple R-squared:  0.9415, Adjusted R-squared:  0.9314
## F-statistic: 94.07 on 213 and 1246 DF,  p-value: < 2.2e-16
```

`summary(H_Lo$Exterior2nd)`

```
## AsbShng AsphShn Brk Cmn BrkFace CBlock CmentBd HdBoard ImStucc MetalSd
##      38      4      22      47      3      126      406      15      447
## Other Plywood Stone Stucco VinylSd Wd Sdng Wd Shng
##      1      270      6      47      1015      391      81
```

Remove Outlier and check

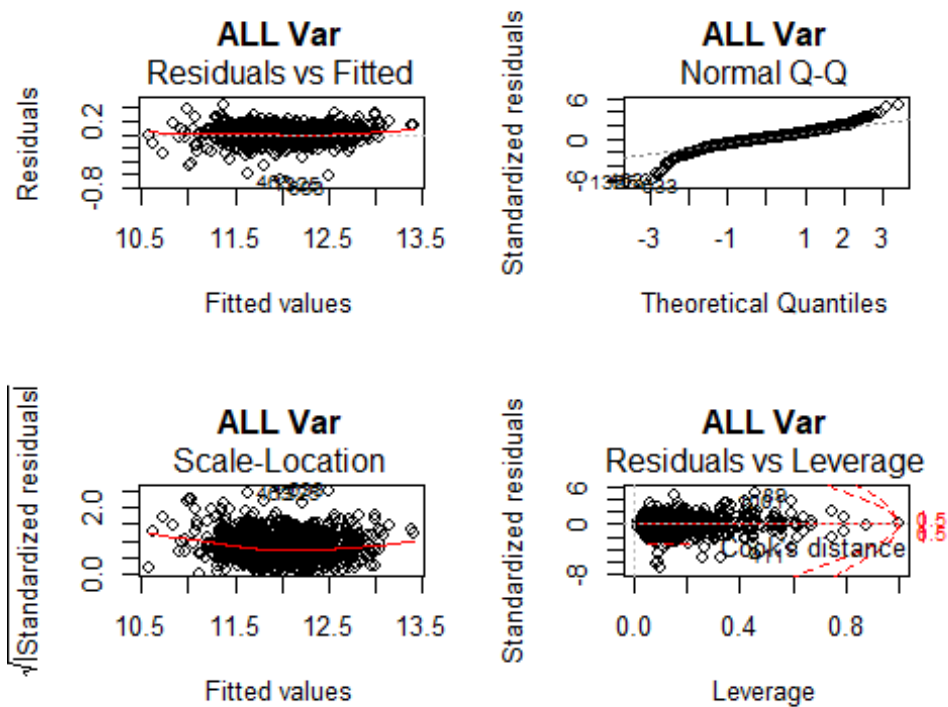
```
a_o<-lm(SalePrice~.,data=H_Lo[TR_RmvOut,-1]) # Removing ID
b_o<-lm(StepF_Log$call,data=H_Lo[TR_RmvOut,-1])
c_o<-lm(StepS_Log$call,data=H_Lo[TR_RmvOut,-1])
par(mfrow=c(2,2))
plot(a_o, main='ALL Var')

## Warning: not plotting observations with leverage one:
## 121, 347, 583, 595, 825, 1003, 1011, 1187, 1230, 1270, 1275, 1320, 1369,
## 1385

## Warning: not plotting observations with leverage one:
## 121, 347, 583, 595, 825, 1003, 1011, 1187, 1230, 1270, 1275, 1320, 1369,
## 1385

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced
```



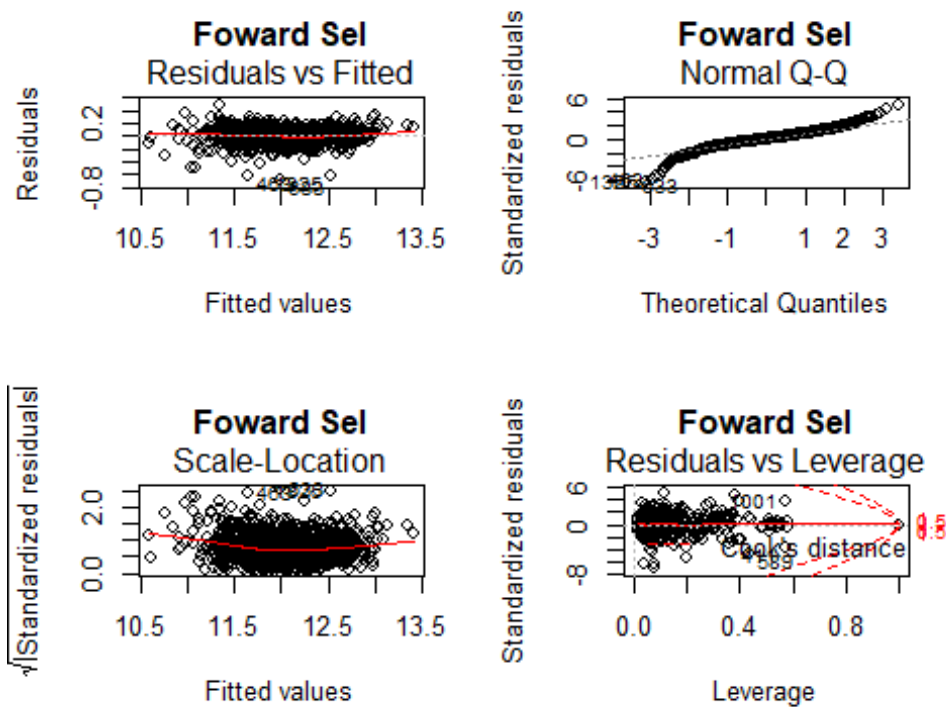
```
plot(b_o, main='Foward Sel')
```

```
## Warning: not plotting observations with leverage one:
## 272, 583, 825, 1003, 1011, 1187, 1230, 1275, 1320, 1369

## Warning: not plotting observations with leverage one:
## 272, 583, 825, 1003, 1011, 1187, 1230, 1275, 1320, 1369

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced
```

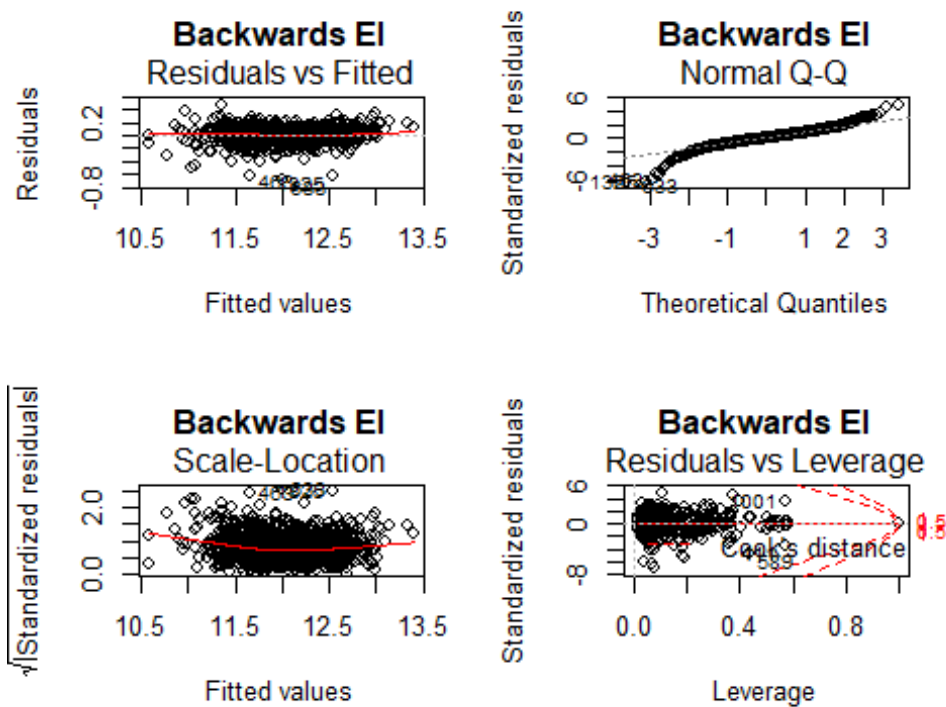
```
plot(c_o, main='Backwards E1')
```

