

Housing Data Exploratory Analysis

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September 26, 2016

```
library(data.table)
library(testthat)
library(gridExtra)
library(corrplot)
library(GGally)
library(ggplot2)
library(e1071)
library(dplyr)
```

```
## -----
```

```
## data.table + dplyr code now lives in dtplyr.
## Please library(dtplyr)!
```

```
## -----
```

```
##
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:GGally':
##
##      nasa
```

```
## The following object is masked from 'package:gridExtra':
##
##      combine
```

```
## The following object is masked from 'package:testthat':
##
##      matches
```

```
## The following objects are masked from 'package:data.table':
##
##      between, last
```

```
## The following objects are masked from 'package:stats':
##
##      filter, lag
```

```
## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union
```

```
cat_var <- names(train)[which(sapply(train, is.character))]
cat_car <- c(cat_var, 'BedroomAbvGr', 'HalfBath', 'KitchenAbvGr', 'BsmtFullBath', 'BsmtHalfBath', 'MSSubClass')
numeric_var <- names(train)[which(sapply(train, is.numeric))]
```

Structure of the data

The housing data set has 1460 rows and 81 features with the target feature Sale Price.

```
dim(train)
```

```
## [1] 1460    81
```

```
str(train)
```

```
## Classes 'data.table' and '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     : chr  "RL" "RL" "RL" "RL" ...
## $ 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       : chr  "Pave" "Pave" "Pave" "Pave" ...
## $ Alley        : chr  NA NA NA NA ...
## $ LotShape     : chr  "Reg" "Reg" "IR1" "IR1" ...
## $ LandContour  : chr  "Lvl" "Lvl" "Lvl" "Lvl" ...
## $ Utilities    : chr  "AllPub" "AllPub" "AllPub" "AllPub" ...
## $ LotConfig    : chr  "Inside" "FR2" "Inside" "Corner" ...
## $ LandSlope    : chr  "Gtl" "Gtl" "Gtl" "Gtl" ...
## $ Neighborhood : chr  "CollgCr" "Veenker" "CollgCr" "Crawfor" ...
## $ Condition1   : chr  "Norm" "Feedr" "Norm" "Norm" ...
## $ Condition2   : chr  "Norm" "Norm" "Norm" "Norm" ...
## $ BldgType     : chr  "1Fam" "1Fam" "1Fam" "1Fam" ...
## $ HouseStyle   : chr  "2Story" "1Story" "2Story" "2Story" ...
## $ 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    : chr  "Gable" "Gable" "Gable" "Gable" ...
## $ RoofMatl     : chr  "CompShg" "CompShg" "CompShg" "CompShg" ...
## $ Exterior1st  : chr  "VinylSd" "MetalSd" "VinylSd" "Wd Sdng" ...
## $ Exterior2nd  : chr  "VinylSd" "MetalSd" "VinylSd" "Wd Shng" ...
## $ MasVnrType   : chr  "BrkFace" "None" "BrkFace" "None" ...
## $ MasVnrArea   : int  196 0 162 0 350 0 186 240 0 0 ...
## $ ExterQual    : chr  "Gd" "TA" "Gd" "TA" ...
## $ ExterCond    : chr  "TA" "TA" "TA" "TA" ...
## $ Foundation   : chr  "PConc" "CBlock" "PConc" "BrkTil" ...
## $ BsmtQual     : chr  "Gd" "Gd" "Gd" "TA" ...
## $ BsmtCond     : chr  "TA" "TA" "TA" "Gd" ...
## $ BsmtExposure : chr  "No" "Gd" "Mn" "No" ...
## $ BsmtFinType1 : chr  "GLQ" "ALQ" "GLQ" "ALQ" ...
## $ BsmtFinSF1   : int  706 978 486 216 655 732 1369 859 0 851 ...
## $ BsmtFinType2 : chr  "Unf" "Unf" "Unf" "Unf" ...
```

```

## $ 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 : chr "GasA" "GasA" "GasA" "GasA" ...
## $ HeatingQC : chr "Ex" "Ex" "Ex" "Gd" ...
## $ CentralAir : chr "Y" "Y" "Y" "Y" ...
## $ Electrical : chr "SBrkr" "SBrkr" "SBrkr" "SBrkr" ...
## $ 1stFlrSF : int 856 1262 920 961 1145 796 1694 1107 1022 1077 ...
## $ 2ndFlrSF : 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 0 1 ...
## $ BsmtHalfBath : int 0 1 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 : chr "Gd" "TA" "Gd" "Gd" ...
## $ TotRmsAbvGrd : int 8 6 6 7 9 5 7 7 8 5 ...
## $ Functional : chr "Typ" "Typ" "Typ" "Typ" ...
## $ Fireplaces : int 0 1 1 1 1 0 1 2 2 2 ...
## $ FireplaceQu : chr NA "TA" "TA" "Gd" ...
## $ GarageType : chr "Attchd" "Attchd" "Attchd" "Detchd" ...
## $ GarageYrBlt : int 2003 1976 2001 1998 2000 1993 2004 1973 1931 1939 ...
## $ GarageFinish : chr "RFn" "RFn" "RFn" "Unf" ...
## $ 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 : chr "TA" "TA" "TA" "TA" ...
## $ GarageCond : chr "TA" "TA" "TA" "TA" ...
## $ PavedDrive : chr "Y" "Y" "Y" "Y" ...
## $ 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 ...
## $ 3SsnPorch : 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 : chr NA NA NA NA ...
## $ Fence : chr NA NA NA NA ...
## $ MiscFeature : chr NA NA 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 : chr "WD" "WD" "WD" "WD" ...
## $ SaleCondition: chr "Normal" "Normal" "Normal" "Abnorml" ...
## $ SalePrice : int 208500 181500 223500 140000 250000 143000 307000 200000 129900 118000 ...
## - attr(*, ".internal.selfref")=<externalptr>

```

Summarize the missing values in the data.

Viewing the first five rows of the data indicates that there are columns which have missing values. The categorical variables with the largest number of missing values are: Alley, FireplaceQu, PoolQC, Fence, and MiscFeature.

- Alley: indicates the type of alley access
- FireplaceQu: Fireplace Quality
- PoolQC: Pool Quality
- Fence: Fence Quality
- MiscFeature: Miscellaneous features not covered in other categories

The missing values indicate that majority of the houses do not have alley access, no pool, no fence and no elevator, 2nd garage, shed or tennis court that is covered by the MiscFeature.

The numeric variables do not have as many missing values but there are still some present. There are 259 values for the LotFrontage, 8 missing values for MasVnrArea and 81 missing values for GarageYrBlt.

- LotFrontage: Linear feet of street connected to property
- GarageYrBlt: Year garage was built
- MasVnrArea: Masonry veneer area in square feet

Definition of Masonry Veneer from google: Veneer masonry is a popular choice for home building and remodeling, because it gives the appearance of a solid brick or stone wall while providing better economy and insulation. It can be used as an addition to conventional wood frame structures, and can be placed on concrete block walls.

Brick veneers are not essential to the structure of the house but are used to change the appearance of the wall while providing better insulation. They tend to only have one brick layer.

```
head(train)
```

```
##      Id MSSubClass MSZoning LotFrontage LotArea Street Alley LotShape
## 1:    1          60      RL          65    8450   Pave   NA      Reg
## 2:    2          20      RL          80    9600   Pave   NA      Reg
## 3:    3          60      RL          68   11250   Pave   NA     IR1
## 4:    4          70      RL          60    9550   Pave   NA     IR1
## 5:    5          60      RL          84   14260   Pave   NA     IR1
## 6:    6          50      RL          85   14115   Pave   NA     IR1
##      LandContour Utilities LotConfig LandSlope Neighborhood Condition1
## 1:          Lvl1    AllPub    Inside      Gtl1    CollgCr      Norm
## 2:          Lvl1    AllPub    FR2        Gtl1    Veenker    Feedr
## 3:          Lvl1    AllPub    Inside      Gtl1    CollgCr      Norm
## 4:          Lvl1    AllPub    Corner      Gtl1    Crawfor      Norm
## 5:          Lvl1    AllPub    FR2        Gtl1    NoRidge      Norm
## 6:          Lvl1    AllPub    Inside      Gtl1    Mitchel      Norm
##      Condition2 BldgType HouseStyle OverallQual OverallCond YearBuilt
## 1:          Norm    1Fam    2Story          7           5      2003
## 2:          Norm    1Fam    1Story          6           8      1976
## 3:          Norm    1Fam    2Story          7           5      2001
## 4:          Norm    1Fam    2Story          7           5      1915
## 5:          Norm    1Fam    2Story          8           5      2000
## 6:          Norm    1Fam    1.5Fin          5           5      1993
##      YearRemodAdd RoofStyle RoofMatl Exterior1st Exterior2nd MasVnrType
## 1:          2003     Gable  CompShg    VinylSd    VinylSd    BrkFace
## 2:          1976     Gable  CompShg    MetalSd    MetalSd      None
## 3:          2002     Gable  CompShg    VinylSd    VinylSd    BrkFace
## 4:          1970     Gable  CompShg    Wd Sdng    Wd Shng      None
## 5:          2000     Gable  CompShg    VinylSd    VinylSd    BrkFace
## 6:          1995     Gable  CompShg    VinylSd    VinylSd      None
```

