

Predicting House Sale Price

```
rm(list = ls())
library(car)

## Loading required package: carData

library(dplyr)

##
## Attaching package: 'dplyr'
## The following object is masked from 'package:car':
##
##      recode
## The following objects are masked from 'package:stats':
##
##      filter, lag
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union

library(ggplot2)
library(zoo)

##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric

library(forcats)

setwd("C:\\Users\\naimesh.chaudhari\\Desktop\\H6 Applied DM")

dftrain <- read.csv("train.csv")
dftest <- read.csv("test.csv")
#there are select cols where a N/A ment none, so we are replacing NA/s with none for thise
validcols <- c("Alley","BsmtQual","BsmtCond", "BsmtExposure", "BsmtFinType1"
              , "BsmtFinType2", "FireplaceQu", "GarageType", "GarageFinish", "GarageQual"
              , "GarageCond", "PoolQC", "Fence", "MiscFeature")

#data is evenly spread
nrow(dftrain)

## [1] 1460

ncol(dftrain)

## [1] 81

nrow(dftrain)

## [1] 1459
```

```
ncol(dftrain)
```

```
## [1] 80
```

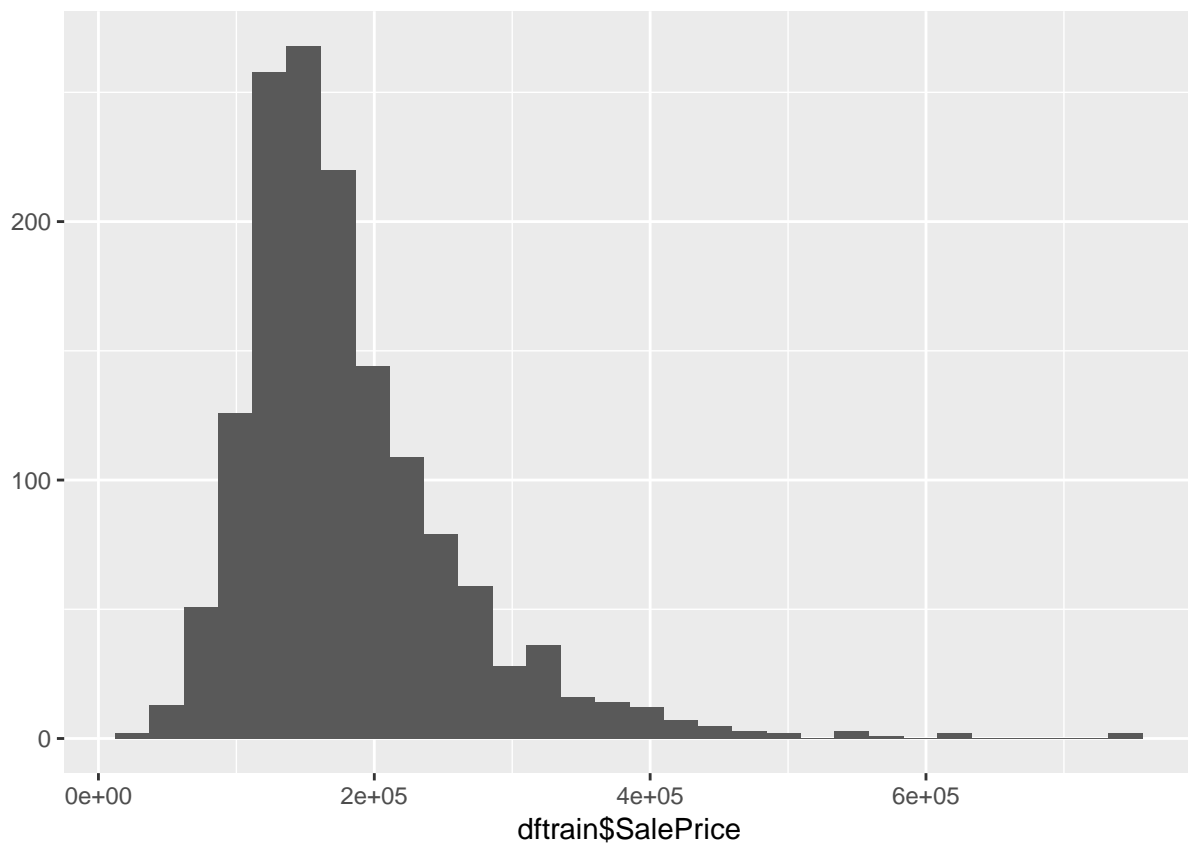
```
summary(dftrain$SalePrice)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  34900  129975  163000  180921  214000  755000
```

```
#skewed data
```

```
qplot(dftrain$SalePrice)
```

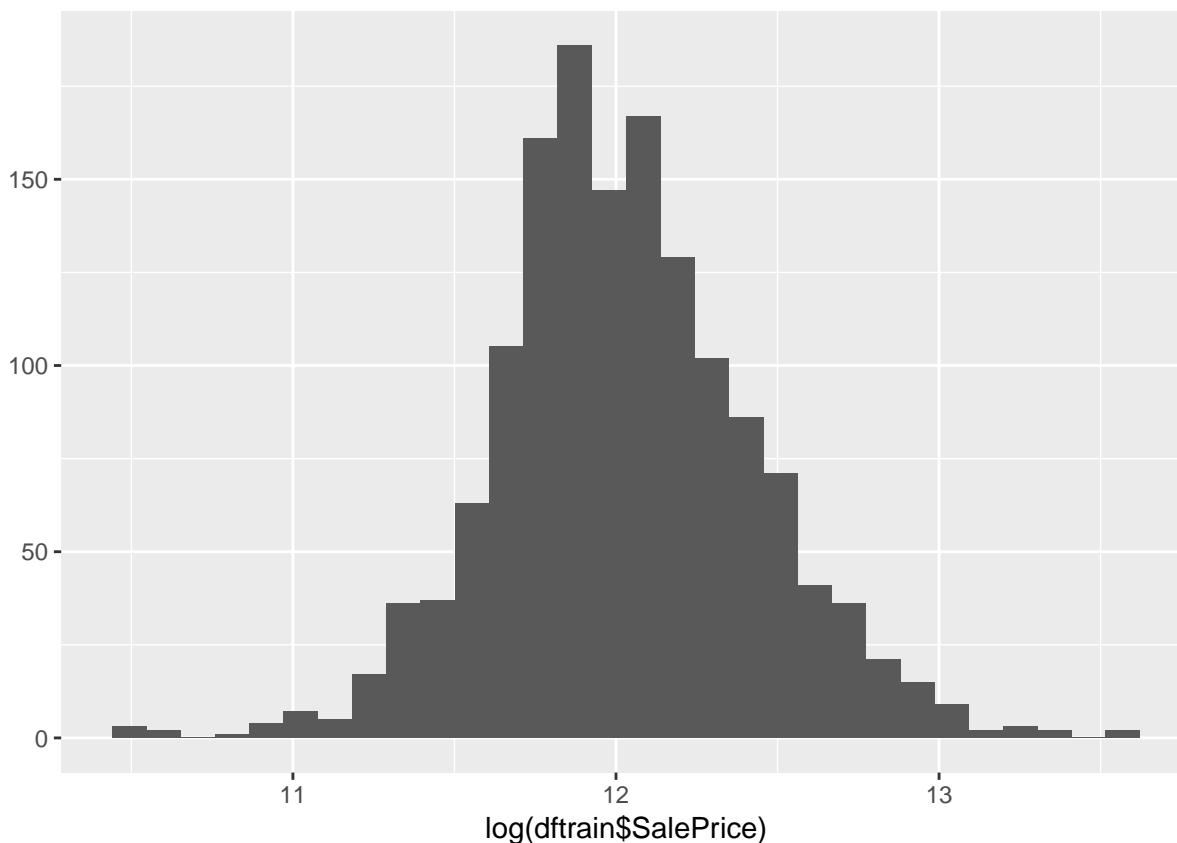
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
#Using normalized data will be better
```

```
qplot(log(dftrain$SalePrice))
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
#everything is numeric or factro
str(dftrain)
```