```
## Warning: not plotting observations with leverage one:
## 121, 272, 825, 1003, 1011, 1187, 1230, 1275, 1320, 1369

## Warning: not plotting observations with leverage one:
## 121, 272, 825, 1003, 1011, 1187, 1230, 1275, 1320, 1369

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced

## Warning in sqrt(crit * p * (1 - hh)/hh): NaNs produced
```



```
plot(a_o,which=c(4))
plot(b_o,which=c(4))
plot(c_o,which=c(4))

#pdtrain<-predict(a,newdata=H_Lo[!is.na(H_Lo$SalePrice),])
pd_Log_LM_All_o<-predict(a_o,newdata=H_Lo[is.na(H_Lo$SalePrice),])

## Warning in predict.lm(a_o, newdata = H_Lo[is.na(H_Lo$SalePrice), ]):
## prediction from a rank-deficient fit may be misleading

pd_Log_LM_f2_o<-predict(b_o,newdata=H_Lo[is.na(H_Lo$SalePrice),])
pd_Log_LM_f3_o<-predict(c_o,newdata=H_Lo[is.na(H_Lo$SalePrice),])
#trainSP<-H_Lo$SalePrice[!is.na(H_Lo$SalePrice)]

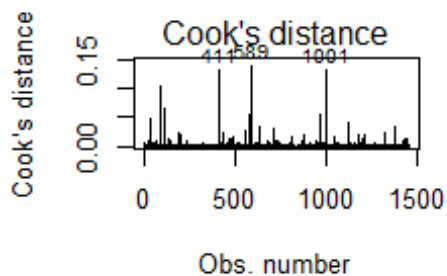
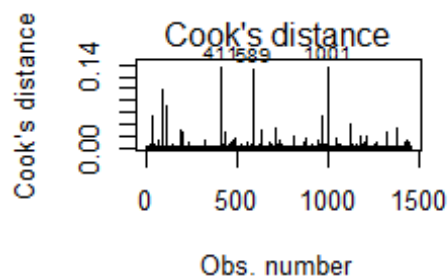
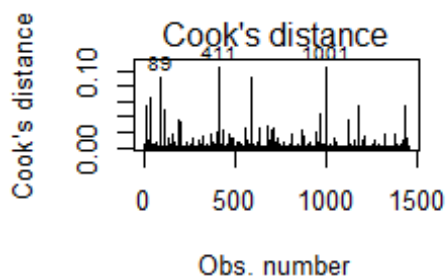
#plot(exp(trainSP),exp(pdtrain), col='blue', ylab='Prediction')
#points(actual,exp(pdtest), col='red')
#abline(b=1,a=0)
#RMSE(exp(trainSP),exp(pdtrain))
#RMSE(actual,exp(pdtest))

print("RMSLE for All Var - Log")

## [1] "RMSLE for All Var - Log"

RMSE(pd_Log_LM_All_o,log(actual))
```

```
## [1] 0.1340808
RMSE(pd_Log_LM_f2_o,log(actual))
## [1] 0.1311971
RMSE(pd_Log_LM_f3_o,log(actual))
## [1] 0.1305666
```



Lasso

Applied - Only done on all variables for Lasso/Ridge/Elasticnet

```
tc<-trainControl(method="cv", number=10)
#rainset -DummyVariables
H_DummyLo_Train<- as.data.frame(model.matrix(~.-1,data=H_Lo[1:1460,-1])) # Remove ID

#TestSet -DummyVariables
H_DummyLo_Test<- as.data.frame(model.matrix(~.-1,data=H_Lo[1461:2919,-c(1,78)
])) # RemoveID, salePrice
#as.data.frame

dim(H_DummyLo_Train)
## [1] 1460 216

set.seed(12334)

myLasso1Log <-train(SalePrice~.,
```

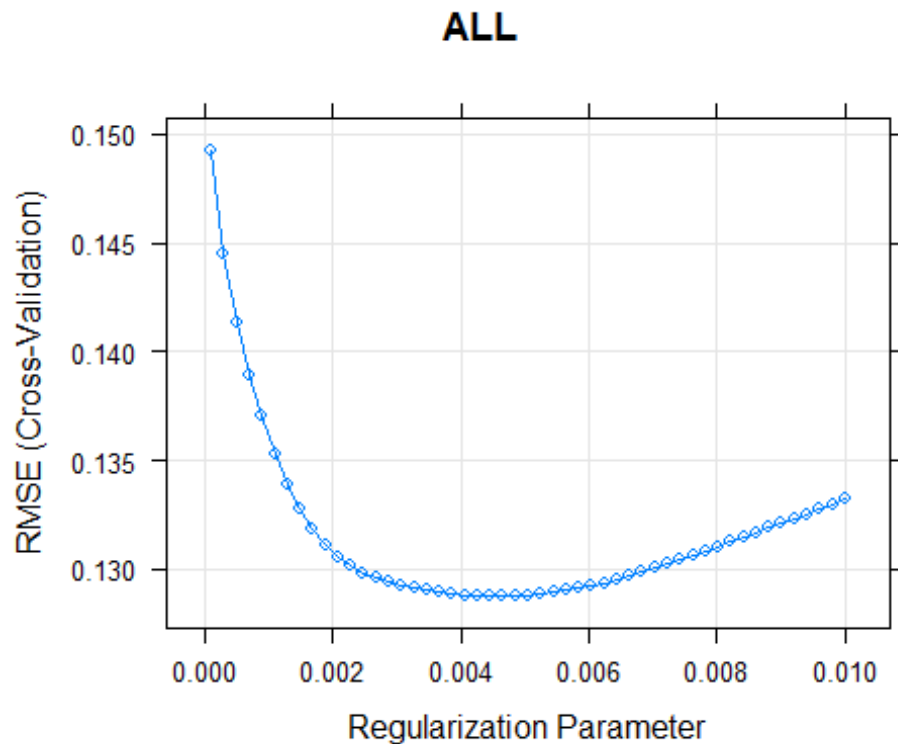
```

        data = H_DummyLo_Train
        ,method='glmnet',
        tuneGrid=expand.grid(alpha=1,lambda=seq(0.0001,0.01,length=5
1)), trControl=tc)
myLasso2Log <-train(SalePrice~., data = H_DummyLo_Train[-c(524,1299),]
        ,method='glmnet',tuneGrid=expand.grid(alpha=1,lambda=seq(0.0001,0.01,length
=51)), trControl=tc)
#myLasso3 <-train(f3, data = H_Dummy_Train
# ,method='glmnet',tuneGrid=expand.grid(alpha=1,lambda=seq(500,1500,length=5
0)), trControl=tc)
#myLasso4 <-train(f4, data = H_Dummy_Train
# ,method='glmnet',tuneGrid=expand.grid(alpha=1,lambda=seq(500,1500,length=5
0)), trControl=tc)

set.seed(12334)