##	MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond	
## 1:	196	Gd	TA	PConc	Gd	TA	
## 2:	0	TA	TA	CBlock	Gd	TA	
## 3:	162	Gd	TA	PConc	Gd	TA	
## 4:	0	TA	TA	BrkTil	TA	Gd	
## 5:	350	Gd	TA	PConc	Gd	TA	
## 6:	0	TA	TA	Wood	Gd	TA	
##	BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	BsmtUnfSF	
## 1:	No	GLQ	706	Unf	0	150	
## 2:	Gd	ALQ	978	Unf	0	284	
## 3:	Mn	GLQ	486	Unf	0	434	
## 4:	No	ALQ	216	Unf	0	540	
## 5:	Av	GLQ	655	Unf	0	490	
## 6:	No	GLQ	732	Unf	0	64	
##	TotalBsmtSF	Heating	HeatingQC	CentralAir	Electrical	1stFlrSF	2ndFlrSF
## 1:	856	GasA	Ex	Y	SBrkr	856	854
## 2:	1262	GasA	Ex	Y	SBrkr	1262	0
## 3:	920	GasA	Ex	Y	SBrkr	920	866
## 4:	756	GasA	Gd	Y	SBrkr	961	756
## 5:	1145	GasA	Ex	Y	SBrkr	1145	1053
## 6:	796	GasA	Ex	Y	SBrkr	796	566
##	LowQualFinSF	GrLivArea	BsmtFullBath	BsmtHalfBath	FullBath	HalfBath	
## 1:	0	1710	1	0	2	1	
## 2:	0	1262	0	1	2	0	
## 3:	0	1786	1	0	2	1	
## 4:	0	1717	1	0	1	0	
## 5:	0	2198	1	0	2	1	
## 6:	0	1362	1	0	1	1	
##	BedroomAbvGr	KitchenAbvGr	KitchenQual	TotRmsAbvGrd	Functional		
## 1:	3	1	Gd	8	Typ		
## 2:	3	1	TA	6	Typ		
## 3:	3	1	Gd	6	Typ		
## 4:	3	1	Gd	7	Typ		
## 5:	4	1	Gd	9	Typ		
## 6:	1	1	TA	5	Typ		
##	Fireplaces	FireplaceQu	GarageType	GarageYrBlt	GarageFinish	GarageCars	
## 1:	0	NA	Attchd	2003	RFn	2	
## 2:	1	TA	Attchd	1976	RFn	2	
## 3:	1	TA	Attchd	2001	RFn	2	
## 4:	1	Gd	Detchd	1998	Unf	3	
## 5:	1	TA	Attchd	2000	RFn	3	
## 6:	0	NA	Attchd	1993	Unf	2	
##	GarageArea	GarageQual	GarageCond	PavedDrive	WoodDeckSF	OpenPorchSF	
## 1:	548	TA	TA	Y	0	61	
## 2:	460	TA	TA	Y	298	0	
## 3:	608	TA	TA	Y	0	42	
## 4:	642	TA	TA	Y	0	35	
## 5:	836	TA	TA	Y	192	84	
## 6:	480	TA	TA	Y	40	30	
##	EnclosedPorch	3SsnPorch	ScreenPorch	PoolArea	PoolQC	Fence	MiscFeature
## 1:	0	0	0	0	NA	NA	NA
## 2:	0	0	0	0	NA	NA	NA
## 3:	0	0	0	0	NA	NA	NA
## 4:	272	0	0	0	NA	NA	NA

```
## 5:      0      0      0      0      NA      NA      NA
## 6:      0     320      0      0      NA MnPrv      Shed
##      MiscVal MoSold YrSold SaleType SaleCondition SalePrice
## 1:      0      2    2008      WD      Normal    208500
## 2:      0      5    2007      WD      Normal    181500
## 3:      0      9    2008      WD      Normal    223500
## 4:      0      2    2006      WD      Abnorml    140000
## 5:      0     12    2008      WD      Normal    250000
## 6:     700     10    2009      WD      Normal    143000
```

```
colSums(sapply(train, is.na))
```

```
##      Id      MSSubClass      MSZoning      LotFrontage      LotArea
##      0      0      0      259      0
##      Street      Alley      LotShape      LandContour      Utilities
##      0      1369      0      0      0
##      LotConfig      LandSlope      Neighborhood      Condition1      Condition2
##      0      0      0      0      0
##      BldgType      HouseStyle      OverallQual      OverallCond      YearBuilt
##      0      0      0      0      0
##      YearRemodAdd      RoofStyle      RoofMatl      Exterior1st      Exterior2nd
##      0      0      0      0      0
##      MasVnrType      MasVnrArea      ExterQual      ExterCond      Foundation
##      8      8      0      0      0
##      BsmtQual      BsmtCond      BsmtExposure      BsmtFinType1      BsmtFinSF1
##      37      37      38      37      0
##      BsmtFinType2      BsmtFinSF2      BsmtUnfSF      TotalBsmtSF      Heating
##      38      0      0      0      0
##      HeatingQC      CentralAir      Electrical      1stFlrSF      2ndFlrSF
##      0      0      1      0      0
##      LowQualFinSF      GrLivArea      BsmtFullBath      BsmtHalfBath      FullBath
##      0      0      0      0      0
##      HalfBath      BedroomAbvGr      KitchenAbvGr      KitchenQual      TotRmsAbvGrd
##      0      0      0      0      0
##      Functional      Fireplaces      FireplaceQu      GarageType      GarageYrBlt
##      0      0      690      81      81
##      GarageFinish      GarageCars      GarageArea      GarageQual      GarageCond
##      81      0      0      81      81
##      PavedDrive      WoodDeckSF      OpenPorchSF      EnclosedPorch      3SsnPorch
##      0      0      0      0      0
##      ScreenPorch      PoolArea      PoolQC      Fence      MiscFeature
##      0      0      1453      1179      1406
##      MiscVal      MoSold      YrSold      SaleType      SaleCondition
##      0      0      0      0      0
##      SalePrice
##      0
```

```
colSums(sapply(train[,.SD, .SDcols = cat_var], is.na))
```

```
##      MSZoning      Street      Alley      LotShape      LandContour
##      0      0      1369      0      0
##      Utilities      LotConfig      LandSlope      Neighborhood      Condition1
##      0      0      0      0      0
```

```
##      Condition2      BldgType      HouseStyle      RoofStyle      RoofMatl
##              0              0              0              0              0
##      Exterior1st      Exterior2nd      MasVnrType      ExterQual      ExterCond
##              0              0              8              0              0
##      Foundation      BsmtQual      BsmtCond      BsmtExposure      BsmtFinType1
##              0              37              37              38              37
##      BsmtFinType2      Heating      HeatingQC      CentralAir      Electrical
##              38              0              0              0              1
##      KitchenQual      Functional      FireplaceQu      GarageType      GarageFinish
##              0              0              690              81              81
##      GarageQual      GarageCond      PavedDrive      PoolQC      Fence
##              81              81              0              1453              1179
##      MiscFeature      SaleType      SaleCondition
##              1406              0              0
```

```
colSums(sapply(train[,.SD, .SDcols = numeric_var], is.na))
```

```
##              Id      MSSubClass      LotFrontage      LotArea      OverallQual
##              0              0              259              0              0
##      OverallCond      YearBuilt      YearRemodAdd      MasVnrArea      BsmtFinSF1
##              0              0              0              8              0
##      BsmtFinSF2      BsmtUnfSF      TotalBsmtSF      1stFlrSF      2ndFlrSF
##              0              0              0              0              0
##      LowQualFinSF      GrLivArea      BsmtFullBath      BsmtHalfBath      FullBath
##              0              0              0              0              0
##      HalfBath      BedroomAbvGr      KitchenAbvGr      TotRmsAbvGrd      Fireplaces
##              0              0              0              0              0
##      GarageYrBlt      GarageCars      GarageArea      WoodDeckSF      OpenPorchSF
##              81              0              0              0              0
##      EnclosedPorch      3SsnPorch      ScreenPorch      PoolArea      MiscVal
##              0              0              0              0              0
##      MoSold      YrSold      SalePrice
##              0              0              0
```

Let's gain some insight on the number of houses that were remodeled. According to the data dictionary, if the YearBuilt date is different from the YearRemodAdd date then the house was remodeled. Comparing these two rows indicates that 696 houses were remodeled and 764 houses were not remodeled.

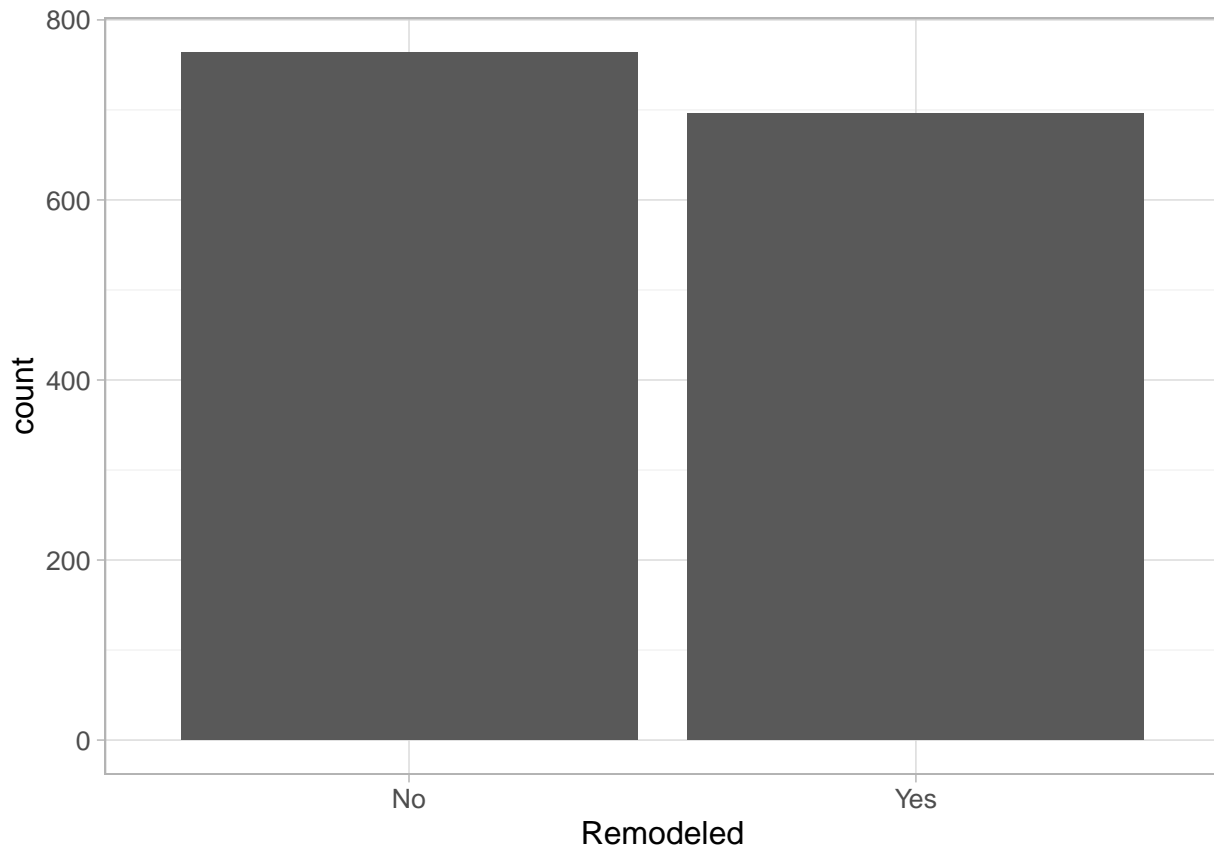
```
sum(train[, 'YearRemodAdd', with = FALSE] != train[, 'YearBuilt', with = FALSE])
```

```
## [1] 696
```

```
cat('Percentage of houses remodeled', sum(train[, 'YearRemodAdd', with = FALSE] != train[, 'YearBuilt', with = FALSE]) / nrow(train), '\n')
```

```
## Percentage of houses remodeled 0.4767123
```

```
train %>% select(YearBuilt, YearRemodAdd) %>% mutate(Remodeled = as.integer(YearBuilt != YearRemodAdd))
```



