

House price

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Understanding the Business Question

Price Recommendation for house Data inspection

Dataset Description

File descriptions

data_description.txt - full description of each column, originally prepared by Dean De Cock

but lightly edited to match the column names used here

sample_submission.csv - a benchmark submission from a linear regression on year and month of sale, lot square footage, and number of bedrooms

Data fields

Here's a brief version of what you'll find in the data description file.

SalePrice : the property's sale price in dollars. This is the target variable that you're trying to predict.

MSSubClass: The building class

MSZoning: The general zoning classification

LotFrontage: Linear feet of street connected to property

LotArea: Lot size in square feet

Street: Type of road access

Alley: Type of alley access

LotShape: General shape of property

LandContour: Flatness of the property

Utilities: Type of utilities available

LotConfig: Lot configuration

LandSlope: Slope of property

Neighborhood: Physical locations within Ames city limits

Condition1: Proximity to main road or railroad

Condition2: Proximity to main road or railroad (if a second is present)

BldgType: Type of dwelling

HouseStyle: Style of dwelling

OverallQual: Overall material and finish quality

OverallCond: Overall condition rating

YearBuilt: Original construction date

YearRemodAdd: Remodel date

RoofStyle: Type of roof

RoofMatl: Roof material

Exterior1st: Exterior covering on house

Exterior2nd: Exterior covering on house (if more than one material)

MasVnrType: Masonry veneer type

MasVnrArea: Masonry veneer area in square feet

ExterQual: Exterior material quality

ExterCond: Present condition of the material on the exterior

Foundation: Type of foundation

BsmtQual: Height of the basement

BsmtCond: General condition of the basement

BsmtExposure: Walkout or garden level basement walls

BsmtFinType1: Quality of basement finished area

BsmtFinSF1: Type 1 finished square feet

BsmtFinType2: Quality of second finished area (if present)

BsmtFinSF2: Type 2 finished square feet

BsmtUnfSF: Unfinished square feet of basement area

TotalBsmtSF: Total square feet of basement area

Heating: Type of heating

HeatingQC: Heating quality and condition

CentralAir: Central air conditioning

Electrical: Electrical system

1stFlrSF: First Floor square feet

2ndFlrSF: Second floor square feet

LowQualFinSF: Low quality finished square feet (all floors)

GrLivArea: Above grade (ground) living area square feet

BsmtFullBath: Basement full bathrooms

BsmtHalfBath: Basement half bathrooms

FullBath: Full bathrooms above grade

HalfBath: Half baths above grade

Bedroom: Number of bedrooms above basement level

Kitchen: Number of kitchens

KitchenQual: Kitchen quality

TotRmsAbvGrd: Total rooms above grade (does not include bathrooms)

Functional: Home functionality rating

Fireplaces: Number of fireplaces

FireplaceQu: Fireplace quality

GarageType: Garage location

GarageYrBlt: Year garage was built

GarageFinish: Interior finish of the garage

GarageCars: Size of garage in car capacity

GarageArea: Size of garage in square feet

GarageQual: Garage quality

GarageCond: Garage condition

PavedDrive: Paved driveway

WoodDeckSF: Wood deck area in square feet

OpenPorchSF: Open porch area in square feet

EnclosedPorch: Enclosed porch area in square feet

3SsnPorch: Three season porch area in square feet

ScreenPorch: Screen porch area in square feet

PoolArea: Pool area in square feet

PoolQC: Pool quality

Fence: Fence quality

MiscFeature: Miscellaneous feature not covered in other categories

MiscVal: \$Value of miscellaneous feature

MoSold: Month Sold

YrSold: Year Sold

SaleType: Type of sale

SaleCondition: Condition of sale

read data from file

```
train <- read.csv("train.csv" , header = TRUE )
```

A summary of the data

```
dim(train)
```

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

```
length(unique(train$Id))
```