plot(myLasso1Log, main='ALL')

```

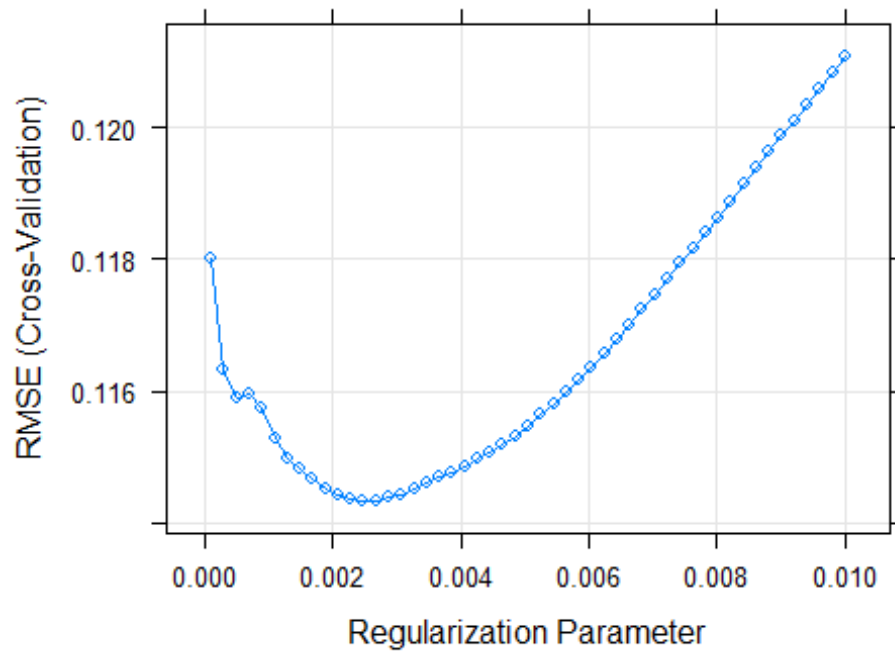


```

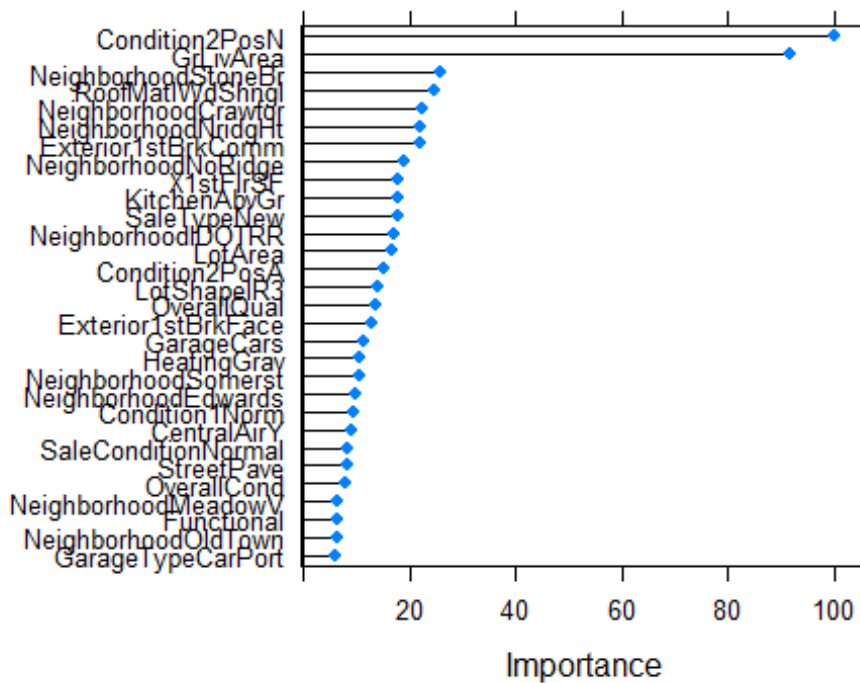
plot(myLasso2Log, main='Outlier Removed')

```

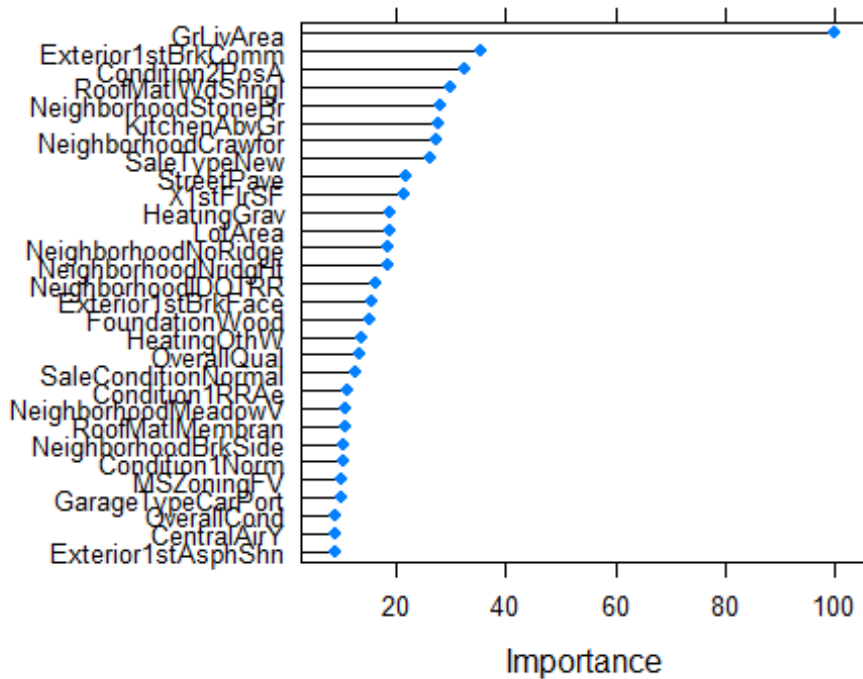
Outlier Removed



```
plot(varImp(myLasso1Log), Scale=F, top=30)
```



```
plot(varImp(myLasso2Log), Scale=F ,top=30)
```



```
varImp(myLasso1Log)
```

```
## glmnet variable importance
##
##   only 20 most important variables shown (out of 215)
##
##               Overall
## Condition2PosN    100.00
## GrLivArea         91.43
## NeighborhoodStoneBr 25.92
## RoofMatlWdShngl   24.90
## NeighborhoodCrawfor 22.51
## NeighborhoodNridgHt 21.91
## Exterior1stBrkComm 21.88
## NeighborhoodNoRidge 18.85
## X1stFlrSF         17.91
## KitchenAbvGr      17.90
## SaleTypeNew       17.76
## NeighborhoodIDOTRR 17.18
## LotArea           16.82
## Condition2PosA    15.04
## LotShapeIR3       14.22
## OverallQual       13.63
## Exterior1stBrkFace 12.97
## GarageCars        11.36
## HeatingGrav       10.76
## NeighborhoodSomerst 10.68
```

```

varImp(myLasso2Log)