Summarize the numeric values and the structure of the data.

```
summary(train[,.SD, .SDcols =numeric_var])
```

```
##      Id      MSSubClass  LotFrontage  LotArea
## Min.   : 1.0    Min.   : 20.0    Min.   : 21.00   Min.   : 1300
## 1st Qu.: 365.8  1st Qu.: 20.0    1st Qu.: 59.00   1st Qu.: 7554
## Median : 730.5  Median : 50.0    Median : 69.00   Median : 9478
## Mean   : 730.5  Mean   : 56.9    Mean   : 70.05   Mean   : 10517
## 3rd Qu.:1095.2  3rd Qu.: 70.0    3rd Qu.: 80.00   3rd Qu.: 11602
## Max.   :1460.0  Max.   :190.0    Max.   :313.00   Max.   :215245
##
##      OverallQual  OverallCond  YearBuilt  YearRemodAdd
## Min.   : 1.000    Min.   :1.000    Min.   :1872   Min.   :1950
## 1st Qu.: 5.000    1st Qu.:5.000    1st Qu.:1954   1st Qu.:1967
## Median : 6.000    Median :5.000    Median :1973   Median :1994
## Mean   : 6.099    Mean   :5.575    Mean   :1971   Mean   :1985
## 3rd Qu.: 7.000    3rd Qu.:6.000    3rd Qu.:2000   3rd Qu.:2004
## Max.   :10.000    Max.   :9.000    Max.   :2010   Max.   :2010
##
##      MasVnrArea  BsmtFinSF1  BsmtFinSF2  BsmtUnfSF
## Min.   : 0.0    Min.   : 0.0    Min.   : 0.00   Min.   : 0.0
## 1st Qu.: 0.0    1st Qu.: 0.0    1st Qu.: 0.00   1st Qu.: 223.0
```



```

## Median : 0.0 Median : 383.5 Median : 0.00 Median : 477.5
## Mean : 103.7 Mean : 443.6 Mean : 46.55 Mean : 567.2
## 3rd Qu.: 166.0 3rd Qu.: 712.2 3rd Qu.: 0.00 3rd Qu.: 808.0
## Max. :1600.0 Max. :5644.0 Max. :1474.00 Max. :2336.0
## NA's :8
## TotalBsmtSF 1stFlrSF 2ndFlrSF LowQualFinSF
## Min. : 0.0 Min. : 334 Min. : 0 Min. : 0.000
## 1st Qu.: 795.8 1st Qu.: 882 1st Qu.: 0 1st Qu.: 0.000
## Median : 991.5 Median :1087 Median : 0 Median : 0.000
## Mean :1057.4 Mean :1163 Mean : 347 Mean : 5.845
## 3rd Qu.:1298.2 3rd Qu.:1391 3rd Qu.: 728 3rd Qu.: 0.000
## Max. :6110.0 Max. :4692 Max. :2065 Max. :572.000
##
## GrLivArea BsmtFullBath BsmtHalfBath FullBath
## Min. : 334 Min. :0.0000 Min. :0.00000 Min. :0.000
## 1st Qu.:1130 1st Qu.:0.0000 1st Qu.:0.00000 1st Qu.:1.000
## Median :1464 Median :0.0000 Median :0.00000 Median :2.000
## Mean :1515 Mean :0.4253 Mean :0.05753 Mean :1.565
## 3rd Qu.:1777 3rd Qu.:1.0000 3rd Qu.:0.00000 3rd Qu.:2.000
## Max. :5642 Max. :3.0000 Max. :2.00000 Max. :3.000
##
## HalfBath BedroomAbvGr KitchenAbvGr TotRmsAbvGrd
## Min. :0.0000 Min. :0.000 Min. :0.000 Min. : 2.000
## 1st Qu.:0.0000 1st Qu.:2.000 1st Qu.:1.000 1st Qu.: 5.000
## Median :0.0000 Median :3.000 Median :1.000 Median : 6.000
## Mean :0.3829 Mean :2.866 Mean :1.047 Mean : 6.518
## 3rd Qu.:1.0000 3rd Qu.:3.000 3rd Qu.:1.000 3rd Qu.: 7.000
## Max. :2.0000 Max. :8.000 Max. :3.000 Max. :14.000
##
## Fireplaces GarageYrBlt GarageCars GarageArea
## Min. :0.000 Min. :1900 Min. :0.000 Min. : 0.0
## 1st Qu.:0.000 1st Qu.:1961 1st Qu.:1.000 1st Qu.: 334.5
## Median :1.000 Median :1980 Median :2.000 Median : 480.0
## Mean :0.613 Mean :1979 Mean :1.767 Mean : 473.0
## 3rd Qu.:1.000 3rd Qu.:2002 3rd Qu.:2.000 3rd Qu.: 576.0
## Max. :3.000 Max. :2010 Max. :4.000 Max. :1418.0
## NA's :81
## WoodDeckSF OpenPorchSF EnclosedPorch 3SsnPorch
## Min. : 0.00 Min. : 0.00 Min. : 0.00 Min. : 0.00
## 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.00
## Median : 0.00 Median : 25.00 Median : 0.00 Median : 0.00
## Mean : 94.24 Mean : 46.66 Mean : 21.95 Mean : 3.41
## 3rd Qu.:168.00 3rd Qu.: 68.00 3rd Qu.: 0.00 3rd Qu.: 0.00
## Max. :857.00 Max. :547.00 Max. :552.00 Max. :508.00
##
## ScreenPorch PoolArea MiscVal MoSold
## Min. : 0.00 Min. : 0.000 Min. : 0.00 Min. : 1.000
## 1st Qu.: 0.00 1st Qu.: 0.000 1st Qu.: 0.00 1st Qu.: 5.000
## Median : 0.00 Median : 0.000 Median : 0.00 Median : 6.000
## Mean : 15.06 Mean : 2.759 Mean : 43.49 Mean : 6.322
## 3rd Qu.: 0.00 3rd Qu.: 0.000 3rd Qu.: 0.00 3rd Qu.: 8.000
## Max. :480.00 Max. :738.000 Max. :15500.00 Max. :12.000
##
## YrSold SalePrice

```

```
## Min.      :2006    Min.      : 34900
## 1st Qu.:2007    1st Qu.:129975
## Median :2008    Median :163000
## Mean      :2008    Mean      :180921
## 3rd Qu.:2009    3rd Qu.:214000
## Max.      :2010    Max.      :755000
##
```

```
cat('Train has', dim(train)[1], 'rows and', dim(train)[2], 'columns.')
```

```
## Train has 1460 rows and 81 columns.
```

```
cat('Test has', dim(test)[1], 'rows and', dim(test)[2], ' columns.')
```

```
## Test has 1459 rows and 80 columns.
```

```
# The percentage of data missing in train.
sum(is.na(train)) / (nrow(train) * ncol(train))
```

```
## [1] 0.05889565
```

```
# The percentage of data missing in test.
sum(is.na(test)) / (nrow(test) * ncol(test))
```

```
## [1] 0.05997258
```

```
# Check for duplicated rows.
```

```
cat("The number of duplicated rows are", nrow(train) - nrow(unique(train)))
```

```
## The number of duplicated rows are 0
```

```
####Convert character to factors
```

```
train[, (cat_var) := lapply(.SD, as.factor), .SDcols = cat_var]
```

```
train_cat <- train[, .SD, .SDcols = cat_var]
train_cont <- train[, .SD, .SDcols = numeric_var]
```

```
plotHist <- function(data_in, i) {
  data <- data.frame(x=data_in[[i]])
  p <- ggplot(data=data, aes(x=factor(x))) + stat_count() + xlab(colnames(data_in)[i]) + theme_light() +
    theme(axis.text.x = element_text(angle = 90, hjust =1))
  return (p)
}
```

```
doPlots <- function(data_in, fun, ii, ncol=3) {
  pp <- list()
  for (i in ii) {
    p <- fun(data_in=data_in, i=i)
```

```

    pp <- c(pp, list(p))
  }
  do.call("grid.arrange", c(pp, ncol=ncol))
}

plotDen <- function(data_in, i){
  data <- data.frame(x=data_in[[i]], SalePrice = data_in$SalePrice)
  p <- ggplot(data= data) + geom_line(aes(x = x), stat = 'density', size = 1,alpha = 1.0) +
    xlab(paste0((colnames(data_in)[i]), '\n', 'Skewness: ',round(skewness(data_in[[i]], na.rm = TRUE), 2)))
  return(p)
}

```

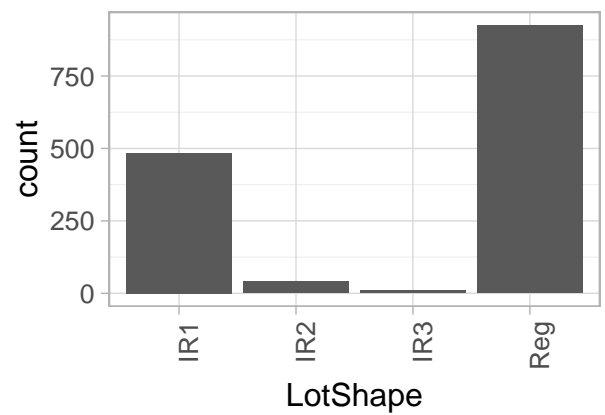
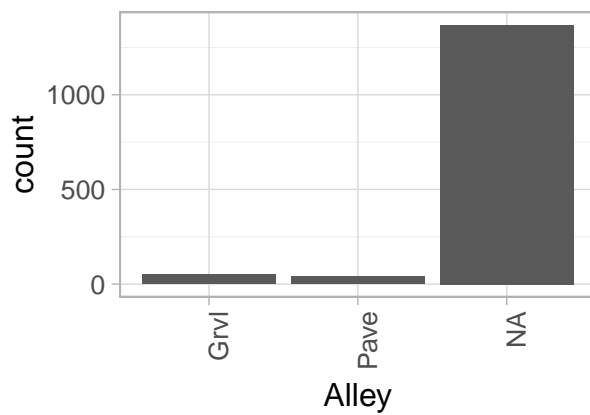
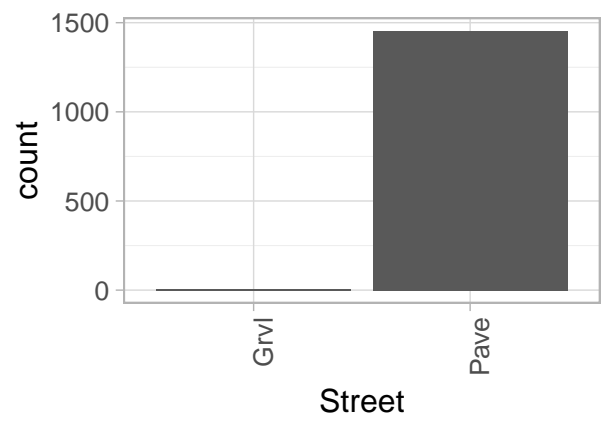
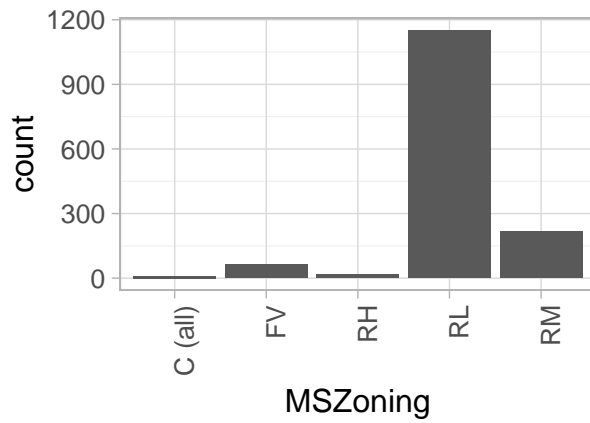
Barplots for the categorical features

The bar plots below offer more insight into the data. MSZoning: bar plot indicates that majority of the houses are located in low density residential areas and medium density residential area.