```
## 'data.frame':   1460 obs. of  81 variables:
## $ Id           : int  1 2 3 4 5 6 7 8 9 10 ...
## $ MSSubClass    : int  60 20 60 70 60 50 20 60 50 190 ...
## $ MSZoning      : Factor w/ 5 levels "C (all)","FV",...: 4 4 4 4 4 4 4 4 5 4 ...
## $ LotFrontage   : int  65 80 68 60 84 85 75 NA 51 50 ...
## $ LotArea       : int  8450 9600 11250 9550 14260 14115 10084 10382 6120 7420 ...
## $ Street        : Factor w/ 2 levels "Grvl","Pave": 2 2 2 2 2 2 2 2 2 2 ...
## $ Alley         : Factor w/ 2 levels "Grvl","Pave": NA NA NA NA NA NA NA NA NA NA ...
## $ LotShape      : Factor w/ 4 levels "IR1","IR2","IR3",...: 4 4 1 1 1 1 4 1 4 4 ...
## $ LandContour   : Factor w/ 4 levels "Bnk","HLS","Low",...: 4 4 4 4 4 4 4 4 4 4 ...
## $ Utilities     : Factor w/ 2 levels "AllPub","NoSeWa": 1 1 1 1 1 1 1 1 1 1 ...
## $ LotConfig     : Factor w/ 5 levels "Corner","CulDSac",...: 5 3 5 1 3 5 5 1 5 1 ...
## $ LandSlope     : Factor w/ 3 levels "Gtl","Mod","Sev": 1 1 1 1 1 1 1 1 1 1 ...
## $ Neighborhood : Factor w/ 25 levels "Blmngtn","Blueste",...: 6 25 6 7 14 12 21 17 18 4 ...
## $ Condition1    : Factor w/ 9 levels "Artery","Feedr",...: 3 2 3 3 3 3 3 5 1 1 ...
## $ Condition2    : Factor w/ 8 levels "Artery","Feedr",...: 3 3 3 3 3 3 3 3 1 ...
## $ BldgType      : Factor w/ 5 levels "1fam","2fmCon",...: 1 1 1 1 1 1 1 1 1 2 ...
## $ HouseStyle    : Factor w/ 8 levels "1.5Fin","1.5Unf",...: 6 3 6 6 6 1 3 6 1 2 ...
## $ OverallQual   : int  7 6 7 7 8 5 8 7 7 5 ...
## $ OverallCond   : int  5 8 5 5 5 5 5 6 5 6 ...
## $ YearBuilt     : int  2003 1976 2001 1915 2000 1993 2004 1973 1931 1939 ...
## $ YearRemodAdd  : int  2003 1976 2002 1970 2000 1995 2005 1973 1950 1950 ...
## $ RoofStyle     : Factor w/ 6 levels "Flat","Gable",...: 2 2 2 2 2 2 2 2 2 2 ...
## $ RoofMatl      : Factor w/ 8 levels "ClyTile","CompShg",...: 2 2 2 2 2 2 2 2 2 2 ...
```

```

## $ Exterior1st : Factor w/ 15 levels "AsbShng","AsphShn",...: 13 9 13 14 13 13 13 7 4 9 ...
## $ Exterior2nd : Factor w/ 16 levels "AsbShng","AsphShn",...: 14 9 14 16 14 14 14 7 16 9 ...
## $ MasVnrType : Factor w/ 4 levels "BrkCmn","BrkFace",...: 2 3 2 3 2 3 4 4 3 3 ...
## $ MasVnrArea : int 196 0 162 0 350 0 186 240 0 0 ...
## $ ExterQual : Factor w/ 4 levels "Ex","Fa","Gd",...: 3 4 3 4 3 4 3 4 4 4 ...
## $ ExterCond : Factor w/ 5 levels "Ex","Fa","Gd",...: 5 5 5 5 5 5 5 5 5 5 ...
## $ Foundation : Factor w/ 6 levels "BrkTil","CBlock",...: 3 2 3 1 3 6 3 2 1 1 ...
## $ BsmtQual : Factor w/ 4 levels "Ex","Fa","Gd",...: 3 3 3 4 3 3 1 3 4 4 ...
## $ BsmtCond : Factor w/ 4 levels "Fa","Gd","Po",...: 4 4 4 2 4 4 4 4 4 4 ...
## $ BsmtExposure : Factor w/ 4 levels "Av","Gd","Mn",...: 4 2 3 4 1 4 1 3 4 4 ...
## $ BsmtFinType1 : Factor w/ 6 levels "ALQ","BLQ","GLQ",...: 3 1 3 1 3 3 3 1 6 3 ...
## $ BsmtFinSF1 : int 706 978 486 216 655 732 1369 859 0 851 ...
## $ BsmtFinType2 : Factor w/ 6 levels "ALQ","BLQ","GLQ",...: 6 6 6 6 6 6 6 2 6 6 ...
## $ BsmtFinSF2 : int 0 0 0 0 0 0 0 32 0 0 ...
## $ BsmtUnfSF : int 150 284 434 540 490 64 317 216 952 140 ...
## $ TotalBsmtSF : int 856 1262 920 756 1145 796 1686 1107 952 991 ...
## $ Heating : Factor w/ 6 levels "Floor","GasA",...: 2 2 2 2 2 2 2 2 2 2 ...
## $ HeatingQC : Factor w/ 5 levels "Ex","Fa","Gd",...: 1 1 1 3 1 1 1 1 1 3 ...
## $ CentralAir : Factor w/ 2 levels "N","Y": 2 2 2 2 2 2 2 2 2 2 ...
## $ Electrical : Factor w/ 5 levels "FuseA","FuseF",...: 5 5 5 5 5 5 5 5 5 2 ...
## $ X1stFlrSF : int 856 1262 920 961 1145 796 1694 1107 1022 1077 ...
## $ X2ndFlrSF : int 854 0 866 756 1053 566 0 983 752 0 ...
## $ LowQualFinSF : int 0 0 0 0 0 0 0 0 0 0 ...
## $ GrLivArea : int 1710 1262 1786 1717 2198 1362 1694 2090 1774 1077 ...
## $ BsmtFullBath : int 1 0 1 1 1 1 1 1 0 1 ...
## $ BsmtHalfBath : int 0 1 0 0 0 0 0 0 0 0 ...
## $ FullBath : int 2 2 2 1 2 1 2 2 2 1 ...
## $ HalfBath : int 1 0 1 0 1 1 0 1 0 0 ...
## $ BedroomAbvGr : int 3 3 3 3 4 1 3 3 2 2 ...
## $ KitchenAbvGr : int 1 1 1 1 1 1 1 1 2 2 ...
## $ KitchenQual : Factor w/ 4 levels "Ex","Fa","Gd",...: 3 4 3 3 3 4 3 4 4 4 ...
## $ TotRmsAbvGrd : int 8 6 6 7 9 5 7 7 8 5 ...
## $ Functional : Factor w/ 7 levels "Maj1","Maj2",...: 7 7 7 7 7 7 7 3 7 ...
## $ Fireplaces : int 0 1 1 1 1 0 1 2 2 2 ...
## $ FireplaceQu : Factor w/ 5 levels "Ex","Fa","Gd",...: NA 5 5 3 5 NA 3 5 5 5 ...
## $ GarageType : Factor w/ 6 levels "2Types","Attchd",...: 2 2 2 6 2 2 2 2 6 2 ...
## $ GarageYrBlt : int 2003 1976 2001 1998 2000 1993 2004 1973 1931 1939 ...
## $ GarageFinish : Factor w/ 3 levels "Fin","RFn","Unf": 2 2 2 3 2 3 2 2 3 2 ...
## $ GarageCars : int 2 2 2 3 3 2 2 2 2 1 ...
## $ GarageArea : int 548 460 608 642 836 480 636 484 468 205 ...
## $ GarageQual : Factor w/ 5 levels "Ex","Fa","Gd",...: 5 5 5 5 5 5 5 5 2 3 ...
## $ GarageCond : Factor w/ 5 levels "Ex","Fa","Gd",...: 5 5 5 5 5 5 5 5 5 5 ...
## $ PavedDrive : Factor w/ 3 levels "N","P","Y": 3 3 3 3 3 3 3 3 3 3 ...
## $ WoodDeckSF : int 0 298 0 0 192 40 255 235 90 0 ...
## $ OpenPorchSF : int 61 0 42 35 84 30 57 204 0 4 ...
## $ EnclosedPorch : int 0 0 0 272 0 0 0 228 205 0 ...
## $ X3SsnPorch : int 0 0 0 0 0 320 0 0 0 0 ...
## $ ScreenPorch : int 0 0 0 0 0 0 0 0 0 0 ...
## $ PoolArea : int 0 0 0 0 0 0 0 0 0 0 ...
## $ PoolQC : Factor w/ 3 levels "Ex","Fa","Gd": NA NA NA NA NA NA NA NA NA NA ...
## $ Fence : Factor w/ 4 levels "GdPrv","GdWo",...: NA NA NA NA NA 3 NA NA NA NA ...
## $ MiscFeature : Factor w/ 4 levels "Gar2","Othr",...: NA NA NA NA NA 3 NA 3 NA NA ...
## $ MiscVal : int 0 0 0 0 0 700 0 350 0 0 ...
## $ MoSold : int 2 5 9 2 12 10 8 11 4 1 ...

```