## glmnet variable importance
##
##   only 20 most important variables shown (out of 215)
##
##               Overall
## GrLivArea      100.00
## Exterior1stBrkComm  35.56
## Condition2PosA    32.57
## RoofMatlWdShngl   29.77
## NeighborhoodStoneBr 28.16
## KitchenAbvGr      27.61
## NeighborhoodCrawfor 27.29
## SaleTypeNew       26.05
## StreetPave        21.90
## X1stFlrSF         21.38
## HeatingGrav       18.95
## LotArea          18.90
## NeighborhoodNoRidge 18.59
## NeighborhoodNridgHt 18.43
## NeighborhoodIDOTRR 16.19
## Exterior1stBrkFace 15.46
## FoundationWood     15.10
## HeatingOthW        13.54
## OverallQual        13.37
## SaleConditionNormal 12.58

myLasso1Log$bestTune

##   alpha  lambda
## 23      1 0.004456

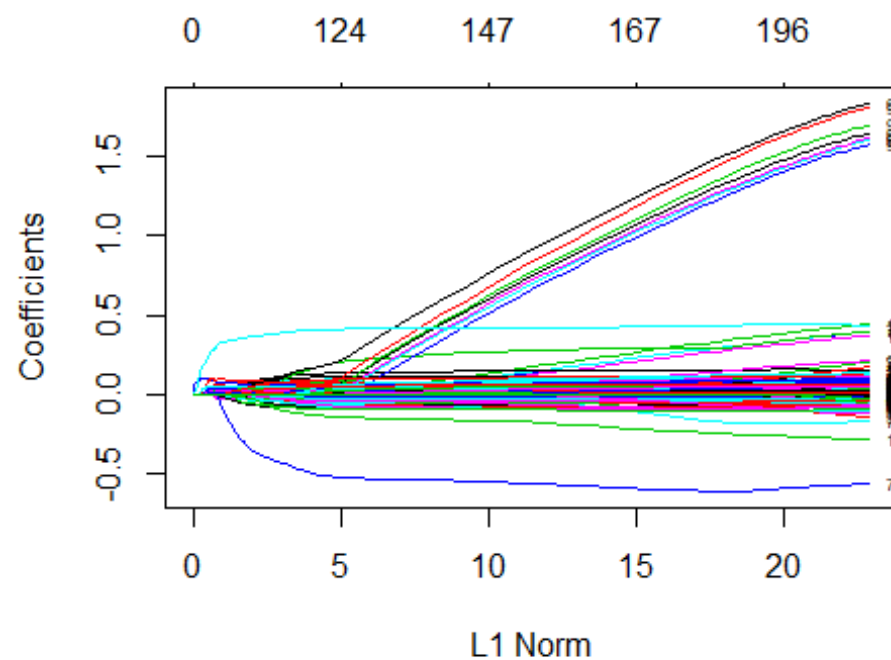
myLasso2Log$bestTune

##   alpha  lambda
## 13      1 0.002476

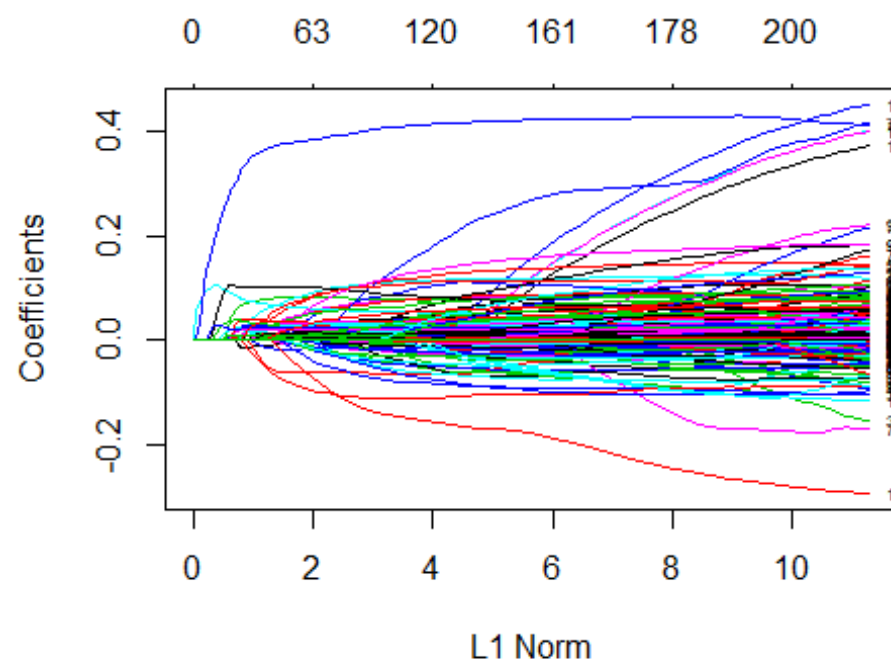
#myLasso3$bestTune
#myLasso4$bestTune

plot(myLasso1Log$finalModel, label=T)

```



```
plot(myLasso2Log$finalModel ,label=T)
```



Predict and Measure RMSE of Log

```
PdmyLasso1Log<-predict(myLasso1Log, newdata=H_DummyLo_Test)
PdmyLasso2Log<-predict(myLasso2Log, newdata=H_DummyLo_Test)
```

```
RMSE(PdmyLasso1Log, log(actual))
```

```
## [1] 0.1269526
```

```
RMSE(PdmyLasso2Log, log(actual))
```

```
## [1] 0.1242114
```

Ridge of Log

```
set.seed(12334)
```

```
myRidge1Log <-train(SalePrice~.,
                    data = H_DummyLo_Train
                    ,method='glmnet',
                    tuneGrid=expand.grid(alpha=0,lambda=seq(0.00001,0.08,length=
51)), trControl=tc)
```

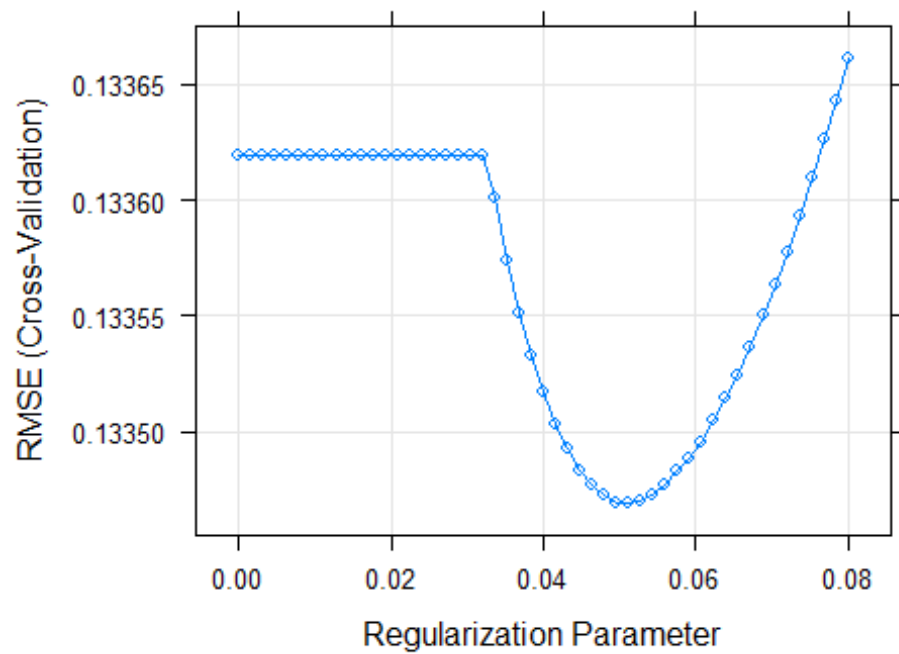
```
set.seed(12334)
```

```
myRidge2Log <-train(SalePrice~., data = H_DummyLo_Train[-c(524,1299),]
                    ,method='glmnet',tuneGrid=expand.grid(alpha=0,lambda=seq(0.0001,0.08,length
=51)), trControl=tc) # without outlier
```