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

```
summary(train)
```

```

##      Id      MSSubClass      MSZoning      LotFrontage
## Min.   :   1.0   Min.    : 20.0   Length:1460   Min.    : 21.00
## 1st Qu.: 365.8   1st Qu.: 20.0   Class :character   1st Qu.: 59.00
## Median : 730.5   Median : 50.0   Mode  :character   Median : 69.00
## Mean   : 730.5   Mean    : 56.9                      Mean   : 70.05
## 3rd Qu.:1095.2   3rd Qu.: 70.0                      3rd Qu.: 80.00
## Max.   :1460.0   Max.    :190.0                      Max.    :313.00
##                                     NA's    :259
##      LotArea      Street      Alley      LotShape
## Min.    : 1300   Length:1460   Length:1460   Length:1460
## 1st Qu.: 7554   Class :character   Class :character   Class :character
## Median : 9478   Mode  :character   Mode  :character   Mode  :character
## Mean    :10517
## 3rd Qu.:11602
## Max.    :215245
##
##      LandContour      Utilities      LotConfig      LandSlope
## Length:1460   Length:1460   Length:1460   Length:1460
## Class :character   Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character   Mode  :character
##
##
##
##      Neighborhood      Condition1      Condition2      BldgType
## Length:1460   Length:1460   Length:1460   Length:1460
## Class :character   Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character   Mode  :character
##
##
##
##      HouseStyle      OverallQual      OverallCond      YearBuilt
## Length:1460   Min.    : 1.000   Min.    :1.000   Min.    :1872
## Class :character   1st Qu.: 5.000   1st Qu.:5.000   1st Qu.:1954
## Mode  :character   Median : 6.000   Median :5.000   Median :1973
##                                     Mean   : 6.099   Mean   :5.575   Mean   :1971
##                                     3rd Qu.: 7.000   3rd Qu.:6.000   3rd Qu.:2000
##                                     Max.    :10.000   Max.    :9.000   Max.    :2010
##
##      YearRemodAdd      RoofStyle      RoofMatl      Exterior1st
## Min.    :1950   Length:1460   Length:1460   Length:1460
## 1st Qu.:1967   Class :character   Class :character   Class :character
## Median :1994   Mode  :character   Mode  :character   Mode  :character
## Mean    :1985
## 3rd Qu.:2004
## Max.    :2010
##
##      Exterior2nd      MasVnrType      MasVnrArea      ExterQual
## Length:1460   Length:1460   Min.    : 0.0   Length:1460
## Class :character   Class :character   1st Qu.: 0.0   Class :character
## Mode  :character   Mode  :character   Median : 0.0   Mode  :character
##                                     Mean   :103.7
##                                     3rd Qu.:166.0
##                                     Max.    :1600.0
##                                     NA's    :8
##      ExterCond      Foundation      BsmtQual      BsmtCond
## Length:1460   Length:1460   Length:1460   Length:1460
## Class :character   Class :character   Class :character   Class :character
## Mode  :character   Mode  :character   Mode  :character   Mode  :character
##
##
##
##      BsmtExposure      BsmtFinType1      BsmtFinSF1      BsmtFinType2
## Length:1460   Length:1460   Min.    : 0.0   Length:1460
## Class :character   Class :character   1st Qu.: 0.0   Class :character
## Mode  :character   Mode  :character   Median :383.5   Mode  :character
##                                     Mean   :443.6
##                                     3rd Qu.:712.2
##                                     Max.    :5644.0
##
##      BsmtFinSF2      BsmtUnfSF      TotalBsmtSF      Heating
## Min.    : 0.00   Min.    : 0.0   Min.    : 0.0   Length:1460
## 1st Qu.: 0.00   1st Qu.:223.0   1st Qu.:795.8   Class :character
## Median : 0.00   Median :477.5   Median :991.5   Mode  :character
## Mean    :46.55   Mean    :567.2   Mean :1057.4
## 3rd Qu.: 0.00   3rd Qu.:808.0   3rd Qu.:1298.2

```

```

## Max. :1474.00 Max. :2336.0 Max. :6110.0
##
## HeatingQC CentralAir Electrical X1stFlrSF
## Length:1460 Length:1460 Length:1460 Min. : 334
## Class :character Class :character Class :character 1st Qu.: 882
## Mode :character Mode :character Mode :character Median :1087
## Mean :1163
## 3rd Qu.:1391
## Max. :4692
##
## X2ndFlrSF LowQualFinSF GrLivArea BsmtFullBath
## Min. : 0 Min. : 0.000 Min. : 334 Min. :0.0000
## 1st Qu.: 0 1st Qu.: 0.000 1st Qu.:1130 1st Qu.:0.0000
## Median : 0 Median : 0.000 Median :1464 Median :0.0000
## Mean : 347 Mean : 5.845 Mean :1515 Mean :0.4253
## 3rd Qu.: 728 3rd Qu.: 0.000 3rd Qu.:1777 3rd Qu.:1.0000
## Max. :2065 Max. :572.000 Max. :5642 Max. :3.0000
##
## BsmtHalfBath FullBath HalfBath BedroomAbvGr
## Min. :0.00000 Min. :0.000 Min. :0.0000 Min. :0.000
## 1st Qu.:0.00000 1st Qu.:1.000 1st Qu.:0.0000 1st Qu.:2.000
## Median :0.00000 Median :2.000 Median :0.0000 Median :3.000
## Mean :0.05753 Mean :1.565 Mean :0.3829 Mean :2.866
## 3rd Qu.:0.00000 3rd Qu.:2.000 3rd Qu.:1.0000 3rd Qu.:3.000
## Max. :2.00000 Max. :3.000 Max. :2.0000 Max. :8.000
##
## KitchenAbvGr KitchenQual TotRmsAbvGrd Functional
## Min. :0.000 Length:1460 Min. : 2.000 Length:1460
## 1st Qu.:1.000 Class :character 1st Qu.: 5.000 Class :character
## Median :1.000 Mode :character Median : 6.000 Mode :character
## Mean :1.047 Mean : 6.518
## 3rd Qu.:1.000 3rd Qu.: 7.000
## Max. :3.000 Max. :14.000
##
## Fireplaces FireplaceQu GarageType GarageYrBlt
## Min. :0.000 Length:1460 Length:1460 Min. :1900
## 1st Qu.:0.000 Class :character Class :character 1st Qu.:1961
## Median :1.000 Mode :character Mode :character Median :1980
## Mean :0.613 Mean :1979
## 3rd Qu.:1.000 3rd Qu.:2002
## Max. :3.000 Max. :2010
## NA's :81
##
## GarageFinish GarageCars GarageArea GarageQual
## Length:1460 Min. :0.000 Min. : 0.0 Length:1460
## Class :character 1st Qu.:1.000 1st Qu.: 334.5 Class :character
## Mode :character Median :2.000 Median : 480.0 Mode :character
## Mean :1.767 Mean : 473.0
## 3rd Qu.:2.000 3rd Qu.: 576.0
## Max. :4.000 Max. :1418.0
##
## GarageCond PavedDrive WoodDeckSF OpenPorchSF
## Length:1460 Length:1460 Min. : 0.00 Min. : 0.00
## Class :character Class :character 1st Qu.: 0.00 1st Qu.: 0.00
## Mode :character Mode :character Median : 0.00 Median : 25.00
## Mean : 94.24 Mean : 46.66
## 3rd Qu.:168.00 3rd Qu.: 68.00
## Max. :857.00 Max. :547.00
##
## EnclosedPorch X3SsnPorch ScreenPorch PoolArea
## Min. : 0.00 Min. : 0.00 Min. : 0.00 Min. : 0.000
## 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.00 1st Qu.: 0.000
## Median : 0.00 Median : 0.00 Median : 0.00 Median : 0.000
## Mean : 21.95 Mean : 3.41 Mean : 15.06 Mean : 2.759
## 3rd Qu.: 0.00 3rd Qu.: 0.00 3rd Qu.: 0.00 3rd Qu.: 0.000
## Max. :552.00 Max. :508.00 Max. :480.00 Max. :738.000
##
## PoolQC Fence MiscFeature MiscVal
## Length:1460 Length:1460 Length:1460 Min. : 0.00
## Class :character Class :character Class :character 1st Qu.: 0.00
## Mode :character Mode :character Mode :character Median : 0.00
## Mean : 43.49
## 3rd Qu.: 0.00
## Max. :15500.00
##
## MoSold YrSold SaleType SaleCondition
## Min. : 1.000 Min. :2006 Length:1460 Length:1460
## 1st Qu.: 5.000 1st Qu.:2007 Class :character Class :character
## Median : 6.000 Median :2008 Mode :character Mode :character
## Mean : 6.322 Mean :2008