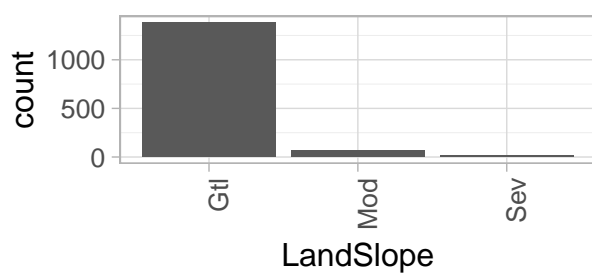
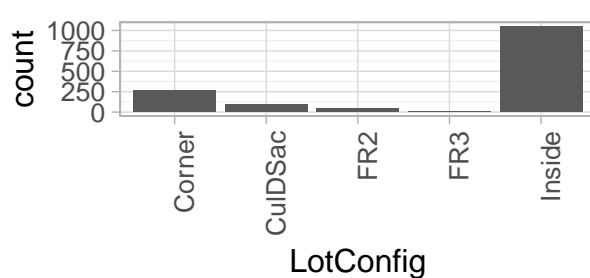
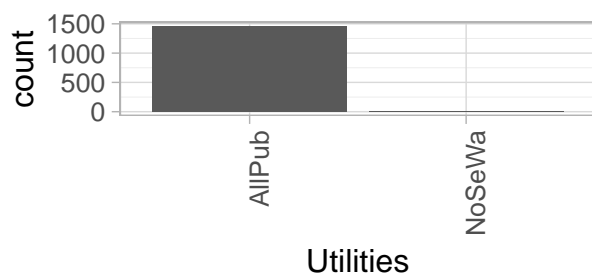
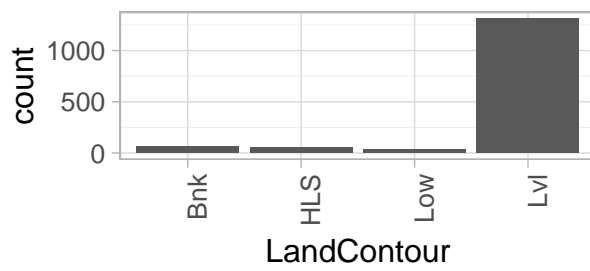
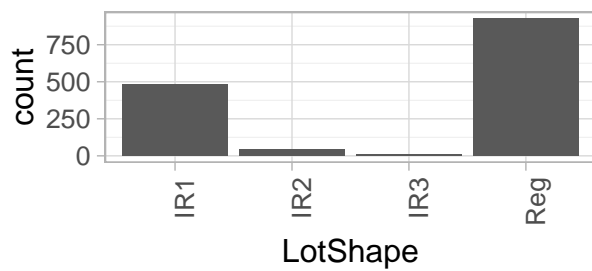
The type of road access to the property tends to be paved and the houses do not have alleys.

- Landcontour: the houses are built on flat properties
- Utilities: Almost all homes have all public utilities (E,G,W, & S)
- LandSlope: most of the properties have a gentle slope

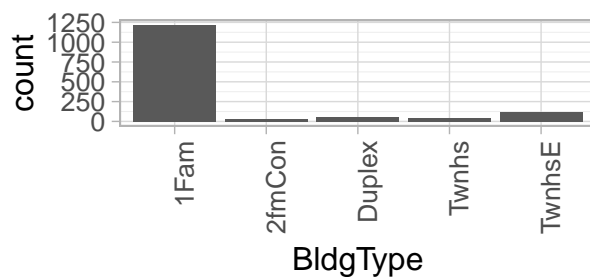
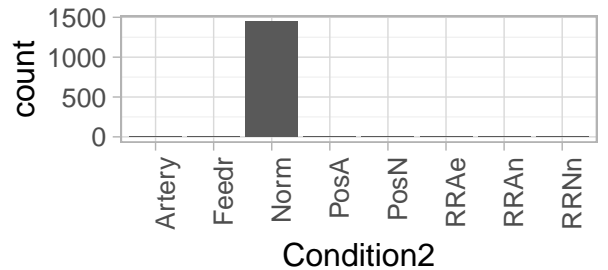
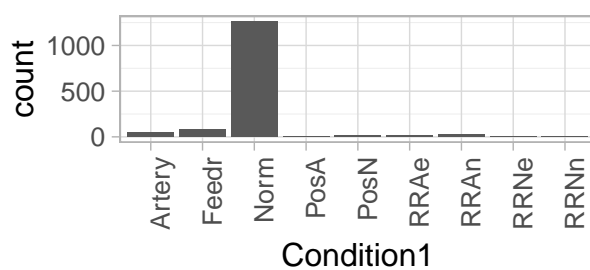
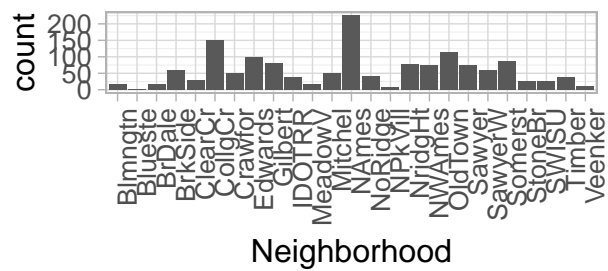
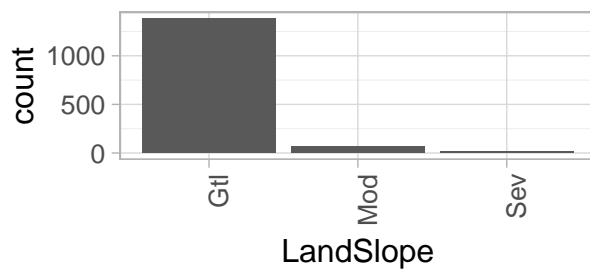
```
doPlots(train_cat, fun = plotHist, ii = 1:4, ncol = 2)
```



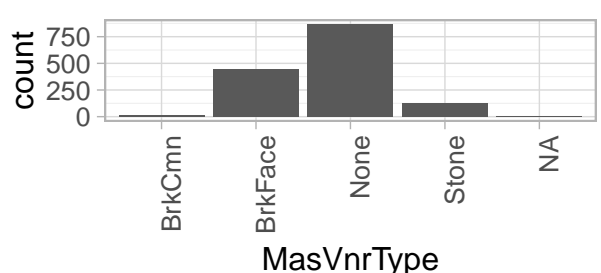
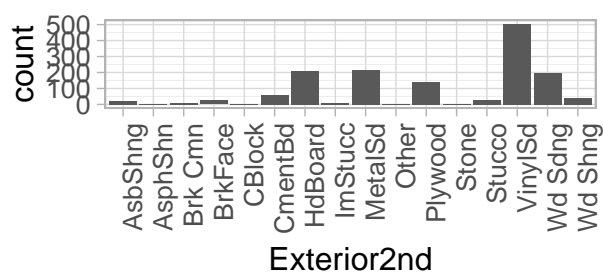
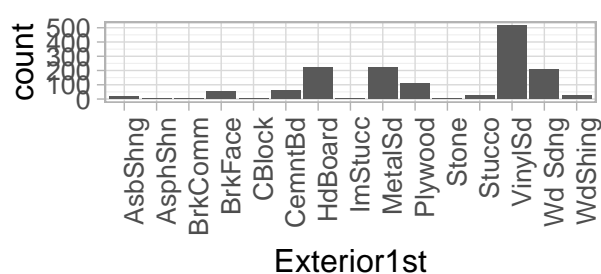
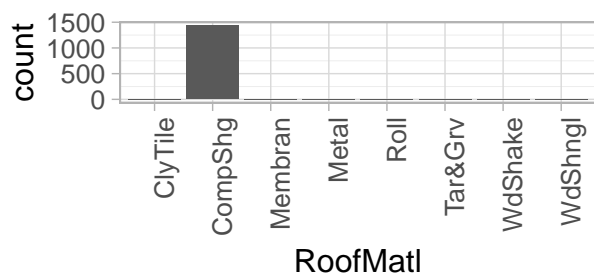
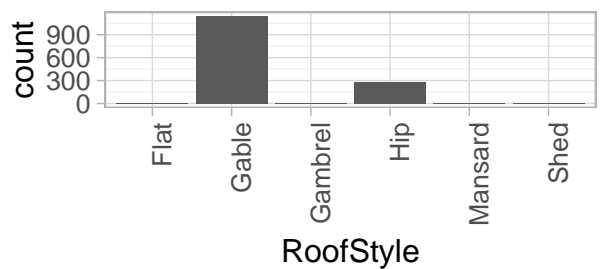
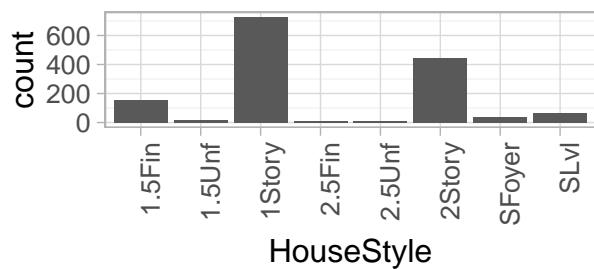
```
doPlots(train_cat, fun = plotHist, ii = 4:8, ncol = 2)
```