```
par(mfrow=c(1,2))
```

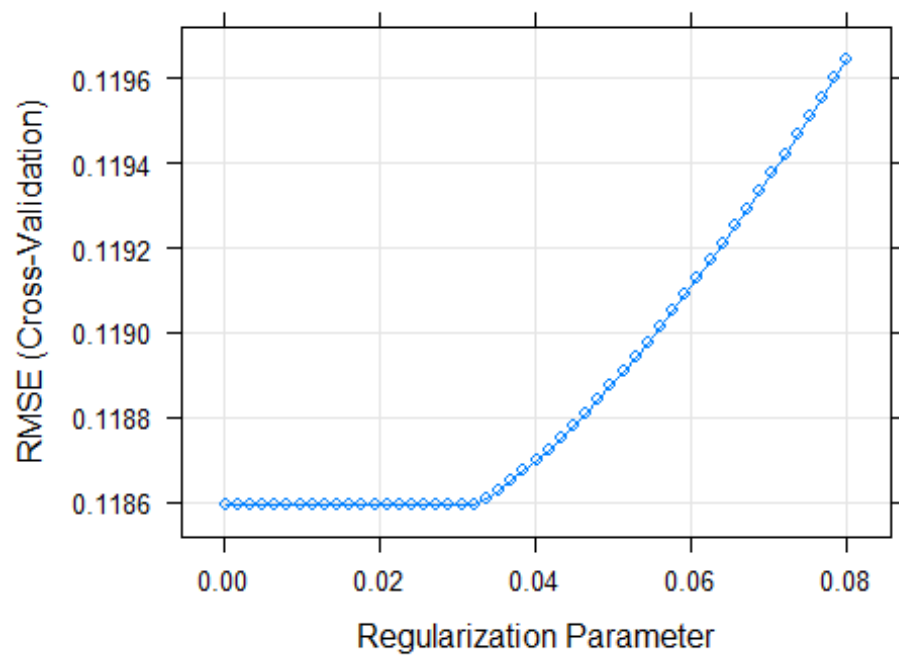
```
plot(myRidge1Log, main='ALL')
```

ALL



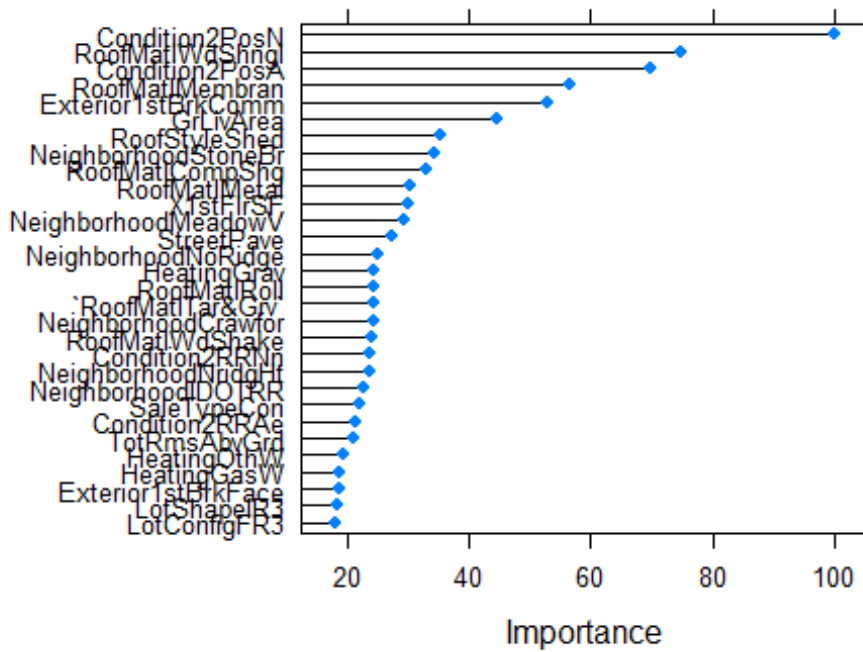
```
plot(myRidge2Log, main='Outlier Removed')
```

Outlier Removed



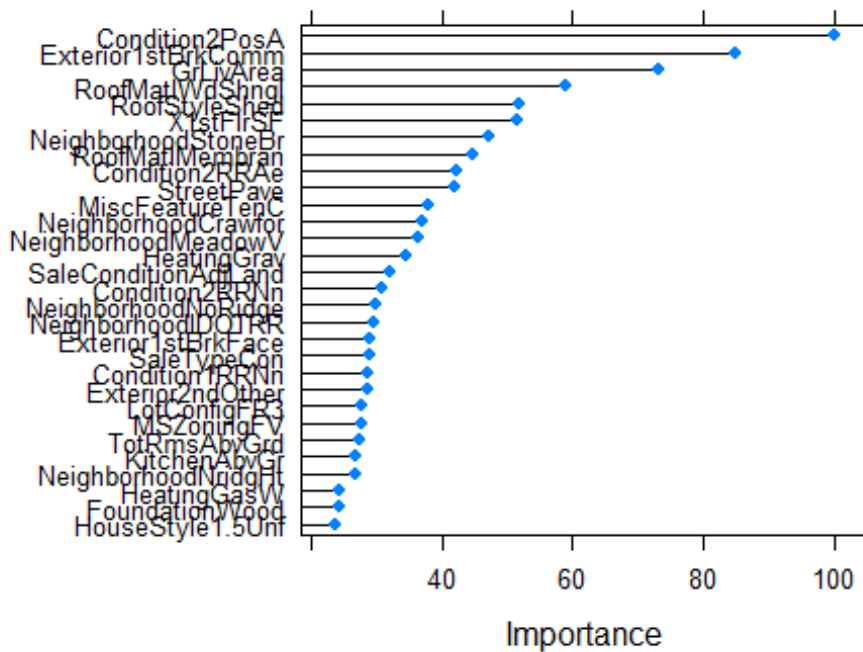
```
plot(varImp(myRidge1Log), Scale=F, top=30, main='ALL')
```

ALL



```
plot(varImp(myRidge2Log), Scale=F ,top=30, main='Outlier Removed')
```

Outlier Removed



```
varImp(myRidge1Log,main='ALL')
```

```
## glmnet variable importance
##
##   only 20 most important variables shown (out of 215)
##
##               Overall
## Condition2PosN    100.00
## RoofMatlWdShngl   74.63
## Condition2PosA    69.64
## RoofMatlMembran   56.52
## Exterior1stBrkComm 52.94
## GrLivArea         44.45
## RoofStyleShed     35.25
## NeighborhoodStoneBr 34.26
## RoofMatlCompShg   33.06
## RoofMatlMetal     30.21
## X1stFlrSF         30.10
## NeighborhoodMeadowV 29.23
## StreetPave        27.52
## NeighborhoodNoRidge 25.13
## HeatingGrav       24.53
## RoofMatlRoll      24.53
## `RoofMatlTar&Grv` 24.28
## NeighborhoodCrawfor 24.27
## RoofMatlWdShake   23.93
## Condition2RRNn    23.83
```

```
varImp(myRidge2Log,main='Outlier Removed')
```

```
## glmnet variable importance
##
##   only 20 most important variables shown (out of 215)
##
##               Overall
## Condition2PosA    100.00
## Exterior1stBrkComm 84.96
## GrLivArea         73.26
## RoofMatlWdShngl   58.77
## RoofStyleShed     51.75
## X1stFlrSF         51.63
## NeighborhoodStoneBr 47.27
## RoofMatlMembran   44.81
## Condition2RRAE    42.14
## StreetPave        42.07
## MiscFeatureTenC   38.04
## NeighborhoodCrawfor 37.03
## NeighborhoodMeadowV 36.24
## HeatingGrav       34.49
## SaleConditionAdjLand 31.94
## Condition2RRNn    30.73
## NeighborhoodNoRidge 29.89
```

```
## NeighborhoodIDOTRR      29.52
## Exterior1stBrkFace      29.05
## SaleTypeCon              28.96