```

```
## 3rd Qu.: 8.000 3rd Qu.:2009
## Max. :12.000 Max. :2010
##
## SalePrice
## Min. : 34900
## 1st Qu.:129975
## Median :163000
## Mean :180921
## 3rd Qu.:214000
## Max. :755000
##
```

Convert categorical variables to factor

```
cat_var <- c("MSZoning" , "Street" , "Alley" , "LotShape" , "LandContour", "Utilities" , "LotConfig" , "LandSlope" ,
"Neighborhood" , "Condition1" , "Condition2" , "BldgType" ,
"HouseStyle", "OverallQual", "OverallCond" , "RoofStyle" , "Exterior1st" , "Exterior2nd" , "MasVnrType",
"ExterQual" ,"ExterCond" , "Foundation" , "BsmtQual" ,"BsmtCond" ,
"BsmtExposure" ,"BsmtFinType1" ,"BsmtFinType2" , "Heating" , "HeatingQC" , "CentralAir" , "Electrical" ,
"KitchenQual" , "Functional" , "Fireplaces" , "FireplaceQu" ,
"GarageType" , "GarageFinish" , "GarageCars" , "GarageQual" ,"GarageCond" , "PavedDrive" , "PoolQC",
"Fence","RoofMatl" ,
"MiscFeature" , "SaleType" , "SaleCondition")
```

```
train[,cat_var] <- lapply(train[,cat_var] , factor)
knitr::kable(summary(train))
```

Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley	LotShape	LandContour	Utilities	LotConfig	LandSlope	Nei
Min. : 1.0	Min. : 20.0	C (all): 10	Min. : 21.00	Min. : 1300	Grvl: 6	Grvl: 50	IR1:484	Bnk: 63	AllPub:1459	Corner : 263	Gtl:1382	NA
1st Qu.: 365.8	1st Qu.: 20.0	FV : 65	1st Qu.: 59.00	1st Qu.: 7554	Pave:1454	Pave: 41	IR2: 41	HLS: 50	NoSeWa: 1	CulDSac: 94	Mod: 65	Col
Median : 730.5	Median : 50.0	RH : 16	Median : 69.00	Median : 9478	NA	NA's:1369	IR3: 10	Low: 36	NA	FR2 : 47	Sev: 13	Old
Mean : 730.5	Mean : 56.9	RL :1151	Mean : 70.05	Mean : 10517	NA	NA	Reg:925	Lvl:1311	NA	FR3 : 4	NA	Edh
3rd Qu.:1095.2	3rd Qu.: 70.0	RM : 218	3rd Qu.: 80.00	3rd Qu.: 11602	NA	NA	NA	NA	NA	Inside :1052	NA	Sor
Max. :1460.0	Max. :190.0	NA	Max. :313.00	Max. :215245	NA	NA	NA	NA	NA	NA	NA	Gill
NA	NA	NA	NA's :259	NA	NA	NA	NA	NA	NA	NA	NA	(Ot

```
knitr::kable(str(train))
```

```

## '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 "Grv1","Pave": 2 2 2 2 2 2 2 2 2 2 ...
## $ Alley          : Factor w/ 2 levels "Grv1","Pave": NA NA NA NA NA NA NA NA NA ...
## $ LotShape       : Factor w/ 4 levels "IR1","IR2","IR3",...: 4 4 1 1 1 1 1 4 1 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    : Factor w/ 10 levels "1","2","3","4",...: 7 6 7 7 8 5 8 7 7 5 ...
## $ OverallCond    : Factor w/ 9 levels "1","2","3","4",...: 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 6 2 6 ...
## $ BsmtFinSF2     : int  0 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 2 5 ...
## $ 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 7 3 7 ...
## $ Fireplaces     : Factor w/ 4 levels "0","1","2","3": 1 2 2 2 2 1 2 3 3 3 ...
## $ 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     : Factor w/ 5 levels "0","1","2","3",...: 3 3 3 4 4 3 3 3 3 2 ...
## $ 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 ...
## $ Fence          : Factor w/ 4 levels "GdPrv","GdWo",...: NA NA NA NA NA 3 NA NA NA ...
## $ MiscFeature     : Factor w/ 4 levels "Gar2","Othr",...: NA NA NA NA NA 3 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    : 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 1 5 ...
## $ SalePrice   : int  208500 181500 223500 140000 250000 143000 307000 200000 129900 118000 ...
```