```
## $ YrSold      : int  2008 2007 2008 2006 2008 2009 2007 2009 2008 2008 ...
## $ SaleType    : Factor w/ 9 levels "COD","Con","ConLD",...: 9 9 9 9 9 9 9 9 9 ...
## $ SaleCondition: Factor w/ 6 levels "Abnorml","AdjLand",...: 5 5 5 1 5 5 5 5 1 5 ...
## $ SalePrice   : int  208500 181500 223500 140000 250000 143000 307000 200000 129900 118000 ...
```

#let look at na's

```
nacols <- as.data.frame(as.table(colSums(is.na(dftrain))))
nacols <- filter(nacols, Freq > 0) %>% arrange(-Freq)
```

```
dftrain[,validcols] <-sapply(dftrain[,validcols],fct_explicit_na)
dftrain[,validcols] <- as.data.frame(sapply(dftrain[,validcols], factor))
dftrain$MasVnrType <- na.approx(dftrain$MasVnrType)
#replacing with most common value
dftrain$Electrical[is.na(dftrain$Electrical)] <- "SBrkr"
```

```
dftrain$LotFrontage <- na.approx(dftrain$LotFrontage)
dftrain$GarageYrBlt <- na.approx(dftrain$GarageYrBlt)
dftrain$MasVnrArea <- na.approx(dftrain$MasVnrArea)
```

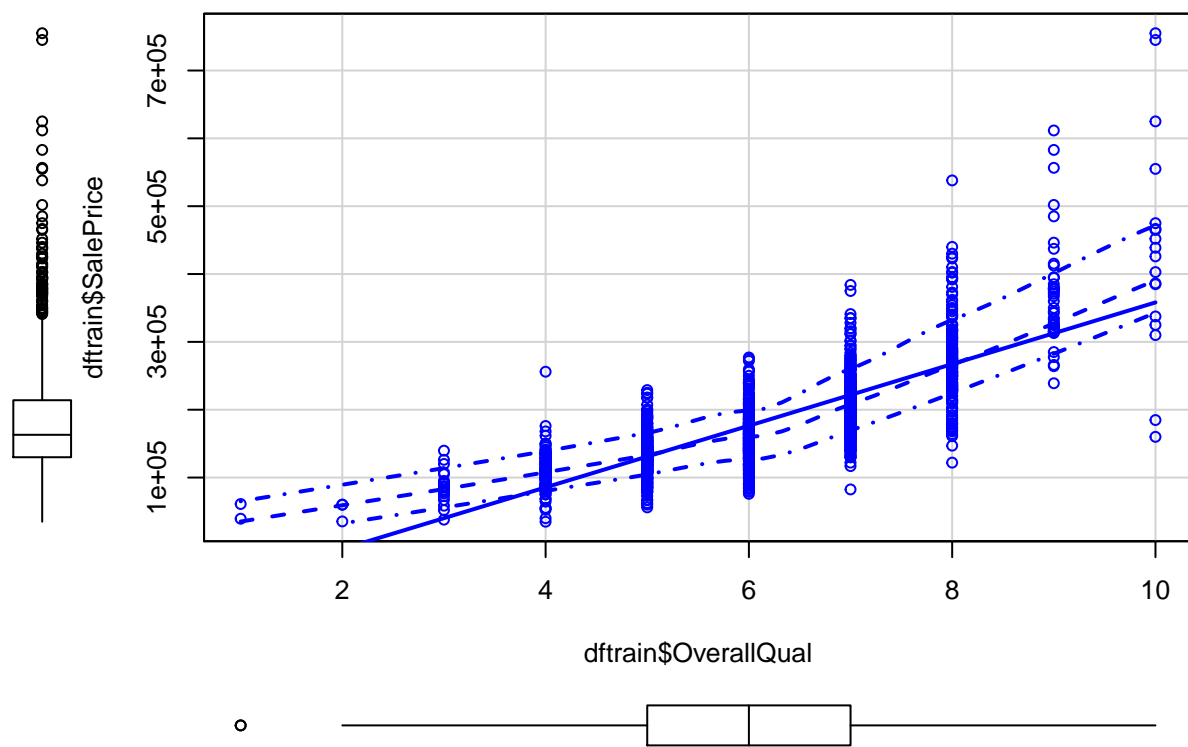
#all missing values are taken care off

#Lets Look @ Correlation

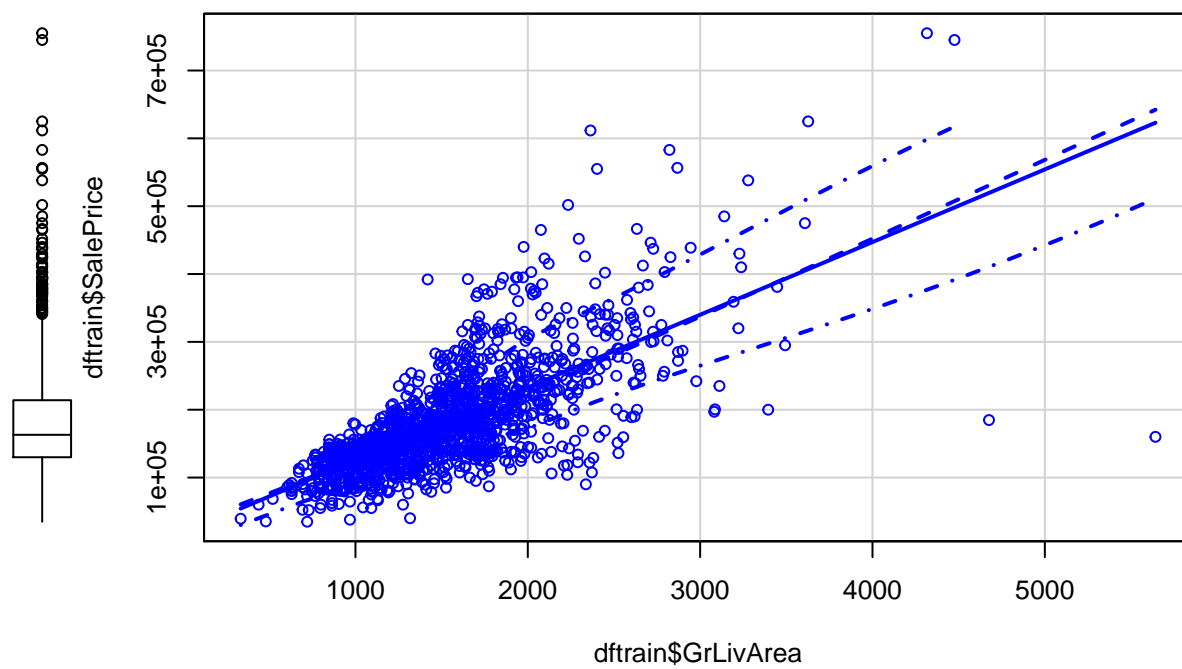
```
numeric <- dftrain[,sapply(dftrain,is.numeric)]
corr <- as.data.frame(as.table(cor(numeric$SalePrice, numeric))) %>% arrange(-Freq)
filter(corr,abs(corr$Freq) >= .5)
```