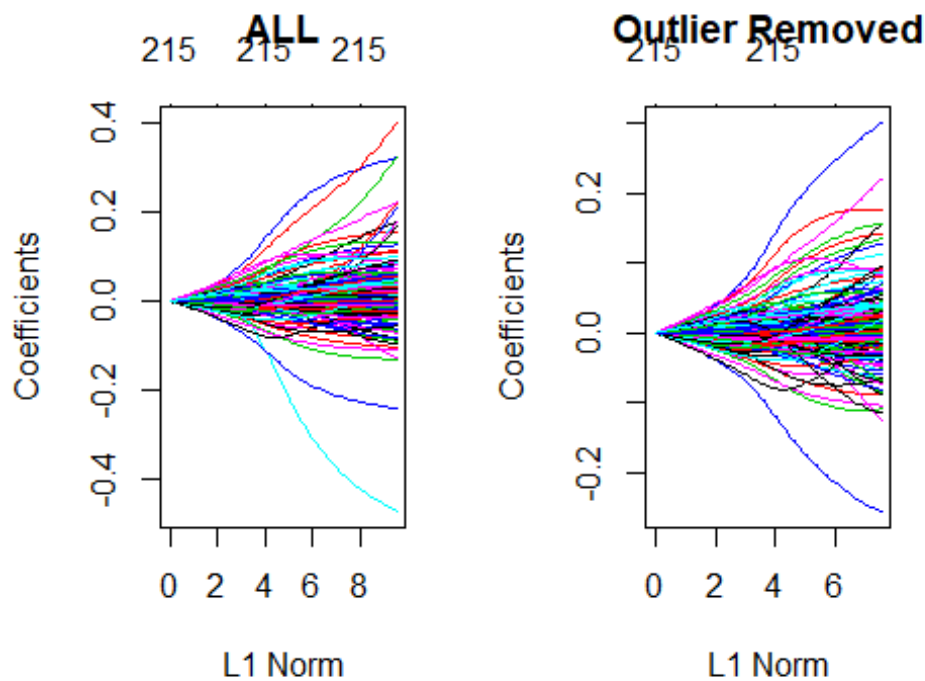
myRidge1Log$bestTune

##      alpha      lambda
## 33      0 0.0512036

myRidge2Log$bestTune

##      alpha      lambda
## 21      0 0.03206

plot(myRidge1Log$finalModel,main='ALL')
plot(myRidge2Log$finalModel,main='Outlier Removed')
```



```
PdmyRidge1Log<-predict(myRidge1Log, newdata=H_DummyLo_Test)
PdmyRidge2Log<-predict(myRidge2Log, newdata=H_DummyLo_Test)

RMSE(PdmyRidge1Log,log(actual))

## [1] 0.1317548

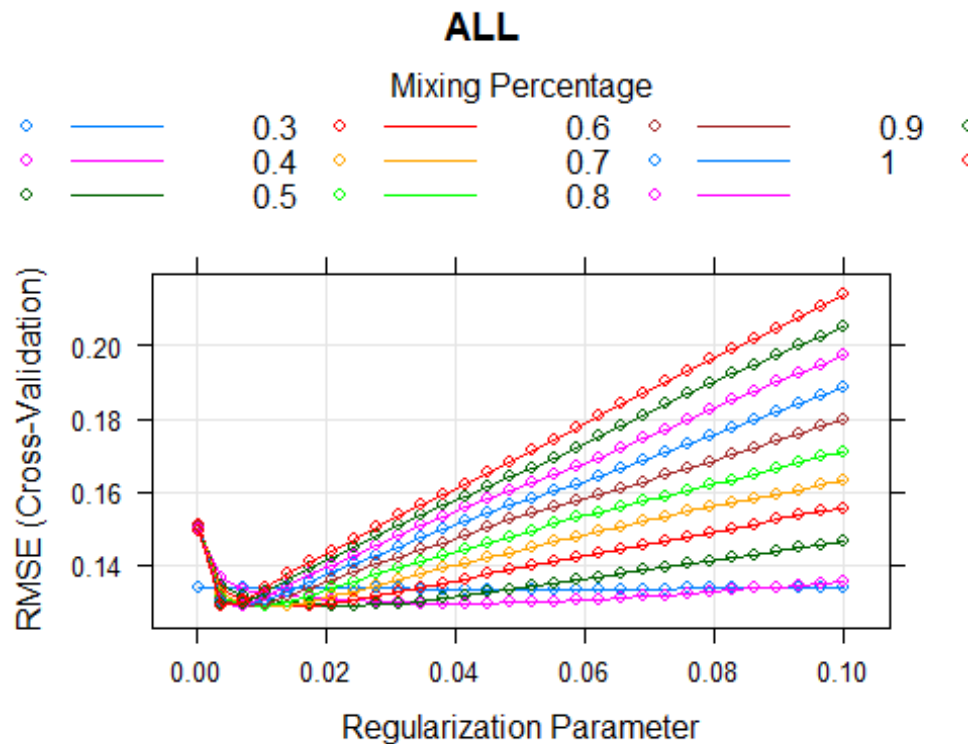
RMSE(PdmyRidge2Log,log(actual))

## [1] 0.1287369
```

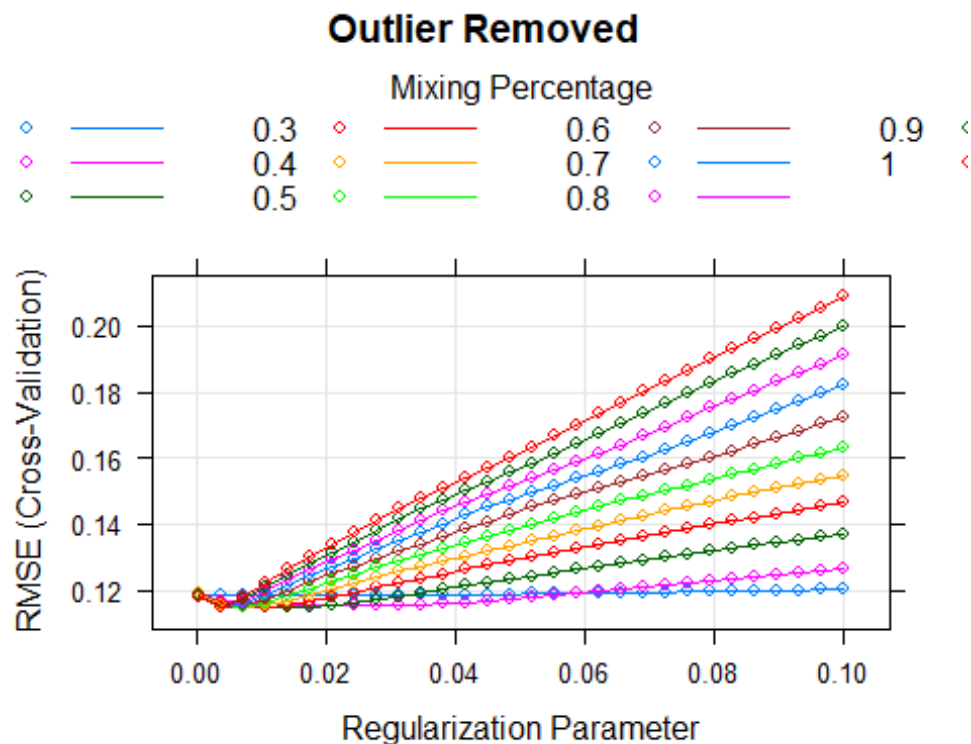
Elasticnet of Log

```
set.seed(12334)
myEla1Log<-train(SalePrice~.,data = H_DummyLo_Train,method='glmnet',tuneGrid=
expand.grid(alpha=seq(0, 1,length=11),lambda=seq(0.0001,0.1,length=30)), trCo
ntrol=tc)
set.seed(12334)
myEla2Log<-train(SalePrice~.,data = H_DummyLo_Train[-c(524,1299),],method='gl
mnet',tuneGrid=expand.grid(alpha=seq(0, 1,length=11),lambda=seq(0.0001,0.1,le
ngth=30)), trControl=tc)

par(mfrow=c(1,2))
plot(myEla1Log, main='ALL')
```



```
plot(myEla2Log, main='Outlier Removed')
```



```
#plot(varImp(myEla1Log), top=25)
#varImp(myEla1Log)
myEla1Log$bestTune

##      alpha      lambda
## 213    0.7 0.006989655

myEla2Log$bestTune

##      alpha      lambda
## 242    0.8 0.003544828

PdEla1Log<-predict(myEla1Log, newdata=H_DummyLo_Test)
PdEla2Log<-predict(myEla2Log, newdata=H_DummyLo_Test)
RMSE(PdEla1Log, log(actual))