Identify missing data

```
mv_summary2      <- data.frame('variables name' = colnames(train))
mv_summary2$freq  <- apply(train , 2 , function(x) sum(is.na(x)))
mv_summary2$pers  <- round(mv_summary2$freq / nrow(train) , 3) * 100
mv_summary2_1     <- as.data.frame(mv_summary2[mv_summary2$pers > 0 & mv_summary2$pers <10, ])
mv_summary2_2     <- as.data.frame(mv_summary2[mv_summary2$pers > 10 , ])
knitr::kable(mv_summary2_1)
```

	variables.name	freq	pers
26	MasVnrType	8	0.5
27	MasVnrArea	8	0.5
31	BsmtQual	37	2.5
32	BsmtCond	37	2.5
33	BsmtExposure	38	2.6
34	BsmtFinType1	37	2.5
36	BsmtFinType2	38	2.6
43	Electrical	1	0.1
59	GarageType	81	5.5
60	GarageYrBlt	81	5.5
61	GarageFinish	81	5.5
64	GarageQual	81	5.5
65	GarageCond	81	5.5

```
knitr::kable(mv_summary2_2)
```

	variables.name	freq	pers
4	LotFrontage	259	17.7
7	Alley	1369	93.8
58	FireplaceQu	690	47.3
73	PoolQC	1453	99.5
74	Fence	1179	80.8
75	MiscFeature	1406	96.3

Removing columns that have more than 10% missing value

```
t1 <- train[,c("LotFrontage","Alley","FireplaceQu" ,"Fence" ,
               "MiscFeature" , "PoolQC")]
```

In my opinion, these columns have high missing values, so it causes problems in our modeling, so I preferred to remove them.

```
train1 <- train[,-which(train %in% t1)]
dim(train1)
```

```
## [1] 1460 75
```

I also removed all the missing values in the rows. This is a good solution to get rid of the missing values

```
train2 <- train1[apply(train1,1,function(x) any(is.na(x))) == F,]
```

For convenience, I divided the data into continuous and discrete parts


```
cat <- train2[,c("MSZoning" , "Street" , "LotShape" , "LandContour", "Utilities" , "LotConfig" , "LandSlope" ,
  "Neighborhood" , "Condition1" , "Condition2" , "BldgType" ,
  "HouseStyle", "OverallQual", "OverallCond" , "RoofStyle" , "Exterior1st" , "Exterior2nd" ,
  "MasVnrType" ,
  "ExterQual" , "ExterCond" , "Foundation" , "BsmtQual" , "BsmtCond" ,
  "BsmtExposure" , "BsmtFinType1" , "BsmtFinType2" , "Heating" , "HeatingQC" , "CentralAir" ,
  "Electrical" ,
  "KitchenQual" , "Functional" , "GarageType" , "GarageFinish" , "GarageCars" , "GarageQual" ,
  "GarageCond" , "PavedDrive" ,
  "RoofMatl" ,
  "SaleType" , "SaleCondition" , "Fireplaces" )]
```

I prefer to convert the columns that contain the date field to age, this helps me more easily determine the relationship between the price and the life of the house.

```
train3 <- train2[, -c(18,19,57,72)]
today <- as.Date("2022" , format = "%Y")
```

```
train3$ageBuilt <- as.Date(as.character(train2$YearBuilt) , format = "%Y")

train3$ageBuilt <- as.numeric(today - train3$ageBuilt)

train3$ageBuilt <- round(train3$ageBuilt/365 )
summary(train3$ageBuilt)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 12.00   21.00   46.00   48.97   66.00  142.00
```

```
train3$ageRemodAdd <- as.Date(as.character(train2$YearRemodAdd), format = "%Y")

train3$ageRemodAdd <- as.numeric(today - train3$ageRemodAdd)

train3$ageRemodAdd <- round(train3$ageRemodAdd / 365 )

summary(train3$ageRemodAdd)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 12.00   18.00   27.50   36.33   54.00   72.00
```

```
train3$GarageageBlt <- as.Date(as.character(train2$GarageYrBlt), format = "%Y")
train3$GarageageBlt <- as.numeric(today - train3$GarageageBlt)
train3$GarageageBlt <- round(train3$GarageageBlt / 365 )

summary(train3$GarageageBlt)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 12.0    20.0    42.0    43.4    60.0   122.0
```