```
##      Var1      Var2      Freq
## 1      A      SalePrice 1.0000000
## 2      A      OverallQual 0.7909816
## 3      A      GrLivArea 0.7086245
## 4      A      GarageCars 0.6404092
## 5      A      GarageArea 0.6234314
## 6      A      TotalBsmtSF 0.6135806
## 7      A      X1stFlrSF 0.6058522
## 8      A      FullBath 0.5606638
## 9      A      TotRmsAbvGrd 0.5337232
## 10     A      YearBuilt 0.5228973
## 11     A      YearRemodAdd 0.5071010
```

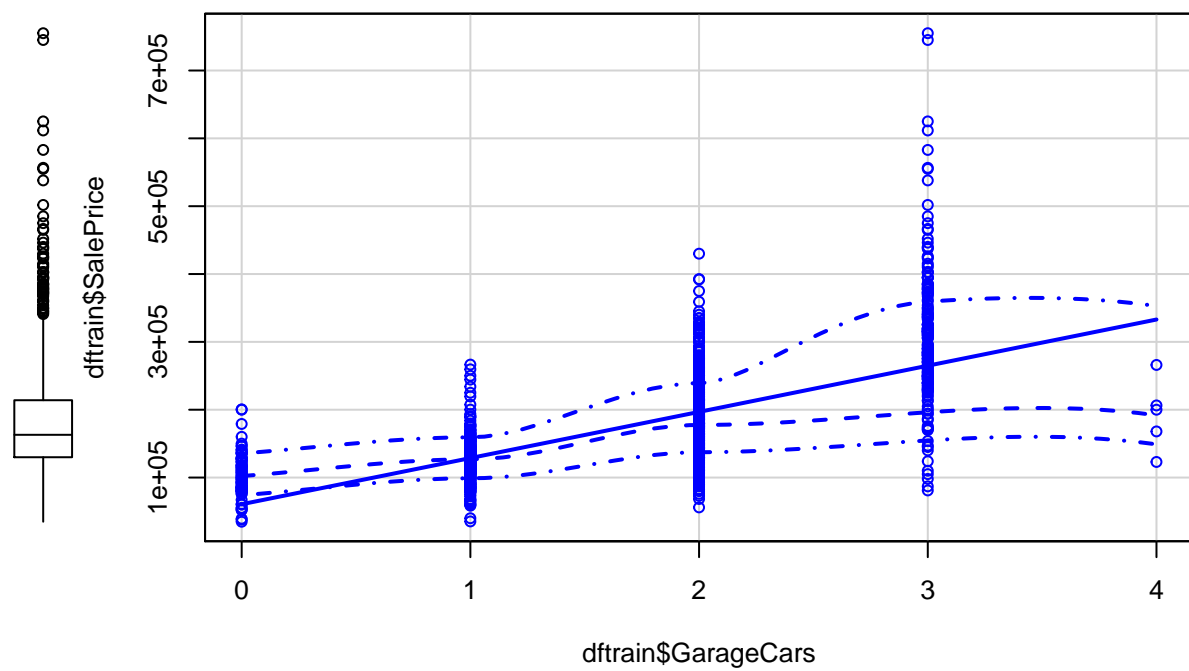
```
scatterplot(dftrain$SalePrice ~ dftrain$OverallQual)
```



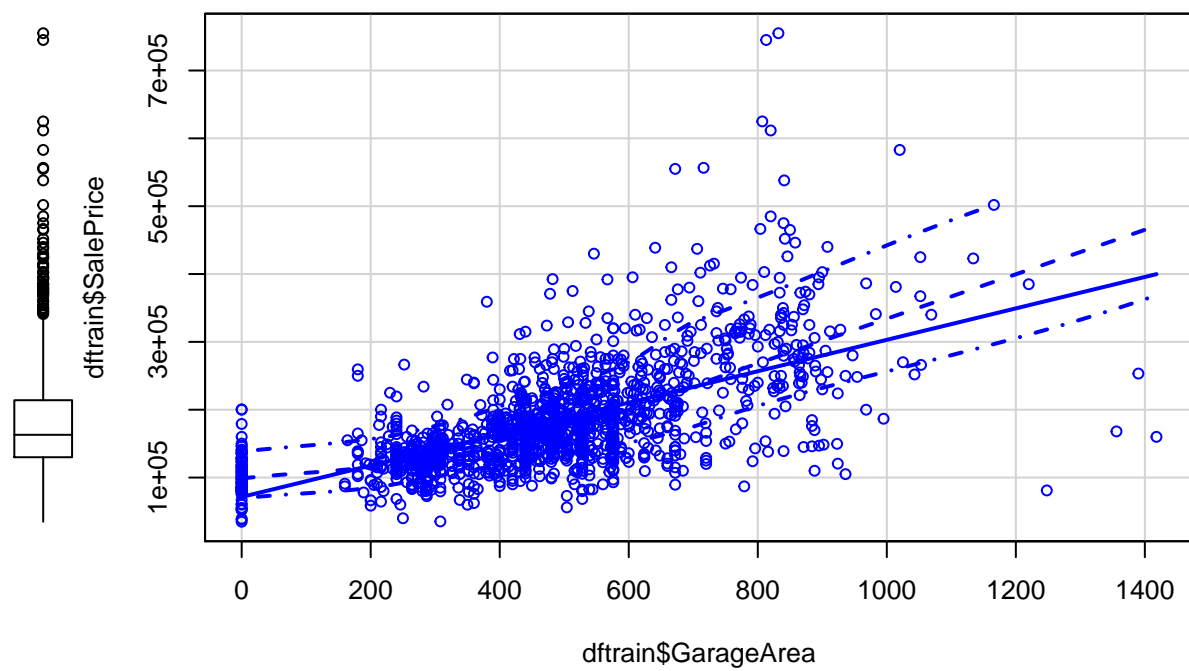
```
scatterplot(dftrain$SalePrice ~ dftrain$GrLivArea)
```



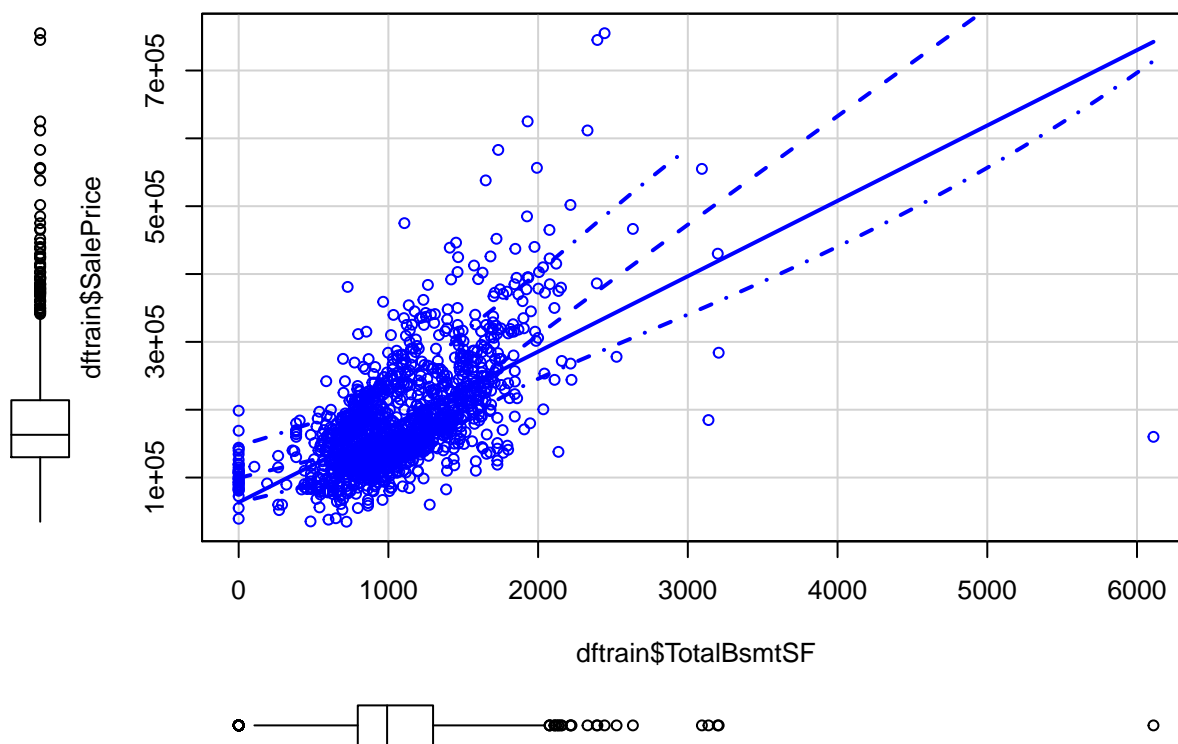
```
scatterplot(dftrain$SalePrice ~ dftrain$GrLivArea)
```