## [1] 0.1273111

RMSE(PdEla2Log, log(actual))

## [1] 0.1243352
```

Original Lasso/Ridge/Elastic - removing outlier didn't help and actually gave worse results (barely) but once log transformed, it improved the results although little.

Random Forest for Log

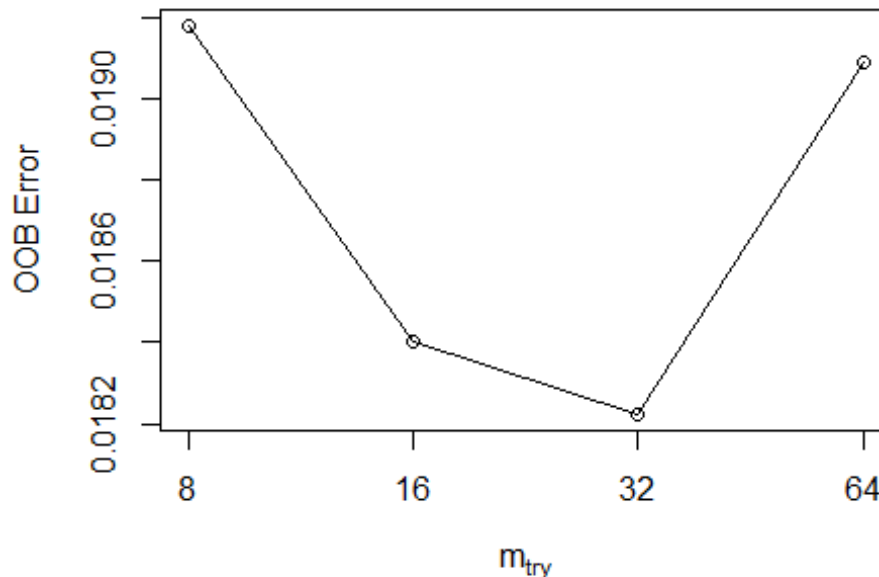
Do a quick tuning.

```

set.seed(100)
tuneRF10<-tuneRF(H_Lo[1:1460,-c(1,78)],H_Lo[1:1460,78], mtryStart=16,StepFactor=1 ,improve=0.0001, Trace=T,Plot=T,ntreeTry=500, doBest = T)

## mtry = 16  OOB error = 0.01840168
## Searching left ...
## mtry = 8      OOB error = 0.01917974
## -0.04228188 1e-04
## Searching right ...
## mtry = 32     OOB error = 0.01822353
## 0.009681035 1e-04
## mtry = 64     OOB error = 0.01908944
## -0.04751592 1e-04

```



Best ntree=32. Apply rf1=randomForest(SalePrice~.,data=Train, ntree=500,mtry=32)

```

RFLog=randomForest(SalePrice~.,data=H_Lo[TR,-1], mtry = 32,ntree=500)
RFLog1<-randomForest(SalePrice~.,data=H_Lo[TR_RmvOut,-1], mtry = 32,ntree=500)
PdRFLog<-predict(RFLog,newdata= H_Lo[1461:2919,-78])
PdRFLog1<-predict(RFLog1,newdata= H_Lo[1461:2919,-78])
RMSE(PdRFLog,log(actual))

## [1] 0.1397059

RMSE(PdRFLog1,log(actual))

## [1] 0.1407842

```


SVM Tuning - some are texted out due to running time

```
#SV1logtune1<-tune.svm(SalePrice~.,data=H_Lo[TR,-1],gamma=0.1*c(0.0001, 0.0005,0.001,0.005, 0.01, 0.15,0.2, 0.3,0.5) ,cost=1:10)

#print(SV1logtune1)
#plot(SV1logtune1)
#sqrt(SV1logtune1$best.performance)
```

Parameter tuning of 'svm':

- sampling method: 10-fold cross validation
- best parameters: gamma cost 0.001 4
- best performance: 0.01571751

Need more fining tuning up to 0.02 - 0.001 4'

```
#SV1logtune2<-tune.svm(SalePrice~.,data=H_Lo[TR,-1],gamma=0.002*(1:10) ,cost=1:10)

#print(SV1logtune2)
#plot(SV1logtune2)
#sqrt(SV1logtune2$best.performance)
```

Finally try for 0.0005 to 0.005 and cost between 1 and 7 - not done due to running time.

```
#SV1logtune3<-tune.svm(SalePrice~.,data=H_Lo[TR,-1],gamma=0.0005*(1:10) ,cost=1:7)

#print(SV1logtune3)
#plot(SV1logtune3)
```

Parameter tuning of 'svm':

- sampling method: 10-fold cross validation
- best parameters: gamma cost 0.0035 2
- best performance: 0.01617483

[1] 0.1271803

```
SVLog1<-svm(SalePrice~.,data=H_Lo[TR,-1], gamma= 0.0035, cost=2)
SVLog2<-svm(SalePrice~.,data=H_Lo[TR_RmvOut,-1], gamma= 0.0035, cost=2 )
```

Plot all to check.

```
pdSVLog1<-predict(SVLog1, newdata=H_Lo[1461:2919,-78])
pdSVLog2<-predict(SVLog2, newdata=H_Lo[1461:2919,-78])
RMSE(pdSVLog1,log(actual))
```

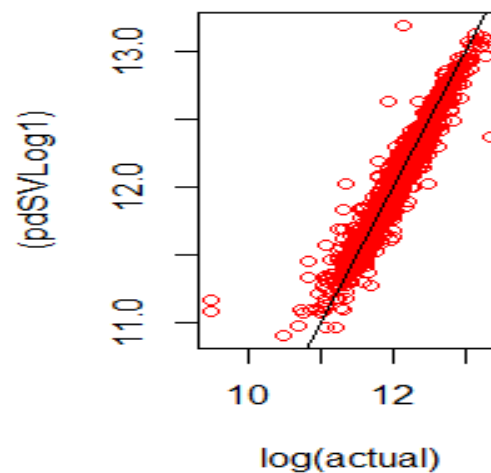
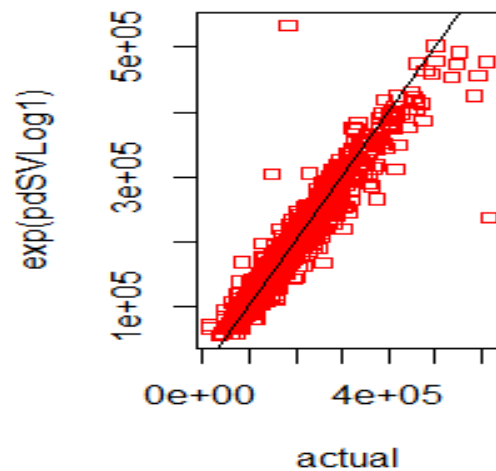
```
## [1] 0.1263945

RMSE(pdSVLog2, log(actual))