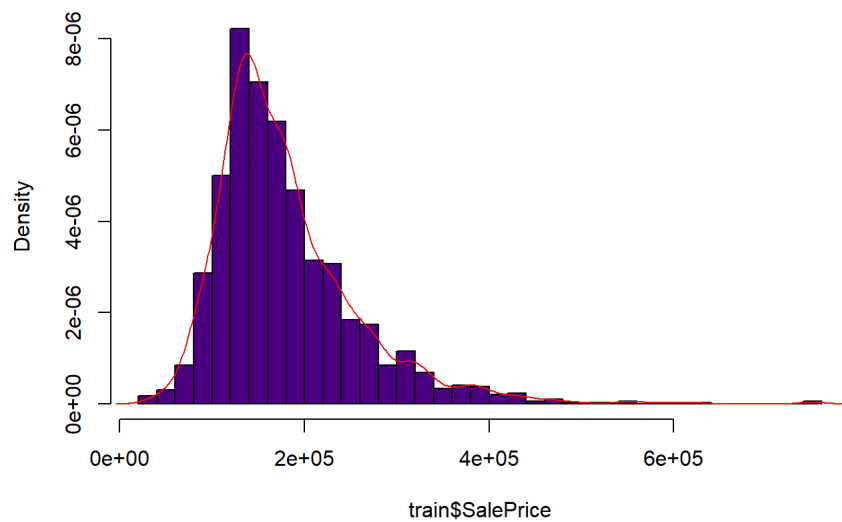
```
#Regarding the sale date, it also helps me to better recognize the rise and fall of prices
train3$ageSold <- as.Date(as.character(train2$YrSold) , format = "%Y")
train3$ageSold <- as.numeric(today - train3$ageSold)
train3$ageSold <- round(train3$ageSold / 365 )
summary(train3$ageSold)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 12.00   13.00   14.00   14.19   15.00   16.00
```

Regarding the sale date, it also helps me to better recognize the rise and fall of prices

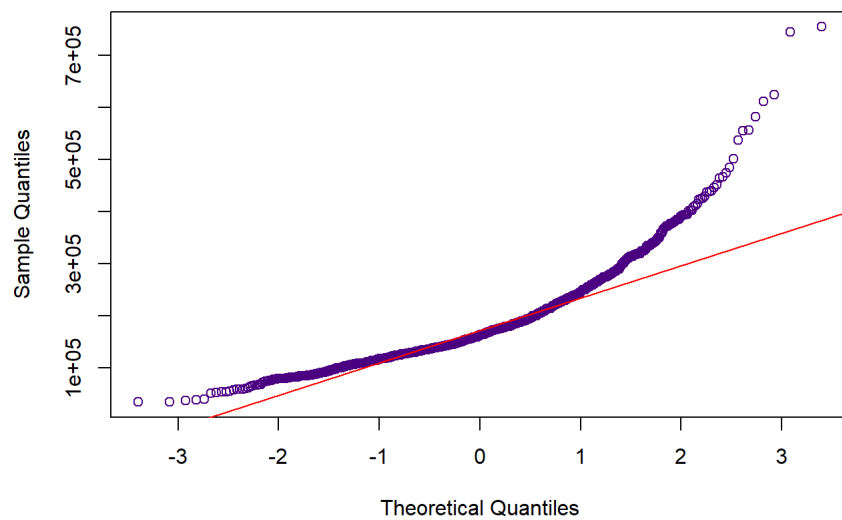
```
par(mfrow = c(1,1))
hist(train3$SalePrice , breaks = 50, probability = TRUE , col = "#4B0082")
lines(density(train3$SalePrice) , col = "red")
```

Histogram of train\$SalePrice

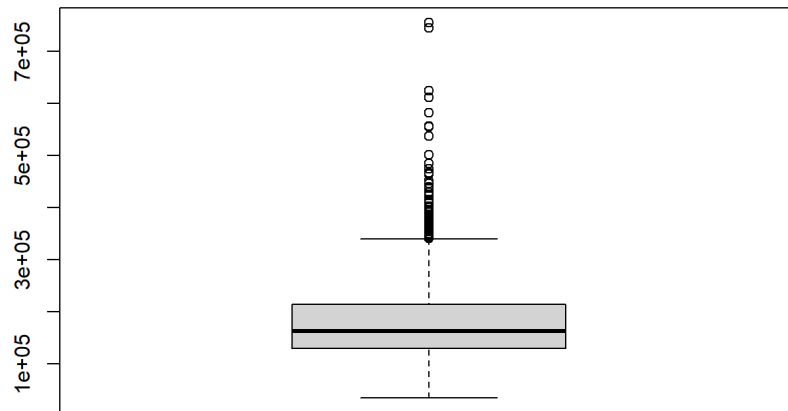


```
qqnorm(train$SalePrice , col = "#480082")  
qqline(train$SalePrice , col = "red")
```