```
scatterplot(dftrain$SalePrice ~ dftrain$GarageArea)
```

```
scatterplot(dftrain$SalePrice ~ dftrain$TotalBsmtSF)
```



```
mymodel <- lm(SalePrice ~ ., data = dftrain)
summary(mymodel)
```

```
##
## Call:
## lm(formula = SalePrice ~ ., data = dftrain)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -177139  -9169        0     9508  177139
##
## Coefficients: (8 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -2.637e+05  1.045e+06  -0.252  0.800823
## Id           1.120e+00  1.546e+00   0.725  0.468648
## MSSubClass   -5.111e+01  8.243e+01  -0.620  0.535397
## MSZoningFV    3.266e+04  1.190e+04   2.745  0.006150 **
## MSZoningRH    2.215e+04  1.187e+04   1.866  0.062219 .
## MSZoningRL    2.492e+04  1.020e+04   2.444  0.014672 *
## MSZoningRM    2.157e+04  9.565e+03   2.255  0.024296 *
## LotFrontage   1.145e+01  3.962e+01   0.289  0.772702
## LotArea       7.195e-01  1.087e-01   6.620  5.40e-11 ***
## StreetPave    3.319e+04  1.216e+04   2.731  0.006415 **
## AlleyGrv1     1.311e+03  4.206e+03   0.312  0.755303
## AlleyPave     6.587e+02  4.745e+03   0.139  0.889622
## LotShapeIR2   5.125e+03  4.198e+03   1.221  0.222331
```

## LotShapeIR3	6.066e+03	8.808e+03	0.689	0.491160	
## LotShapeReg	1.804e+03	1.596e+03	1.130	0.258521	
## LandContourHLS	7.569e+03	5.111e+03	1.481	0.138866	
## LandContourLow	-1.157e+04	6.375e+03	-1.815	0.069820	.
## LandContourLvl	5.423e+03	3.691e+03	1.469	0.142036	
## UtilitiesNoSeWa	-3.734e+04	2.629e+04	-1.420	0.155851	
## LotConfigCulDSac	7.855e+03	3.256e+03	2.413	0.015979	*
## LotConfigFR2	-7.804e+03	4.011e+03	-1.946	0.051893	.
## LotConfigFR3	-1.737e+04	1.252e+04	-1.387	0.165672	
## LotConfigInside	-1.510e+03	1.778e+03	-0.849	0.395974	
## LandSlopeMod	7.449e+03	3.970e+03	1.876	0.060859	.
## LandSlopeSev	-4.221e+04	1.138e+04	-3.708	0.000219	***
## NeighborhoodBlueste	7.306e+03	1.917e+04	0.381	0.703137	
## NeighborhoodBrDale	-2.371e+03	1.092e+04	-0.217	0.828157	
## NeighborhoodBrkSide	-5.787e+03	9.416e+03	-0.615	0.538955	
## NeighborhoodClearCr	-1.460e+04	9.146e+03	-1.596	0.110696	
## NeighborhoodCollgCr	-1.012e+04	7.214e+03	-1.403	0.160882	
## NeighborhoodCrawfor	1.178e+04	8.510e+03	1.384	0.166630	
## NeighborhoodEdwards	-2.148e+04	7.952e+03	-2.702	0.006993	**
## NeighborhoodGilbert	-1.147e+04	7.611e+03	-1.508	0.131905	
## NeighborhoodIDOTRR	-1.203e+04	1.068e+04	-1.126	0.260234	
## NeighborhoodMeadowV	-6.580e+03	1.108e+04	-0.594	0.552671	
## NeighborhoodMitchel	-2.095e+04	8.115e+03	-2.582	0.009953	**
## NeighborhoodNames	-1.734e+04	7.788e+03	-2.227	0.026129	*
## NeighborhoodNoRidge	2.572e+04	8.348e+03	3.081	0.002110	**
## NeighborhoodNPkVill	1.339e+04	1.394e+04	0.961	0.336901	
## NeighborhoodNridgHt	1.832e+04	7.469e+03	2.452	0.014329	*
## NeighborhoodNWAmes	-1.755e+04	7.982e+03	-2.199	0.028086	*
## NeighborhoodOldTown	-1.431e+04	9.589e+03	-1.492	0.135989	
## NeighborhoodSawyer	-1.116e+04	8.051e+03	-1.387	0.165846	
## NeighborhoodSawyerW	-2.786e+03	7.730e+03	-0.360	0.718623	
## NeighborhoodSomerst	-2.599e+03	8.953e+03	-0.290	0.771603	
## NeighborhoodStoneBr	3.952e+04	8.201e+03	4.819	1.63e-06	***
## NeighborhoodSWISU	-8.824e+03	9.663e+03	-0.913	0.361352	
## NeighborhoodTimber	-9.756e+03	8.073e+03	-1.209	0.227061	
## NeighborhoodVeenker	-1.888e+02	1.042e+04	-0.018	0.985555	
## Condition1Feedr	7.156e+03	5.015e+03	1.427	0.153810	
## Condition1Norm	1.633e+04	4.182e+03	3.904	9.99e-05	***
## Condition1PosA	1.022e+04	9.965e+03	1.026	0.305180	
## Condition1PosN	1.493e+04	7.425e+03	2.011	0.044505	*
## Condition1RRAE	-1.542e+04	9.047e+03	-1.705	0.088484	.
## Condition1RRAN	1.295e+04	6.937e+03	1.867	0.062083	.
## Condition1RRNE	-3.388e+03	1.744e+04	-0.194	0.845990	
## Condition1RRNN	1.168e+04	1.281e+04	0.912	0.362203	
## Condition2Feedr	-5.846e+03	2.335e+04	-0.250	0.802362	
## Condition2Norm	-1.058e+04	2.022e+04	-0.523	0.600952	
## Condition2PosA	4.318e+04	3.691e+04	1.170	0.242276	
## Condition2PosN	-2.382e+05	2.755e+04	-8.645	< 2e-16	***
## Condition2RRAE	-1.303e+05	6.482e+04	-2.010	0.044604	*
## Condition2RRAN	-2.397e+04	3.141e+04	-0.763	0.445463	
## Condition2RRNN	-2.724e+03	2.704e+04	-0.101	0.919775	
## BldgType2fmCon	-3.712e+03	1.243e+04	-0.299	0.765316	
## BldgTypeDuplex	-7.345e+03	7.372e+03	-0.996	0.319316	
## BldgTypeTwnhs	-1.993e+04	9.897e+03	-2.013	0.044291	*

## BldgTypeTwnhsE	-1.571e+04	8.990e+03	-1.747	0.080844	.
## HouseStyle1.5Unf	1.219e+04	7.940e+03	1.535	0.125012	
## HouseStyle1Story	5.461e+03	4.370e+03	1.250	0.211605	
## HouseStyle2.5Fin	-1.723e+04	1.232e+04	-1.398	0.162373	
## HouseStyle2.5Unf	-9.624e+03	9.210e+03	-1.045	0.296258	
## HouseStyle2Story	-5.987e+03	3.483e+03	-1.719	0.085839	.
## HouseStyleSFoyer	1.169e+03	6.247e+03	0.187	0.851635	
## HouseStyleSLvl	3.881e+03	5.529e+03	0.702	0.482870	
## OverallQual	6.813e+03	1.008e+03	6.757	2.19e-11	***
## OverallCond	5.792e+03	8.705e+02	6.654	4.32e-11	***
## YearBuilt	3.201e+02	7.644e+01	4.187	3.03e-05	***
## YearRemodAdd	1.055e+02	5.565e+01	1.895	0.058332	.
## RoofStyleGable	9.492e+03	1.843e+04	0.515	0.606585	
## RoofStyleGambrel	1.282e+04	2.016e+04	0.636	0.524898	
## RoofStyleHip	9.253e+03	1.850e+04	0.500	0.617022	
## RoofStyleMansard	2.041e+04	2.137e+04	0.955	0.339697	
## RoofStyleShed	1.001e+05	3.449e+04	2.901	0.003783	**
## RoofMatlCompShg	5.701e+05	5.251e+04	10.857	< 2e-16	***
## RoofMatlMembran	6.666e+05	6.243e+04	10.678	< 2e-16	***
## RoofMatlMetal	6.353e+05	6.206e+04	10.237	< 2e-16	***
## RoofMatlRoll	5.572e+05	5.814e+04	9.583	< 2e-16	***
## RoofMatlTar&Grv	5.722e+05	5.638e+04	10.149	< 2e-16	***
## RoofMatlWdShake	5.616e+05	5.478e+04	10.253	< 2e-16	***
## RoofMatlWdShngl	6.249e+05	5.344e+04	11.694	< 2e-16	***
## Exterior1stAsphShn	-2.623e+04	3.308e+04	-0.793	0.427906	
## Exterior1stBrkComm	-3.836e+03	2.773e+04	-0.138	0.889977	
## Exterior1stBrkFace	7.572e+03	1.274e+04	0.594	0.552447	
## Exterior1stCBlock	-1.479e+04	2.722e+04	-0.543	0.587035	
## Exterior1stCemntBd	-1.254e+04	1.897e+04	-0.661	0.508515	
## Exterior1stHdBoard	-1.332e+04	1.291e+04	-1.032	0.302418	
## Exterior1stImStucc	-2.322e+04	2.812e+04	-0.826	0.409113	
## Exterior1stMetalSd	-6.671e+03	1.457e+04	-0.458	0.647075	
## Exterior1stPlywood	-1.415e+04	1.274e+04	-1.111	0.266997	
## Exterior1stStone	-1.454e+03	2.426e+04	-0.060	0.952213	
## Exterior1stStucco	-7.620e+03	1.407e+04	-0.542	0.588144	
## Exterior1stVinylSd	-1.474e+04	1.330e+04	-1.108	0.267907	
## Exterior1stWd Sdng	-1.424e+04	1.236e+04	-1.152	0.249387	
## Exterior1stWdShing	-9.940e+03	1.334e+04	-0.745	0.456477	
## Exterior2ndAsphShn	1.231e+04	2.216e+04	0.556	0.578560	
## Exterior2ndBrk Cmn	