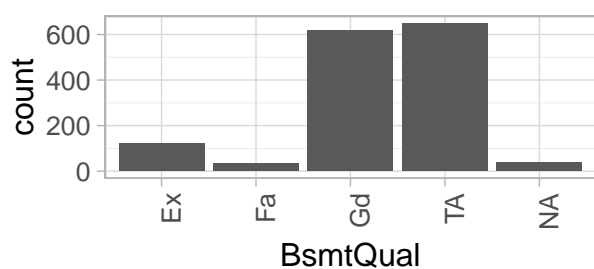
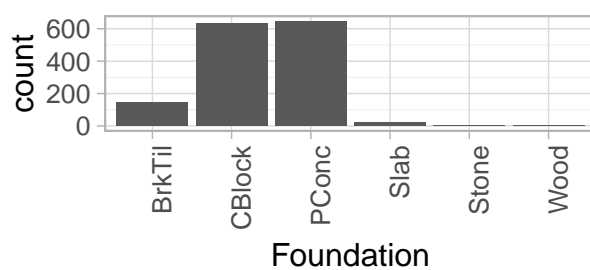
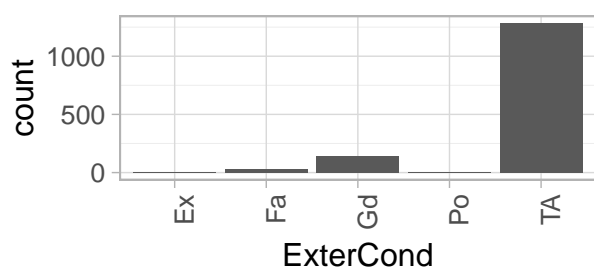
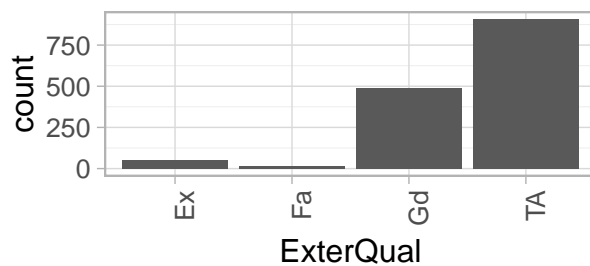
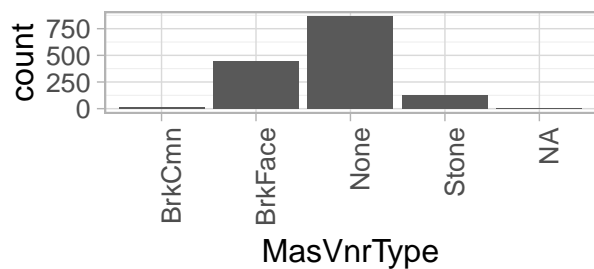
```
doPlots(train_cat, fun = plotHist, ii = 8:12, ncol = 2)
```



```
doPlots(train_cat, fun = plotHist, ii = 13:18, ncol = 2)
```

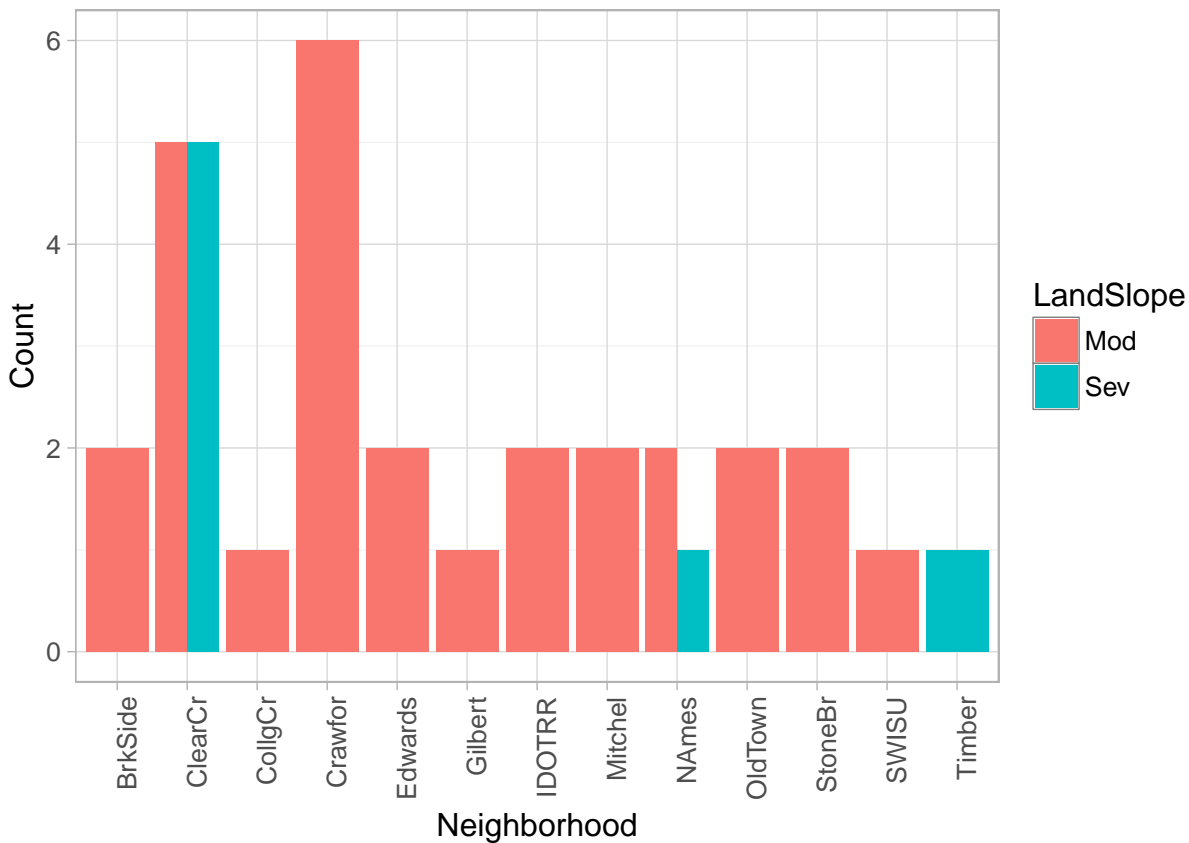


```
doPlots(train_cat, fun = plotHist, ii = 18:22, ncol = 2)
```



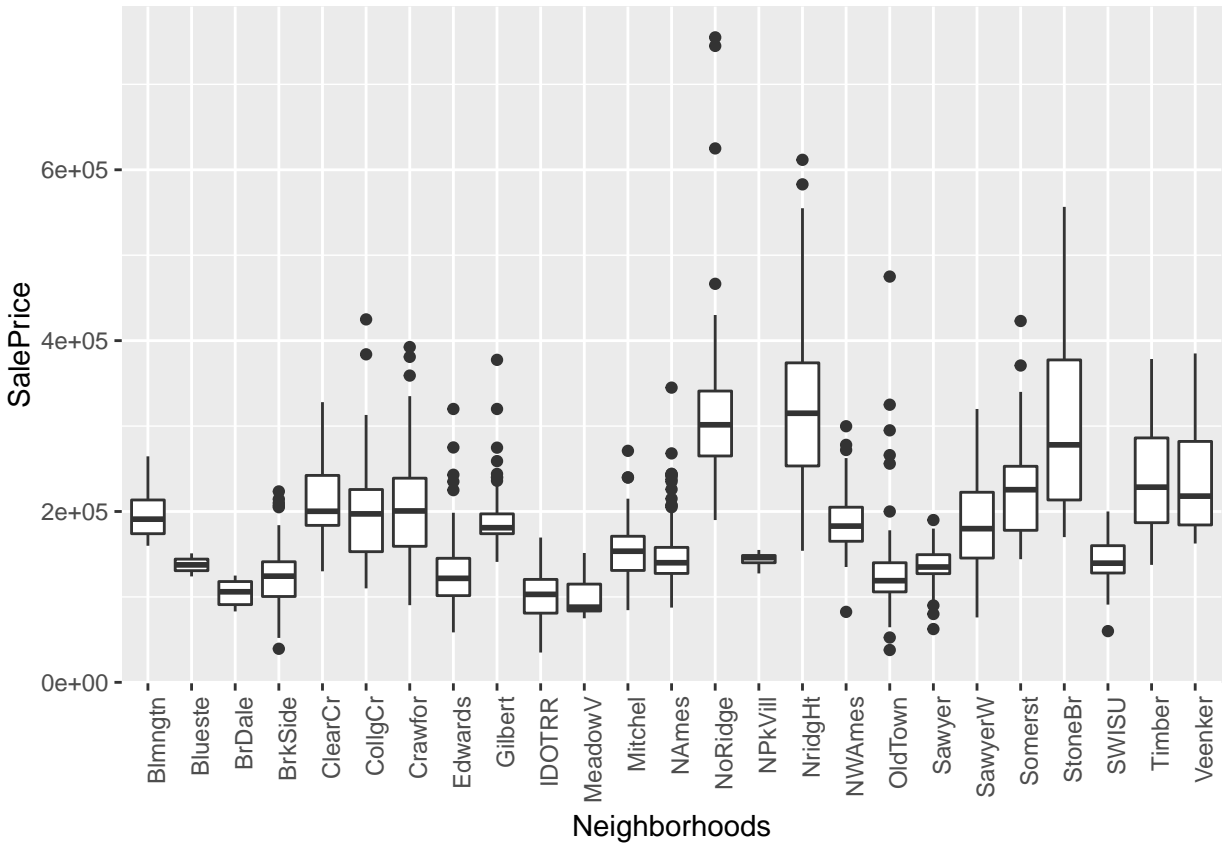
The houses that have sever landslope are located in the Clear Creek and Timberland. The houses with moderate landslope are present in more neighborhood. The Clear Creek and the Crawford neighborhoods seem to have high slopes.

```
train %>% select(LandSlope, Neighborhood, SalePrice) %>% filter(LandSlope == c('Sev', 'Mod')) %>% arrange(
```

Plotting a boxplot between the neighborhoods and sale price shows that BrookSide and South & West of Iowa State University have cheap houses. While Northridge and Northridge Heights are rich neighborhoods with several outliers in terms of price.

```
train %>% select(Neighborhood, SalePrice) %>% ggplot(aes(factor(Neighborhood), SalePrice)) + geom_boxplot
```