Normal Q-Q Plot



```
par(mfrow = c(1,1))  
boxplot(train$SalePrice)
```



```
library("moments")

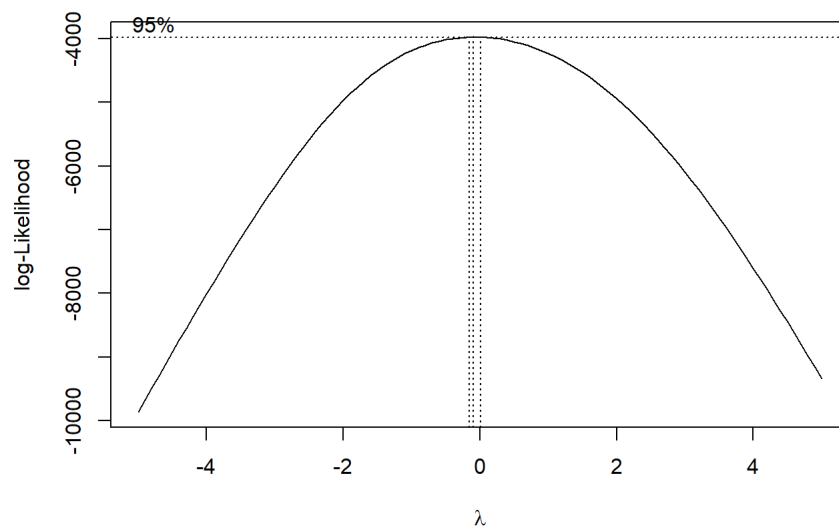
jarque.test(train$SalePrice)
```

```
##
## Jarque-Bera Normality Test
##
## data: train$SalePrice
## JB = 3438.9, p-value < 2.2e-16
## alternative hypothesis: greater
```

```
#pvalue < 0 -> h0 reject
```

The graphs and statistical tests indicate that the response variable does not follow a normal distribution, which makes sense to me because a smaller percentage of people are very wealthy. I would like to approximate them to a normal distribution, this may help me. For this purpose, I use Box-Cox conversion.

```
library("MASS")
box_result <- boxcox(train$SalePrice ~ 1, lambda = seq(-5, 5, 0.1))
```

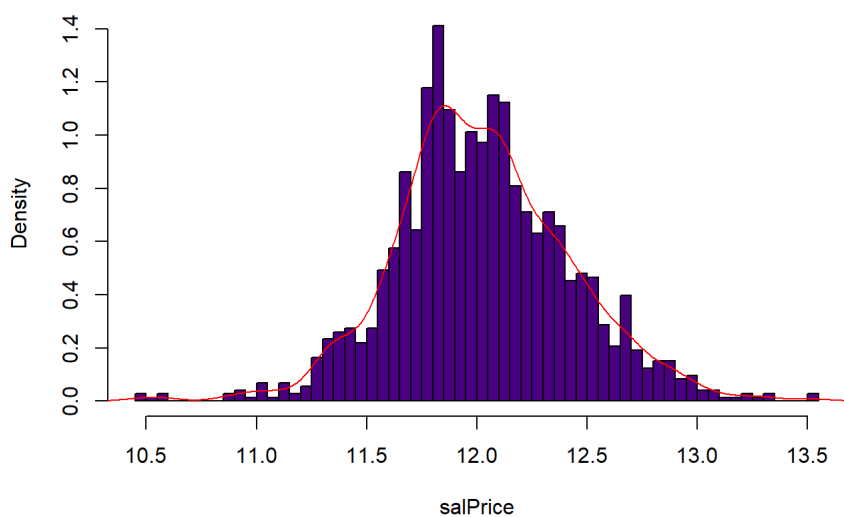


```
box_result <- data.frame(box_result)
lambda <- box_result[which(box_result$y == max(box_result$y)),]
salPrice <- log(train$SalePrice)
```

It is clear that zero is inside this confidence interval Therefore, I use logarithmic transformation

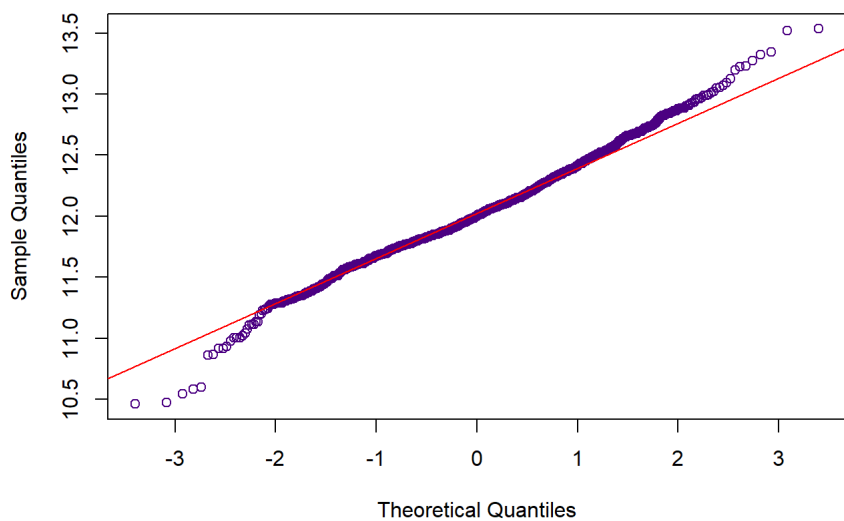
```
hist(salPrice , breaks = 50 , probability = TRUE , col = "#4B0082")
lines(density(salPrice) , col = "red")
```