5.394e+03	2.003e+04	0.269	0.787726	
## Exterior2ndBrkFace	4.075e+03	1.319e+04	0.309	0.757417	
## Exterior2ndCBlock	NA	NA	NA	NA	
## Exterior2ndCmentBd	1.282e+04	1.867e+04	0.686	0.492573	
## Exterior2ndHdBoard	8.344e+03	1.240e+04	0.673	0.501006	
## Exterior2ndImStucc	1.732e+04	1.432e+04	1.209	0.226755	
## Exterior2ndMetalSd	6.439e+03	1.418e+04	0.454	0.649870	
## Exterior2ndOther	-1.753e+04	2.703e+04	-0.648	0.516897	
## Exterior2ndPlywood	6.778e+03	1.203e+04	0.563	0.573419	
## Exterior2ndStone	-1.139e+04	1.711e+04	-0.665	0.505925	
## Exterior2ndStucco	5.917e+03	1.360e+04	0.435	0.663565	
## Exterior2ndVinylSd	1.362e+04	1.278e+04	1.066	0.286832	
## Exterior2ndWd Sdng	1.225e+04	1.192e+04	1.027	0.304501	
## Exterior2ndWd Shng	5.657e+03	1.244e+04	0.455	0.649322	
## MasVnrType	2.706e+03	1.279e+03	2.116	0.034574	*

## MasVnrArea	2.012e+01	5.033e+00	3.997	6.81e-05	***
## ExterQualFa	-7.030e+03	1.107e+04	-0.635	0.525509	
## ExterQualGd	-2.069e+04	4.762e+03	-4.345	1.51e-05	***
## ExterQualTA	-1.975e+04	5.293e+03	-3.731	0.000200	***
## ExterCondFa	-2.409e+03	1.804e+04	-0.134	0.893795	
## ExterCondGd	-7.025e+03	1.720e+04	-0.408	0.683079	
## ExterCondPo	8.649e+03	3.164e+04	0.273	0.784588	
## ExterCondTA	-4.017e+03	1.717e+04	-0.234	0.815049	
## FoundationCBlock	3.017e+03	3.165e+03	0.953	0.340617	
## FoundationPConc	4.044e+03	3.411e+03	1.186	0.236021	
## FoundationSlab	-7.249e+03	1.002e+04	-0.723	0.469696	
## FoundationStone	9.551e+03	1.138e+04	0.839	0.401591	
## FoundationWood	-2.673e+04	1.473e+04	-1.815	0.069826	.
## BsmtQualEx	-3.241e+04	3.616e+04	-0.896	0.370185	
## BsmtQualFa	-4.419e+04	3.595e+04	-1.229	0.219255	
## BsmtQualGd	-5.055e+04	3.593e+04	-1.407	0.159732	
## BsmtQualTA	-4.693e+04	3.583e+04	-1.310	0.190570	
## BsmtCondFa	-2.686e+03	4.232e+03	-0.635	0.525712	
## BsmtCondGd	-2.733e+03	3.244e+03	-0.842	0.399690	
## BsmtCondPo	6.626e+04	2.995e+04	2.212	0.027145	*
## BsmtCondTA	NA	NA	NA	NA	
## BsmtExposureAv	1.141e+04	2.293e+04	0.497	0.618977	
## BsmtExposureGd	2.567e+04	2.302e+04	1.115	0.265141	
## BsmtExposureMn	7.880e+03	2.300e+04	0.343	0.731962	
## BsmtExposureNo	6.265e+03	2.289e+04	0.274	0.784348	
## BsmtFinType1ALQ	-2.655e+03	2.911e+03	-0.912	0.361868	
## BsmtFinType1BLQ	1.864e+02	3.118e+03	0.060	0.952345	
## BsmtFinType1GLQ	2.939e+03	2.706e+03	1.086	0.277712	
## BsmtFinType1LwQ	-6.044e+03	3.763e+03	-1.606	0.108517	
## BsmtFinType1Rec	-2.585e+03	3.156e+03	-0.819	0.412896	
## BsmtFinType1Unf	NA	NA	NA	NA	
## BsmtFinSF1	3.853e+01	5.320e+00	7.243	7.80e-13	***
## BsmtFinType2ALQ	2.861e+04	2.493e+04	1.148	0.251347	
## BsmtFinType2BLQ	1.559e+04	2.466e+04	0.632	0.527303	
## BsmtFinType2GLQ	2.606e+04	2.545e+04	1.024	0.306014	
## BsmtFinType2LwQ	1.430e+04	2.469e+04	0.579	0.562637	
## BsmtFinType2Rec	1.825e+04	2.463e+04	0.741	0.458951	
## BsmtFinType2Unf	2.026e+04	2.457e+04	0.824	0.409903	
## BsmtFinSF2	3.169e+01	9.036e+00	3.507	0.000470	***
## BsmtUnfSF	2.106e+01	4.874e+00	4.321	1.68e-05	***
## TotalBsmtSF	NA	NA	NA	NA	
## HeatingGasA	9.939e+03	2.554e+04	0.389	0.697258	
## HeatingGasW	7.519e+03	2.635e+04	0.285	0.775371	
## HeatingGrav	1.686e+03	2.802e+04	0.060	0.952018	
## HeatingOthW	-1.100e+04	3.142e+04	-0.350	0.726287	
## HeatingWall	2.308e+04	2.968e+04	0.777	0.437017	
## HeatingQCFA	5.855e+02	4.701e+03	0.125	0.900914	
## HeatingQCGd	-3.900e+03	2.061e+03	-1.892	0.058726	.
## HeatingQCPo	2.043e+03	2.650e+04	0.077	0.938572	
## HeatingQCTA	-3.229e+03	2.063e+03	-1.565	0.117929	
## CentralAirY	-2.206e+02	3.863e+03	-0.057	0.954467	
## ElectricalFuseF	5.145e+01	5.738e+03	0.009	0.992848	
## ElectricalFuseP	-8.581e+03	1.859e+04	-0.462	0.644388	
## ElectricalMix	-4.451e+04	4.441e+04	-1.002	0.316499	

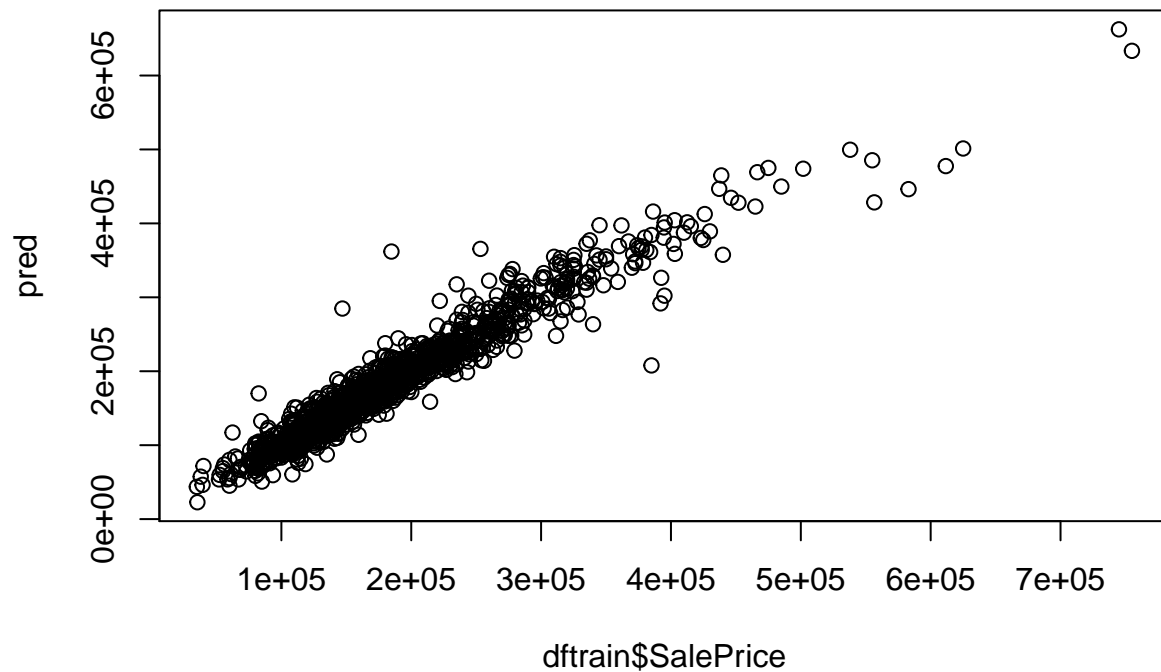
## ElectricalSBrkr	-2.127e+03	2.943e+03	-0.723	0.469984	
## X1stFlrSF	4.433e+01	5.635e+00	7.867	8.00e-15	***
## X2ndFlrSF	6.252e+01	5.657e+00	11.051	< 2e-16	***
## LowQualFinSF	-2.835e+00	1.902e+01	-0.149	0.881544	
## GrLivArea	NA	NA	NA	NA	
## BsmtFullBath	1.522e+03	1.974e+03	0.771	0.440857	
## BsmtHalfBath	-4.899e+02	3.016e+03	-0.162	0.870971	
## FullBath	3.651e+03	2.196e+03	1.663	0.096605	.
## HalfBath	1.871e+03	2.088e+03	0.896	0.370598	
## BedroomAbvGr	-3.684e+03	1.363e+03	-2.702	0.006984	**
## KitchenAbvGr	-1.356e+04	5.666e+03	-2.394	0.016834	*
## KitchenQualFa	-1.964e+04	6.189e+03	-3.172	0.001549	**
## KitchenQualGd	-2.352e+04	3.473e+03	-6.773	1.96e-11	***
## KitchenQualTA	-2.255e+04	3.922e+03	-5.750	1.13e-08	***
## TotRmsAbvGrd	1.798e+03	9.524e+02	1.888	0.059260	.
## FunctionalMaj2	-8.908e+02	1.432e+04	-0.062	0.950427	
## FunctionalMin1	7.300e+03	8.559e+03	0.853	0.393897	
## FunctionalMin2	8.936e+03	8.586e+03	1.041	0.298234	
## FunctionalMod	-5.054e+03	1.051e+04	-0.481	0.630719	
## FunctionalSev	-4.272e+04	2.897e+04	-1.474	0.140664	
## FunctionalTyp	1.842e+04	7.425e+03	2.481	0.013221	*
## Fireplaces	6.174e+03	2.550e+03	2.422	0.015600	*
## FireplaceQuEx	-8.729e+03	6.211e+03	-1.405	0.160146	
## FireplaceQuFa	-9.463e+03	5.324e+03	-1.777	0.075756	.
## FireplaceQuGd	-5.974e+03	3.409e+03	-1.753	0.079907	.
## FireplaceQuPo	3.626e+03	6.375e+03	0.569	0.569618	
## FireplaceQuTA	-5.021e+03	3.562e+03	-1.410	0.158904	
## GarageType2Types	-2.959e+04	1.212e+04	-2.442	0.014748	*
## GarageTypeAttchd	-9.589e+03	4.670e+03	-2.053	0.040280	*
## GarageTypeBasment	-5.228e+03	7.819e+03	-0.669	0.503825	
## GarageTypeBuiltIn	-9.819e+03	5.571e+03	-1.762	0.078256	.
## GarageTypeCarPort	-4.677e+03	1.054e+04	-0.444	0.657170	
## GarageTypeDetchd	-6.602e+03	4.637e+03	-1.424	0.154816	
## GarageYrBlt	-2.086e+01	5.835e+01	-0.357	0.720829	
## GarageFinishFin	6.560e+02	2.419e+03	0.271	0.786287	
## GarageFinishRFn	-1.772e+03	2.151e+03	-0.824	0.410220	
## GarageFinishUnf	NA	NA	NA	NA	
## GarageCars	3.990e+03	2.266e+03	1.761	0.078535	.
## GarageArea	1.826e+01	7.847e+00	2.327	0.020123	*
## GarageQualEx	1.192e+05	2.980e+04	3.999	6.74e-05	***
## GarageQualFa	-6.064e+03	4.850e+03	-1.250	0.211464	
## GarageQualGd	-1.230e+03	7.676e+03	-0.160	0.872757	
## GarageQualPo	-2.287e+04	2.388e+04	-0.957	0.338575	
## GarageQualTA	NA	NA	NA	NA	
## GarageCondEx	-1.135e+05	3.440e+04	-3.298	0.001003	**
## GarageCondFa	-1.841e+03	5.398e+03	-0.341	0.733169	
## GarageCondGd	-2.795e+03	9.202e+03	-0.304	0.761408	
## GarageCondPo	3.998e+03	1.386e+04	0.288	0.773023	
## GarageCondTA	NA	NA	NA	NA	
## PavedDriveP	-3.430e+03	5.540e+03	-0.619	0.535994	
## PavedDriveY	-2.380e+02	3.452e+03	-0.069	0.945046	
## WoodDeckSF	1.513e+01	5.863e+00	2.581	0.009964	**
## OpenPorchSF	4.665e-01	1.154e+01	0.040	0.967765	
## EnclosedPorch	3.354e+00	1.243e+01	0.270	0.787378	