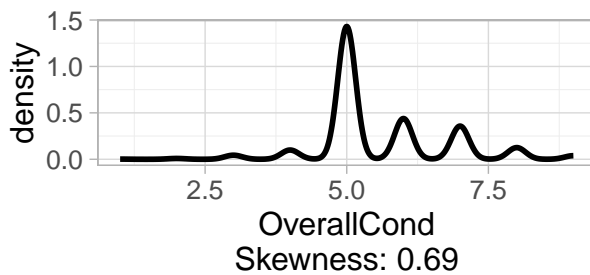
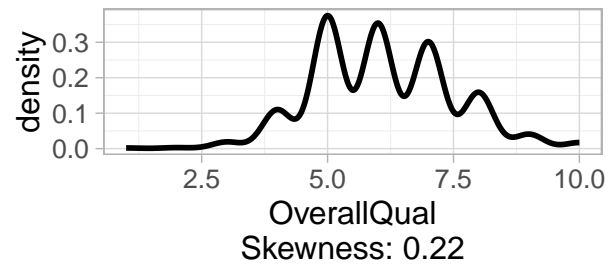
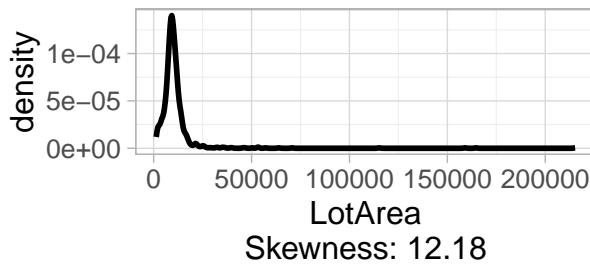
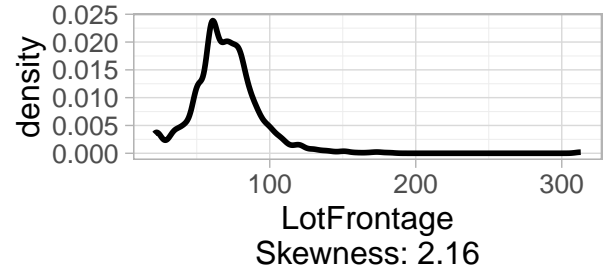
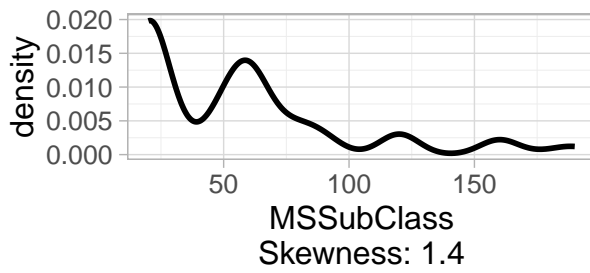


Density plots for numeric variables.

Density plots of the features indicates that the features are skewed. The density plot for YearBuilt shows that the data set contains a mix of new and old houses. It shows a downturn in the number of houses in recent years, possibly due to the housing crisis.

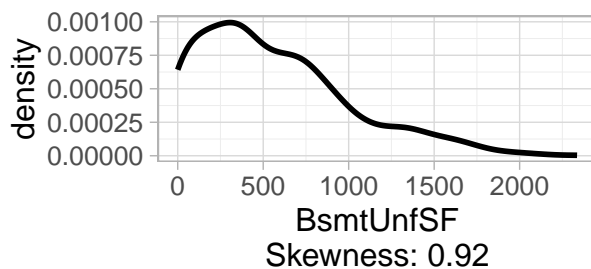
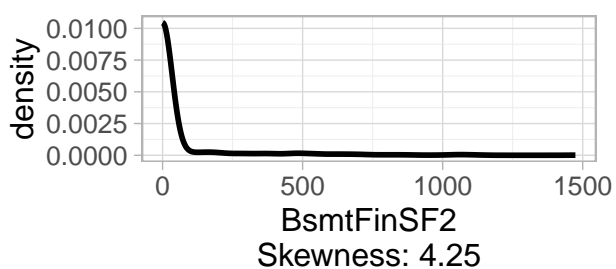
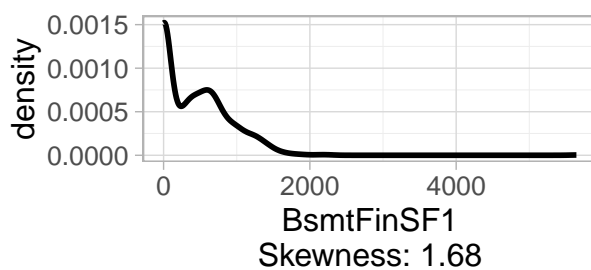
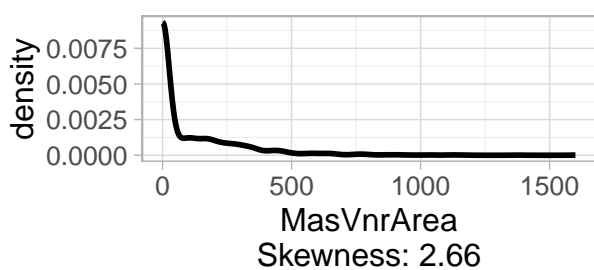
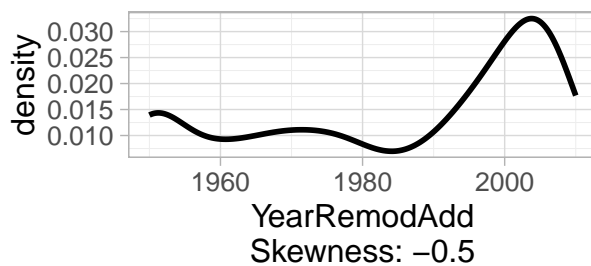
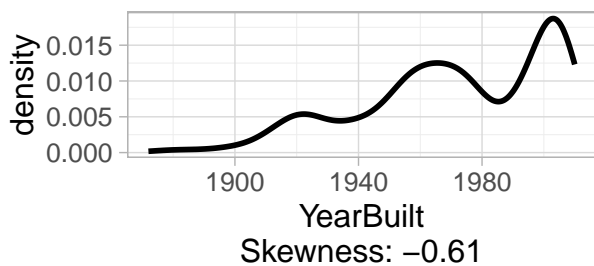
```
doPlots(train_cont, fun = plotDen, ii = 2:6, ncol = 2)
```

```
## Warning: Removed 259 rows containing non-finite values (stat_density).
```

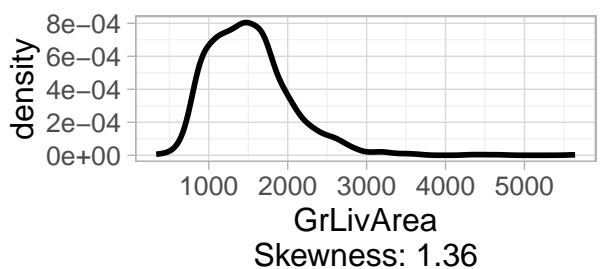
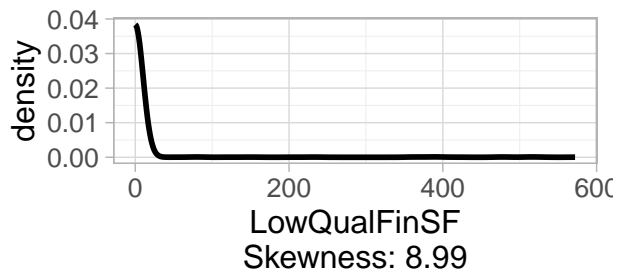
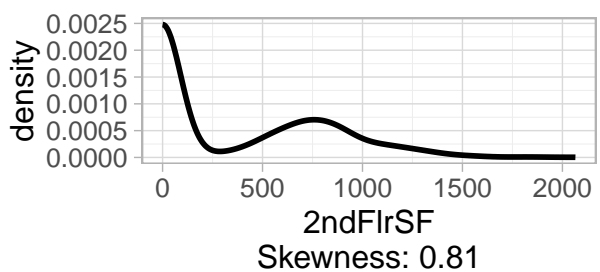
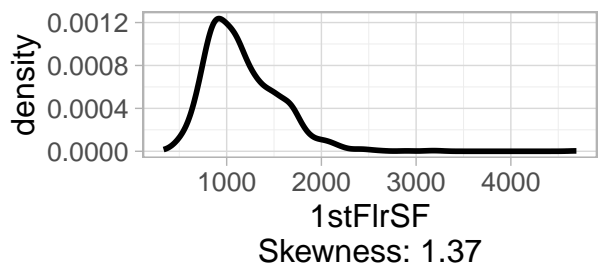
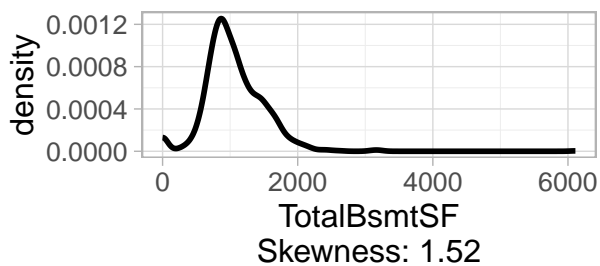


```
doPlots(train_cont, fun = plotDen, ii = 7:12, ncol = 2)
```

```
## Warning: Removed 8 rows containing non-finite values (stat_density).
```

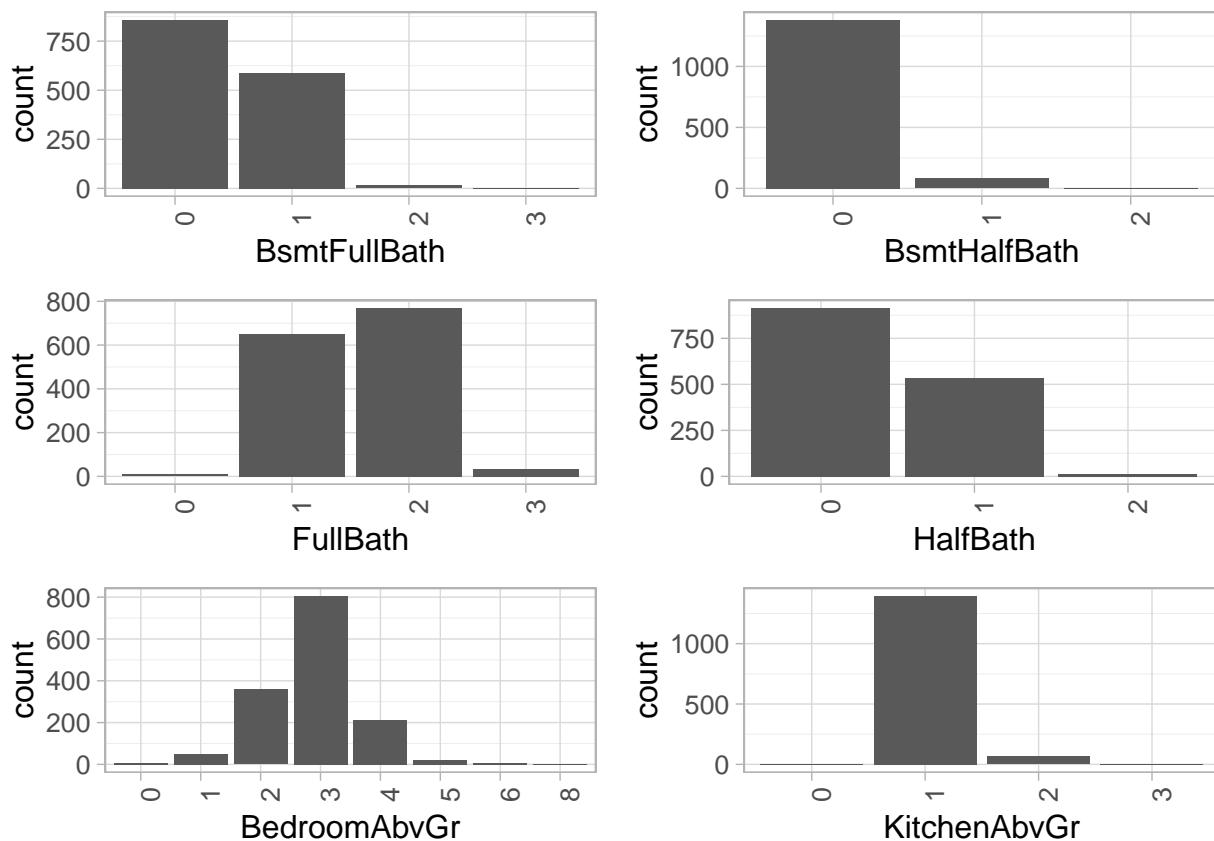


```
doPlots(train_cont, fun = plotDen, ii = 13:17, ncol = 2)
```