Histogram of salPrice



```
qqnorm(salPrice , col = "#4B0082")
qqline(salPrice , col = "red")
```

Normal Q-Q Plot



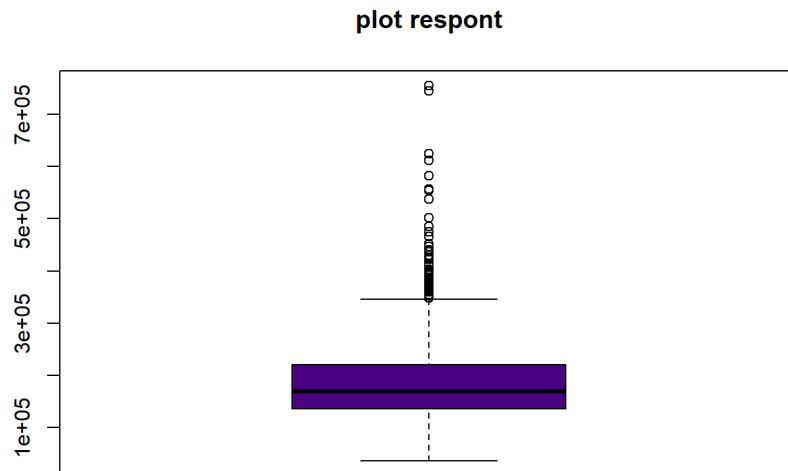
```
jarque.test(salPrice)
```

```
##
## Jarque-Bera Normality Test
##
## data: salPrice
## JB = 42.767, p-value = 5.167e-10
## alternative hypothesis: greater
```

```
anscombe.test(salPrice)
```

```
##
##  Anscombe-Glynn kurtosis test
##
## data:  salPrice
## kurt = 3.8027, z = 4.6820, p-value = 2.84e-06
## alternative hypothesis: kurtosis is not equal to 3
```

```
par(mfrow = c(1,1))
boxplot(train3$SalePrice ,main = "plot resport", col = "#4B0082")
```



It seems that the data is far from the

normal distribution

```
tukey_u <- quantile(train3$SalePrice , probs = 0.75) + 1.5 * IQR(train3$SalePrice)
sum(train3$SalePrice > tukey_u)
```

```
## [1] 56
```

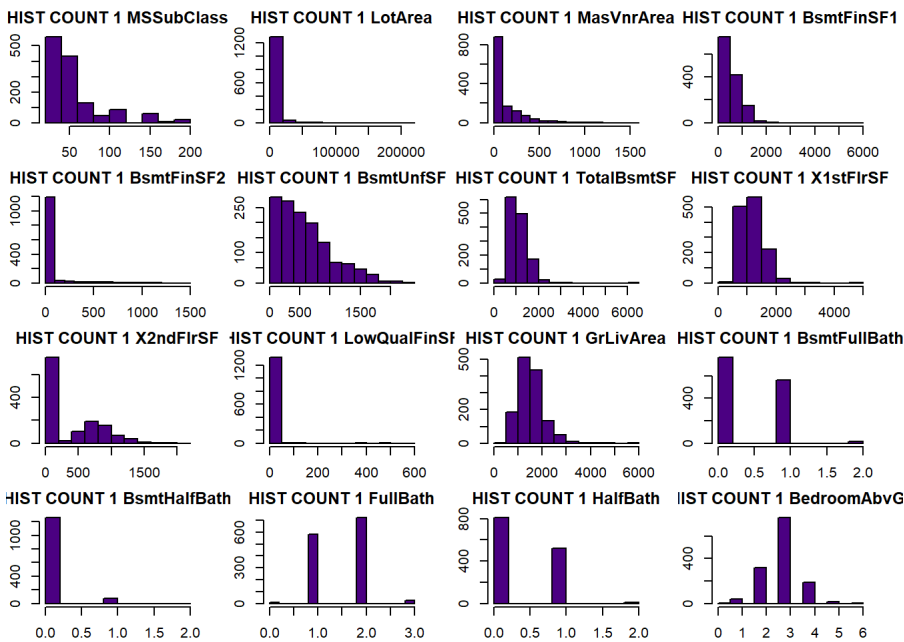
The number of continuous data is very high, so I will convert them into two parts

```
count <- train3[, -which(train3 %in% cat)]
cunt1 <- count[, c(1:17)]
cunt2 <- count[, c(18:33)]

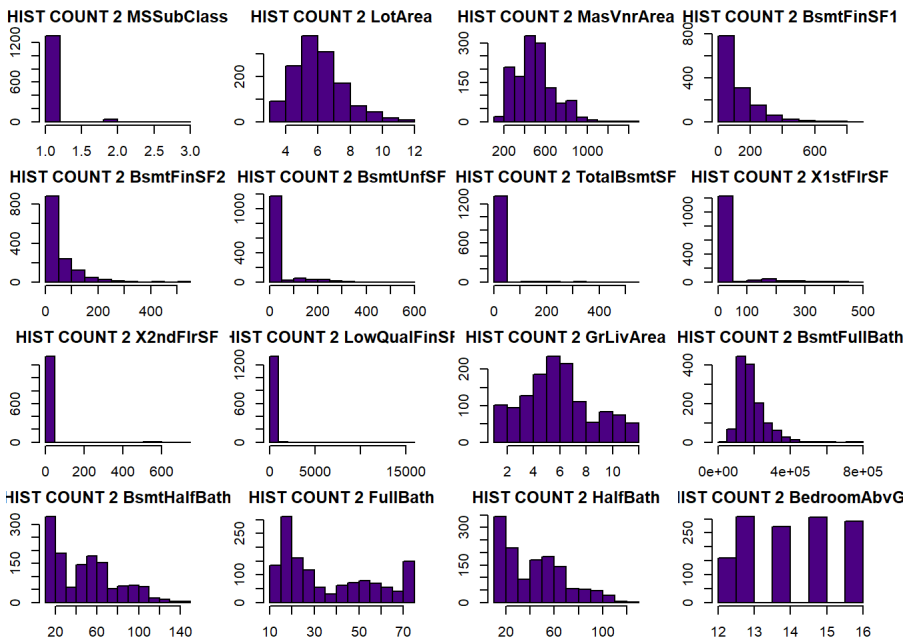
cunt1 <- cunt1[, -1]
cunt1$SalePrice <- cunt2$SalePrice
```