```
## X3SsnPorch      3.522e+01  2.232e+01   1.578 0.114840
## ScreenPorch     3.592e+01  1.247e+01   2.881 0.004031 **
## PoolArea        6.792e+02  2.267e+02   2.996 0.002793 **
## PoolQCEX       -2.509e+05  1.227e+05  -2.045 0.041069 *
## PoolQCFa       -4.089e+05  1.493e+05  -2.738 0.006267 **
## PoolQCGd       -3.799e+05  1.469e+05  -2.587 0.009806 **
## FenceGdPrv     -8.964e+03  3.650e+03  -2.456 0.014181 *
## FenceGdWo      -9.230e+02  3.565e+03  -0.259 0.795776
## FenceMnPrv      5.697e+02  2.260e+03   0.252 0.801079
## FenceMnWw      -5.754e+03  7.454e+03  -0.772 0.440311
## MiscFeatureGar2 -7.928e+02  9.695e+04  -0.008 0.993477
## MiscFeatureOthr  1.456e+04  2.016e+04   0.722 0.470269
## MiscFeatureShed  2.402e+03  5.685e+03   0.422 0.672791
## MiscFeatureTenC  3.033e+04  4.726e+04   0.642 0.521154
## MiscVal         8.497e-02  6.095e+00   0.014 0.988879
## MoSold         -4.734e+02  2.444e+02  -1.937 0.052960 .
## YrSold         -5.753e+02  5.122e+02  -1.123 0.261565
## SaleTypeCon     2.572e+04  1.751e+04   1.469 0.142163
## SaleTypeConLD    1.637e+04  9.665e+03   1.694 0.090503 .
## SaleTypeConLI    4.766e+03  1.154e+04   0.413 0.679668
## SaleTypeConLw    9.828e+02  1.213e+04   0.081 0.935419
## SaleTypeCWD     1.495e+04  1.283e+04   1.165 0.244094
## SaleTypeNew     2.102e+04  1.538e+04   1.366 0.172093
## SaleTypeOth     6.782e+03  1.448e+04   0.468 0.639637
## SaleTypeWD      -4.508e+02  4.166e+03  -0.108 0.913840
## SaleConditionAdjLand 9.341e+03  1.458e+04   0.641 0.521798
## SaleConditionAlloca 8.081e+02  8.824e+03   0.092 0.927044
## SaleConditionFamily 1.003e+03  6.076e+03   0.165 0.868855
## SaleConditionNormal 6.754e+03  2.897e+03   2.331 0.019909 *
## SaleConditionPartial -7.710e+01  1.481e+04  -0.005 0.995846
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 22560 on 1209 degrees of freedom
## Multiple R-squared:  0.9332, Adjusted R-squared:  0.9194
## F-statistic: 67.55 on 250 and 1209 DF,  p-value: < 2.2e-16
```