The histograms below show that majority of the houses have 2 full baths, 0 half baths, and have an average of 3 bedrooms.

```
doPlots(train_cont, fun = plotHist, ii = 18:23, ncol = 2)
```

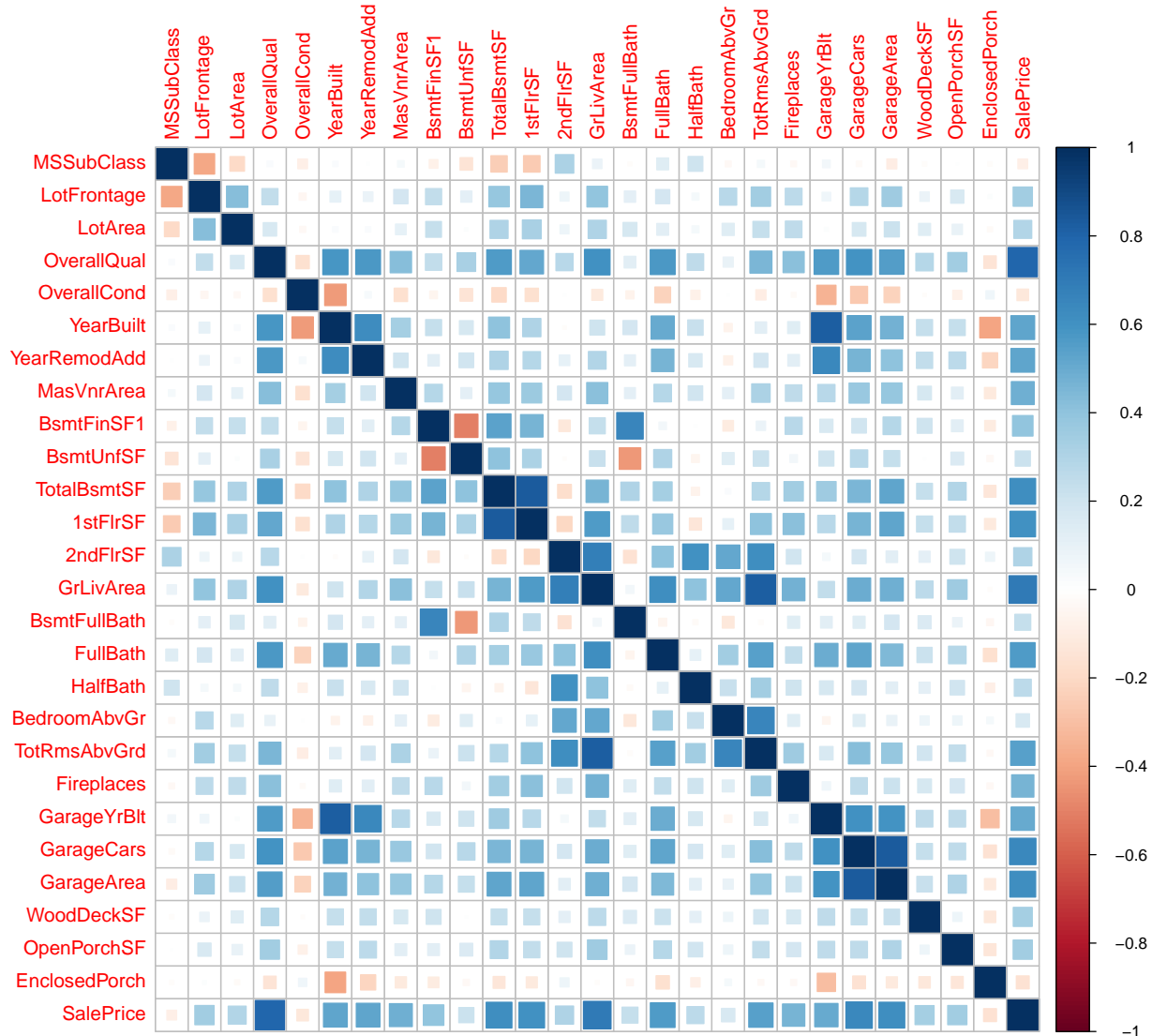


Explore the correlation

```
correlations <- cor(na.omit(train_cont[,-1, with = FALSE]))

# correlations
row_indic <- apply(correlations, 1, function(x) sum(x > 0.3 | x < -0.3) > 1)

correlations<- correlations[row_indic ,row_indic ]
corrplot(correlations, method="square")
```



Plot scatter plot for variables that have high correlation.

The correlation matrix below shows that there are several variables that are strongly and positively correlated with housing price.

High positive correlation:

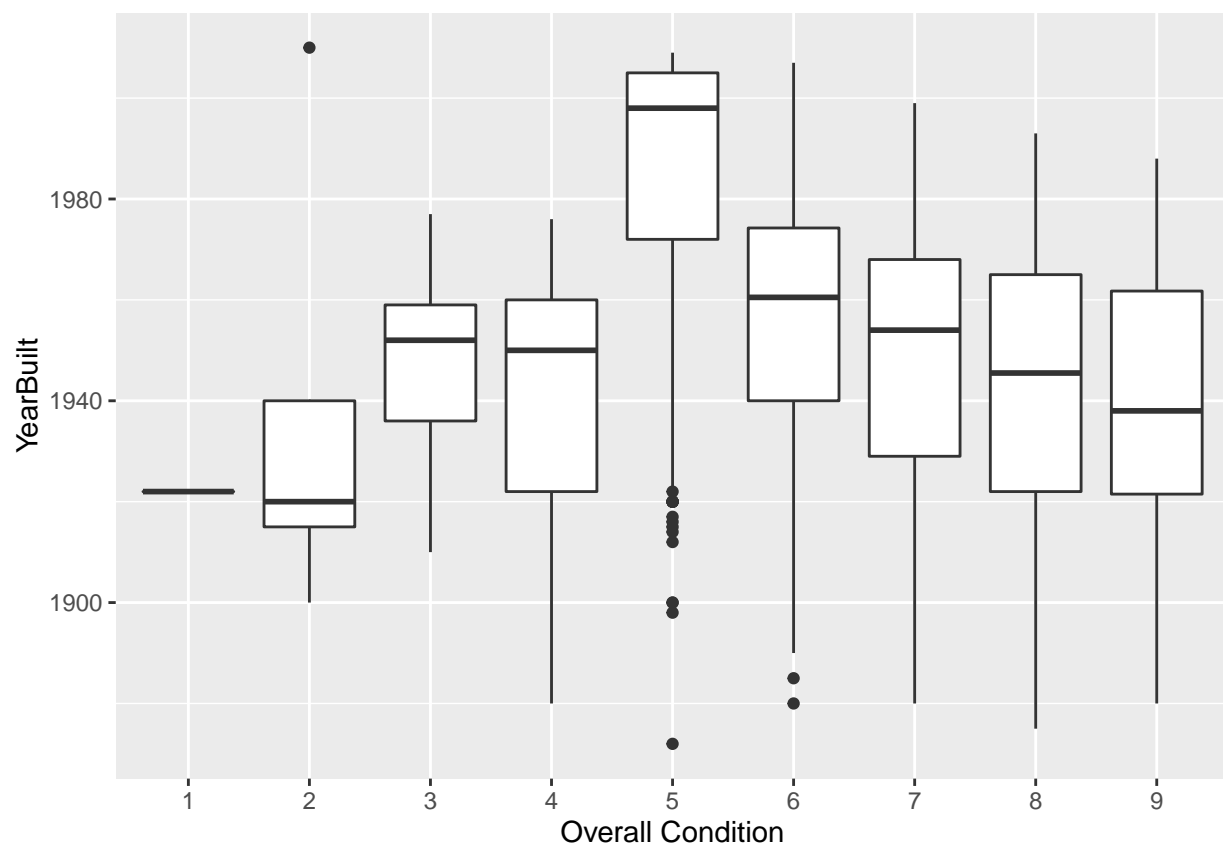
- OverallQual
- YearBuilt
- YearRemodAdd
- MasvnrArea
- BsmtFinSF1
- TotalBsmtSF

- 1stFlrSF
- GrLiveArea
- FullBath
- TotRmsAbvGrd
- FirePlaces
- GarageYrBlt
- GarageCars
- GarageArea
- WoodDeckSF
- OpenPorchSF

The number of enclosed porches are negatively correlated with year built. It seems that potential housebuyers do not want an enclosed porch and house developers have been building less enclosed porches in recent years. It is also negatively correlated with SalePrice, which makes sense.