## [1] 0.1254436

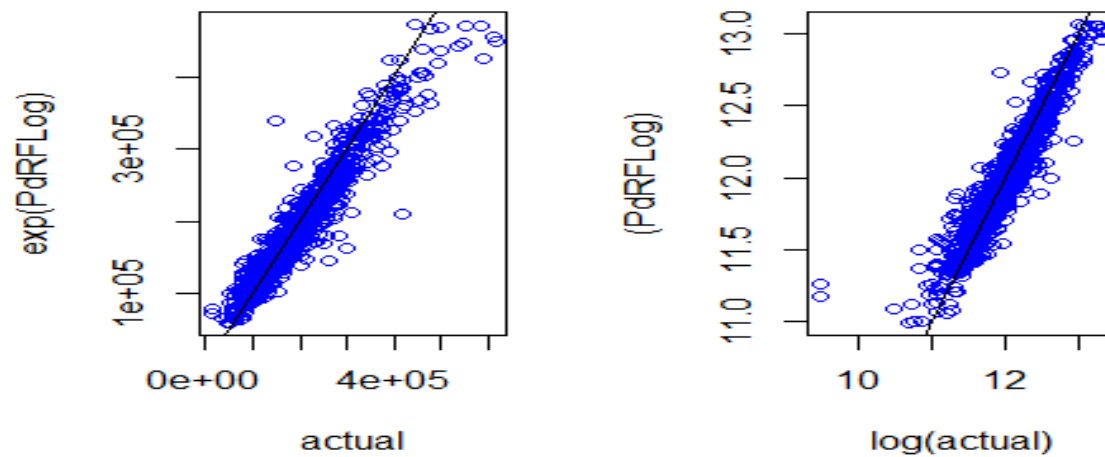
par(mfrow=c(1,2))
plot(actual, exp(pdSVLog1), pch=22,col='red')
abline(a=0,b=1)

plot(log(actual), (pdSVLog1), col='red')
abline(a=0,b=1)
```



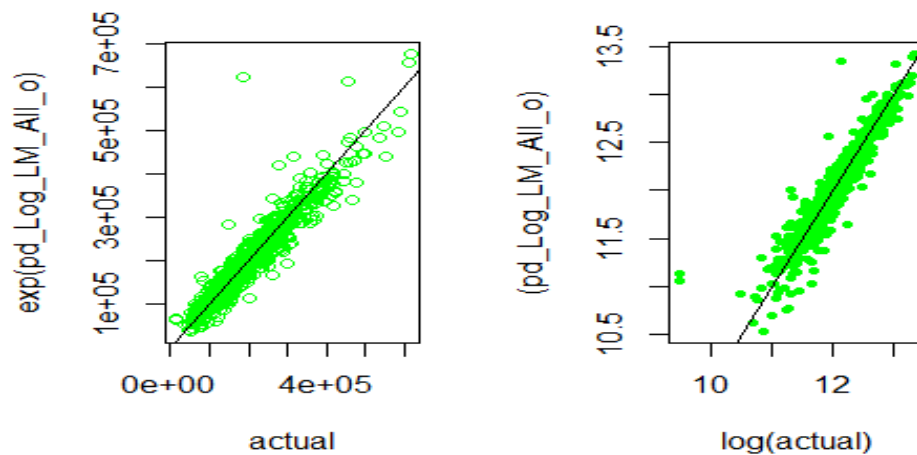
SVM - Logged Plot

```
par(mfrow=c(1,2))
plot(actual, exp(PdRFLog), col='blue')
abline(a=0,b=1)
plot(log(actual), (PdRFLog), col='blue')
abline(a=0,b=1)
```



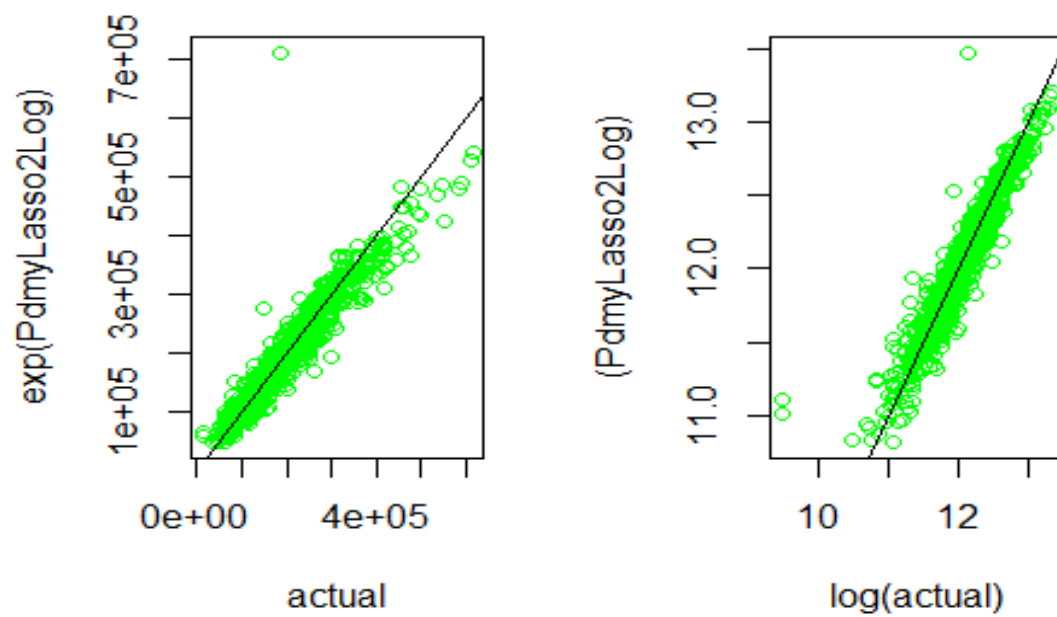
Random Forest – actual/Pred

```
par(mfrow=c(1,2))
plot(actual, exp(pd_Log_LM_All_o), col='green')
abline(a=0,b=1)
plot(log(actual), (pd_Log_LM_All_o), pch=20,col='green')
abline(a=0,b=1)
```



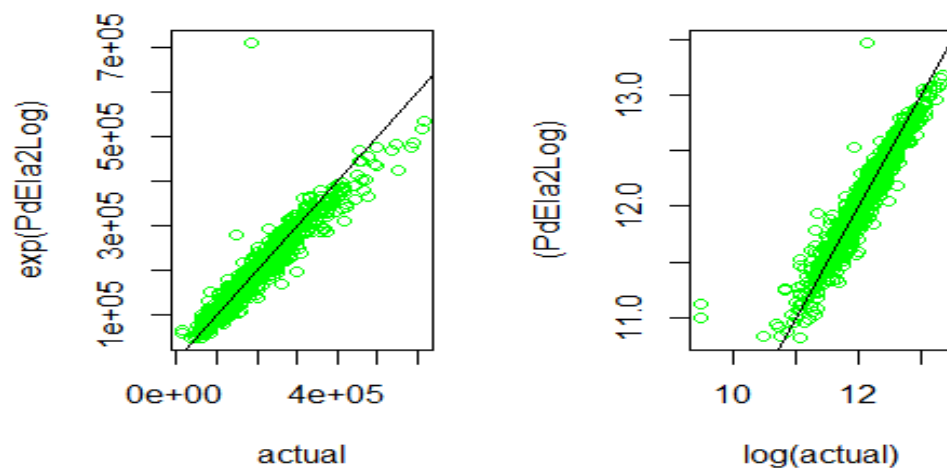
Linear Regression without outliers.

```
par(mfrow=c(1,2))
plot(actual, exp(PdmyLasso2Log), col='green')
abline(a=0,b=1)
plot(log(actual), (PdmyLasso2Log), col='green')
abline(a=0,b=1)
```



Lasso without outliers.

```
par(mfrow=c(1,2))
plot(actual, exp(PdEla2Log), col='green')
abline(a=0,b=1)
plot(log(actual), (PdEla2Log), col='green')
abline(a=0,b=1)
```



Elasticnet w/o outliers.