```
pred <- predict(mymodel, dftrain)
```

```
## Warning in predict.lm(mymodel, dftrain): prediction from a rank-deficient
## fit may be misleading
```

```
plot(dftrain$SalePrice, pred)
```



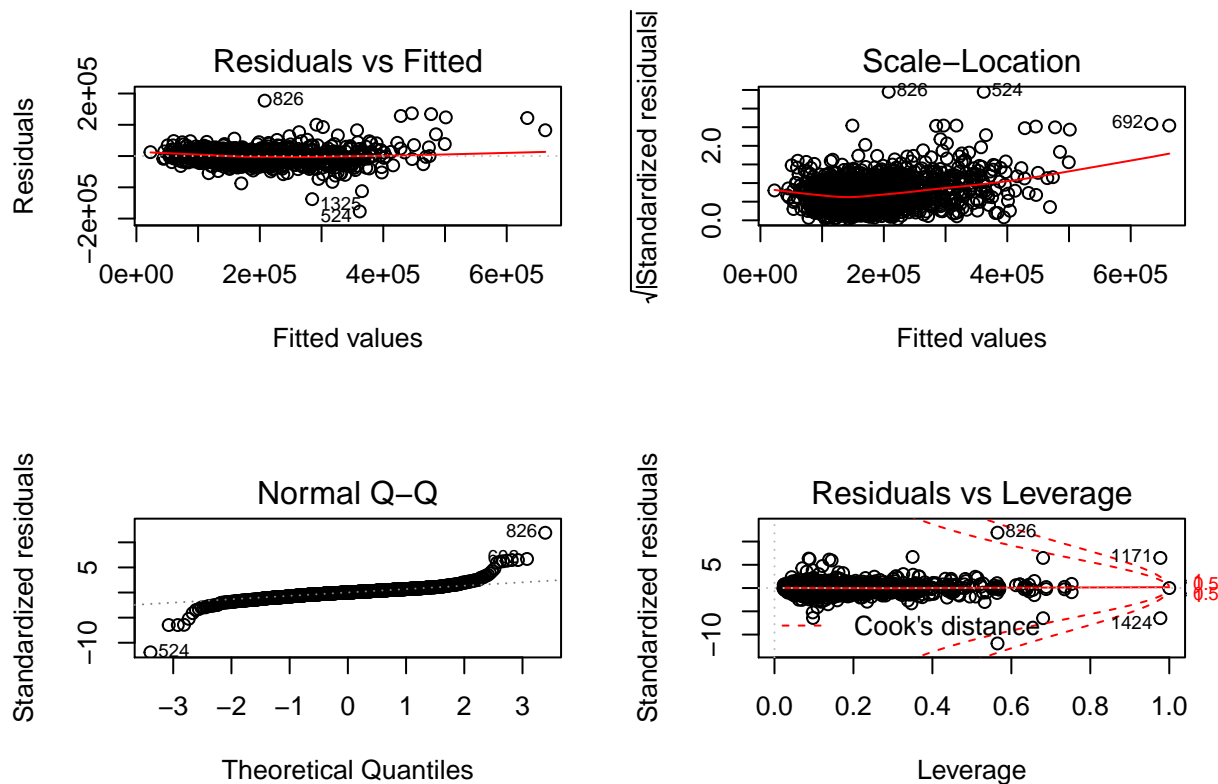
```
layout(matrix(c(1,2,3,4),2,2)) # optional 4 graphs/page
plot(mymodel)
```

```
## Warning: not plotting observations with leverage one:
## 121, 186, 251, 272, 326, 333, 347, 376, 399, 584, 596, 811, 945, 949, 1004, 1012, 1188, 1231, 1271

## Warning: not plotting observations with leverage one:
## 121, 186, 251, 272, 326, 333, 347, 376, 399, 584, 596, 811, 945, 949, 1004, 1012, 1188, 1231, 1271

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

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

```
dfest[,validcols] <-sapply(dfest[,validcols],fct_explicit_na)
dfest[,validcols] <- as.data.frame(sapply(dfest[,validcols], factor))
dfest$MasVnrType <- na.approx(dfest$MasVnrType)
dfest$LotFrontage <- na.approx(dfest$LotFrontage)
dfest$GarageYrBlt <- na.approx(dfest$GarageYrBlt)
dfest$MasVnrArea <- na.approx(dfest$MasVnrArea)
```

```
as.data.frame(as.table(colSums(is.na(dfest)))) %>% arrange(-Freq)
```

```
##      Var1 Freq
## 1  MSZoning      4
## 2  Utilities      2
## 3  BsmtFullBath      2
## 4  BsmtHalfBath      2
## 5   Functional      2
## 6 Exterior1st      1
## 7 Exterior2nd      1
## 8   BsmtFinSF1      1
## 9   BsmtFinSF2      1
## 10  BsmtUnfSF      1
## 11 TotalBsmtSF      1
## 12 KitchenQual      1
## 13 GarageCars      1
## 14 GarageArea      1
## 15   SaleType      1
```

## 16	Id	0
## 17	MSSubClass	0
## 18	LotFrontage	0
## 19	LotArea	0
## 20	Street	0
## 21	Alley	0
## 22	LotShape	0
## 23	LandContour	0
## 24	LotConfig	0
## 25	LandSlope	0
## 26	Neighborhood	0
## 27	Condition1	0
## 28	Condition2	0
## 29	BldgType	0
## 30	HouseStyle	0
## 31	OverallQual	0
## 32	OverallCond	0
## 33	YearBuilt	0
## 34	YearRemodAdd	0
## 35	RoofStyle	0
## 36	RoofMatl	0
## 37	MasVnrType	0
## 38	MasVnrArea	0
## 39	ExterQual	0
## 40	ExterCond	0
## 41	Foundation	0
## 42	BsmtQual	0
## 43	BsmtCond	0
## 44	BsmtExposure	0
## 45	BsmtFinType1	0
## 46	BsmtFinType2	0
## 47	Heating	0
## 48	HeatingQC	0
## 49	CentralAir	0
## 50	Electrical	0
## 51	X1stFlrSF	0
## 52	X2ndFlrSF	0
## 53	LowQualFinSF	0
## 54	GrLivArea	0
## 55	FullBath	0
## 56	HalfBath	0
## 57	BedroomAbvGr	0
## 58	KitchenAbvGr	0
## 59	TotRmsAbvGrd	0
## 60	Fireplaces	0
## 61	FireplaceQu	0
## 62	GarageType	0
## 63	GarageYrBlt	0
## 64	GarageFinish	0
## 65	GarageQual	0
## 66	GarageCond	0
## 67	PavedDrive	0
## 68	WoodDeckSF	0
## 69	OpenPorchSF	0

```
## 70 EnclosedPorch    0
## 71    X3SsnPorch    0
## 72    ScreenPorch    0
## 73        PoolArea    0
## 74        PoolQC    0
## 75         Fence    0
## 76   MiscFeature    0
## 77       MiscVal    0
## 78        MoSold    0
## 79        YrSold    0
## 80 SaleCondition    0
```

```
pred <- predict(mymodel, dftest)
```

```
## Warning in predict.lm(mymodel, dftest): prediction from a rank-deficient
## fit may be misleading
```