There is some slight negative correlation between OverallCond and SalePrice. There is also strong negative correlation between Yearbuilt and OverallCond. It seems to be that recently built houses tend to be in worse Overall Condition.

```
train %>% select(OverallCond, YearBuilt) %>% ggplot(aes(factor(OverallCond), YearBuilt)) + geom_boxplot()
```

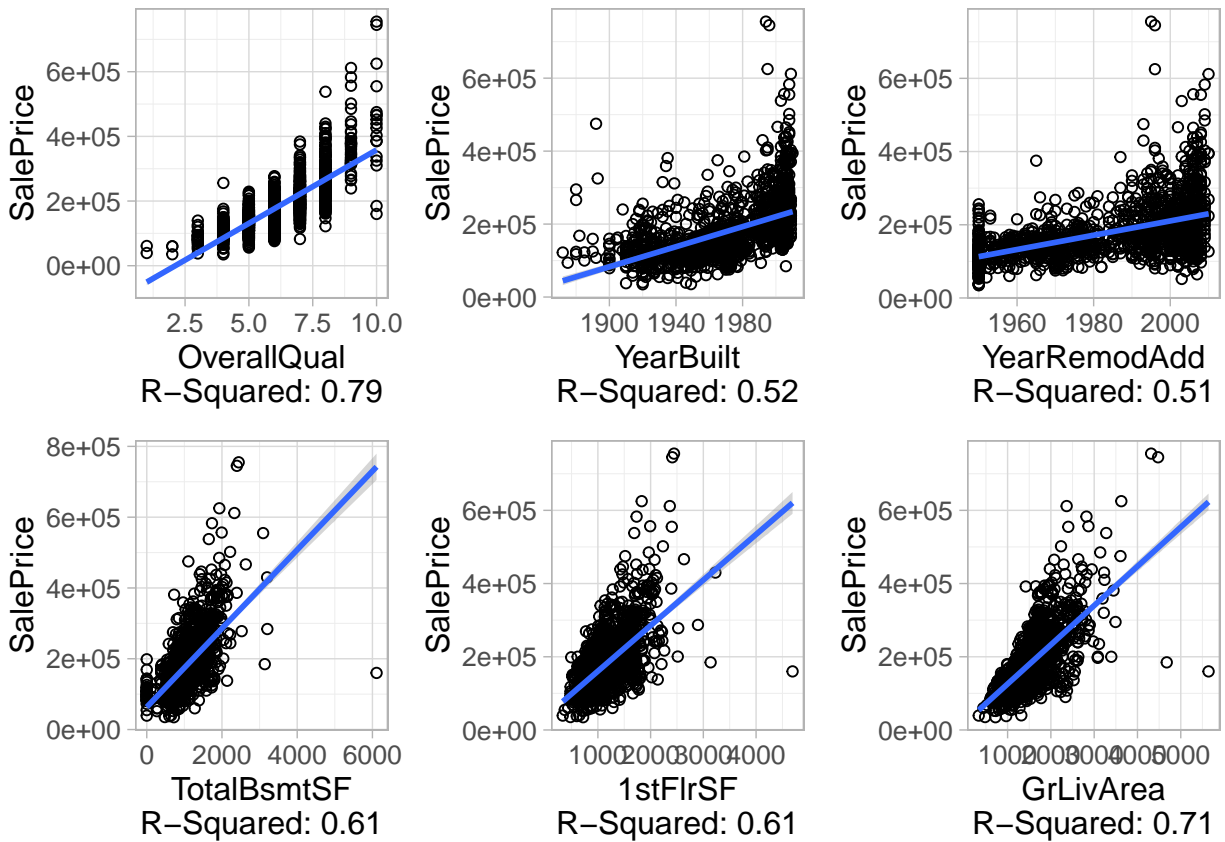


```
plotCorr <- function(data_in, i){
  data <- data.frame(x = data_in[[i]], SalePrice = data_in$SalePrice)
  p <- ggplot(data, aes(x = x, y = SalePrice)) + geom_point(shape = 1, na.rm = TRUE) + geom_smooth(method = "lm")
  return(suppressWarnings(p))
}
```



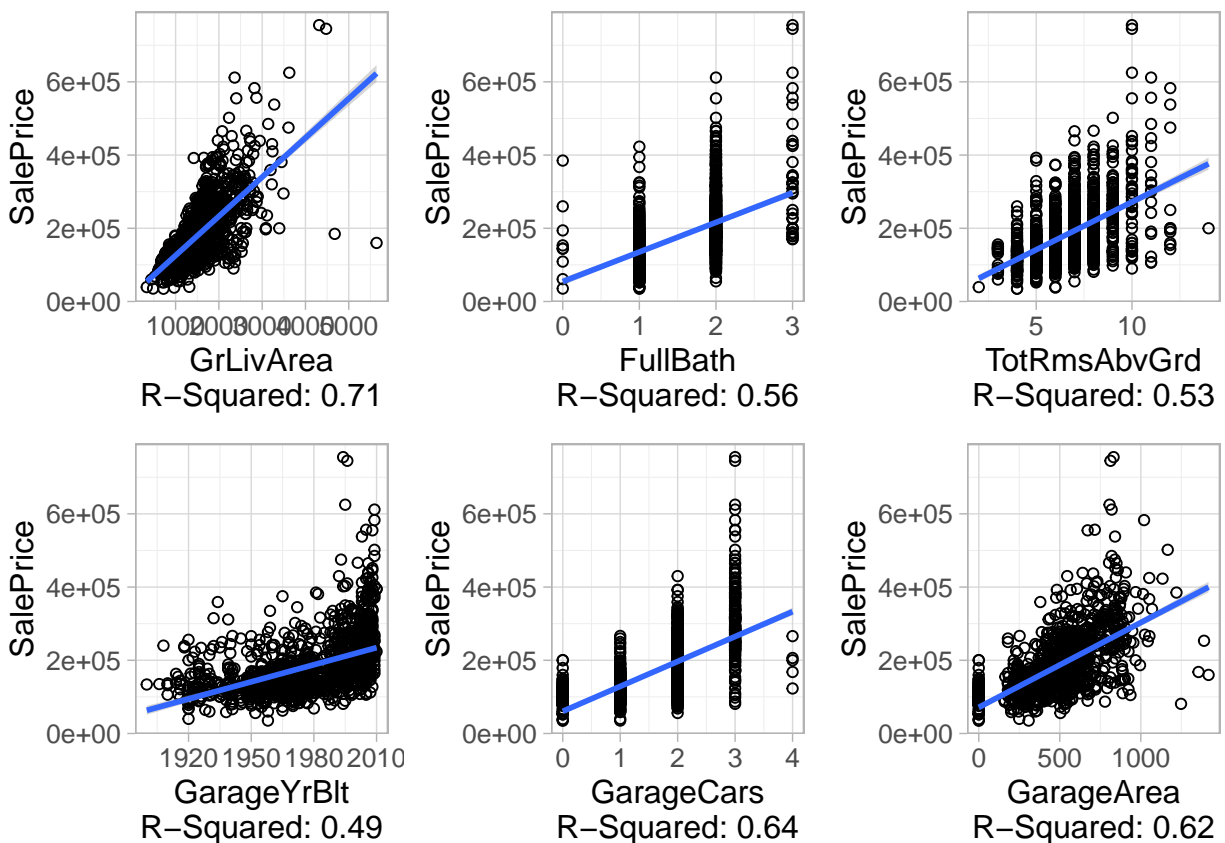
```
highcorr <- c(names(correlations[, 'SalePrice'])[which(correlations[, 'SalePrice'] > 0.5)], names(correlations[, 'SalePrice'])[which(correlations[, 'SalePrice'] < -0.5)])
data_corr <- train[, highcorr, with = FALSE]

doPlots(data_corr, fun = plotCorr, ii = 1:6)
```



```
doPlots(data_corr, fun = plotCorr, ii = 6:11)
```

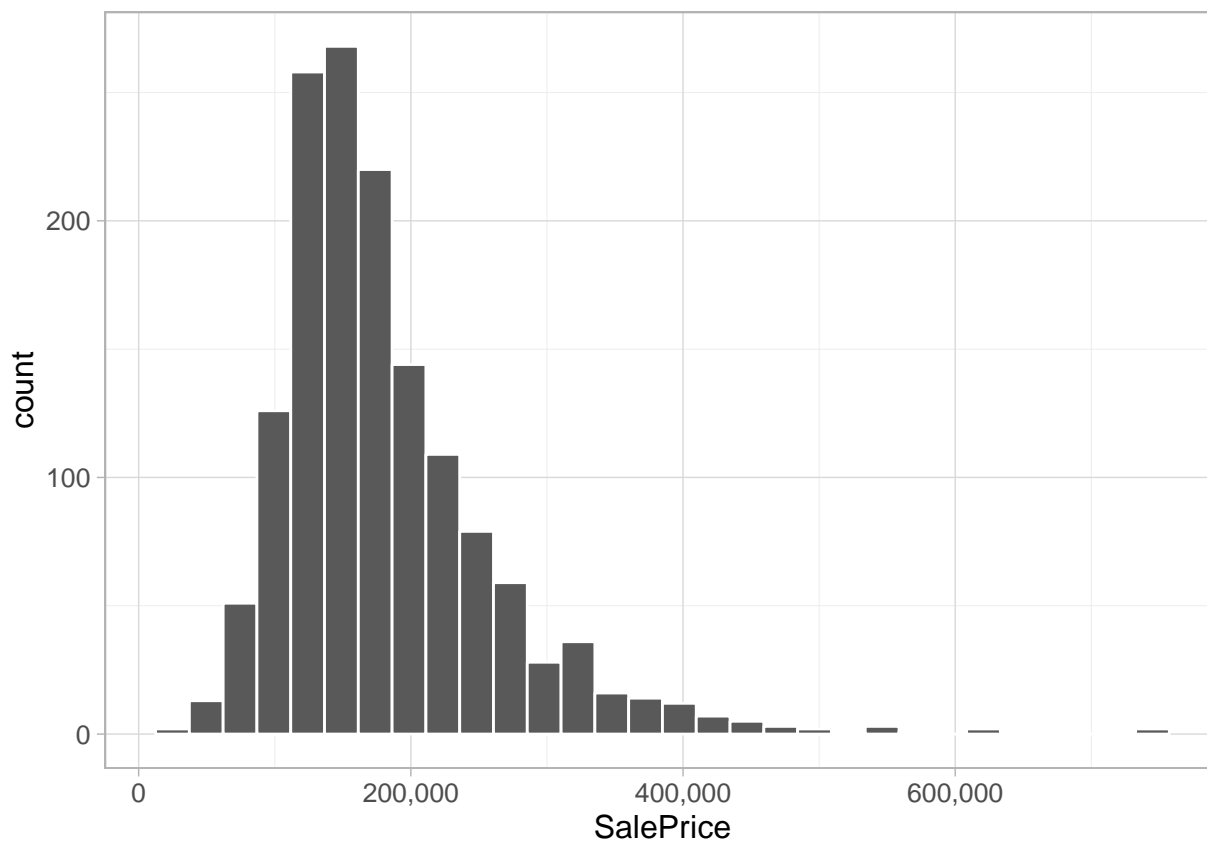
```
## Warning: Removed 81 rows containing non-finite values (stat_smooth).
```



The histogram for the response variable SalePrice shows that it is skewed. Taking the log of the variable normalizes it.

```
library(scales)
ggplot(train, aes(x=SalePrice)) + geom_histogram(col = 'white') + theme_light() + scale_x_continuous(lab=

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



```
summary(train[,.(SalePrice)])
```

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

```
#Normalize distribution
ggplot(train, aes(x=log(SalePrice+1))) + geom_histogram(col = 'white') + theme_light()
```

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