Histogram of continuous data

```
par(mar = c(2,2,2,2))
par(mfrow = c(4,4))
for (i in 1:16) {
  hist(cunt1[,i] , xlab = "" , col = "#4B0082", main = paste("HIST COUNT 1" ,
                                                             names(cunt1[i])))
}
```



```
par(mar = c(2,2,2,2))
par(mfrow = c(4,4))
for (i in 1:16) {
  hist(cunt2[,i] , xlab = "" , col = "#4B0082" , main = paste("HIST COUNT 2" ,
    names(cunt1[i])))
}
```



Checking the correlation of

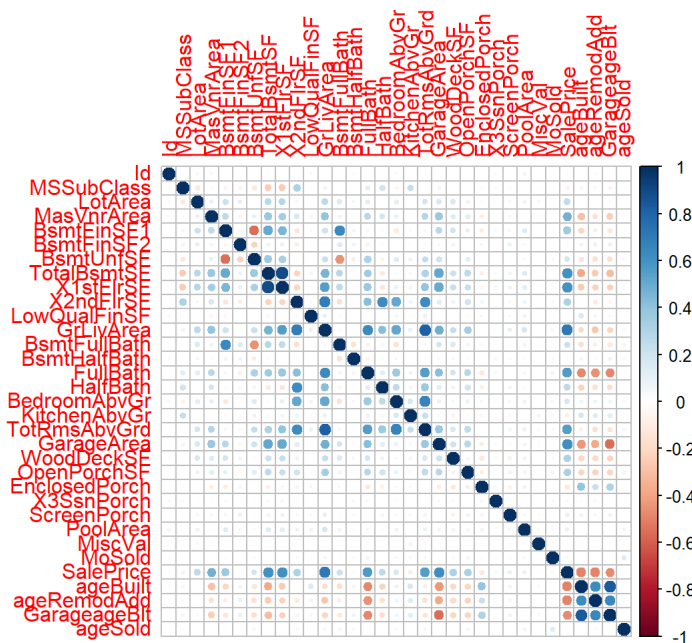
continuous data with the response variable

```
cros_tab <- round(cor(count),2)
```

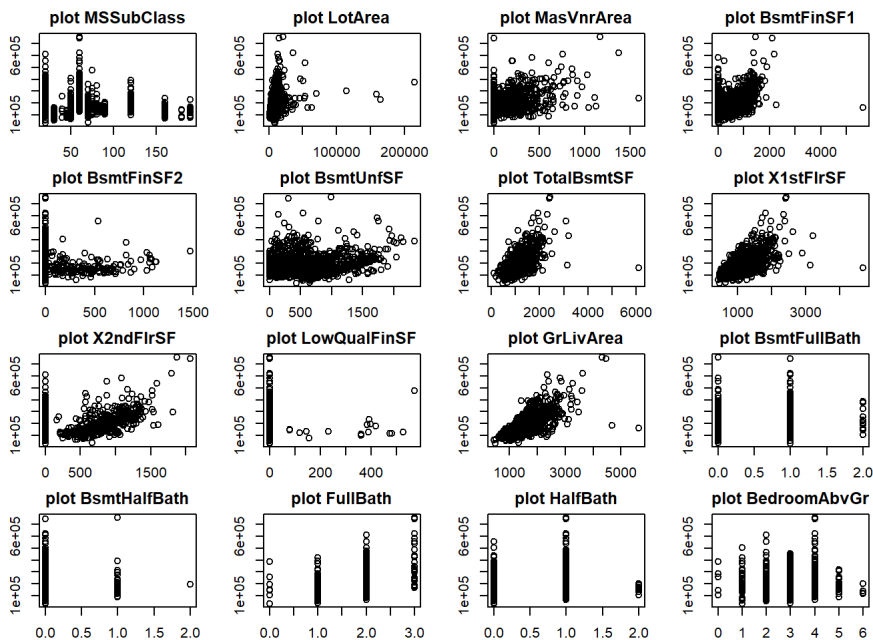
```
library("corrplot")
```

```
## corrplot 0.92 loaded
```

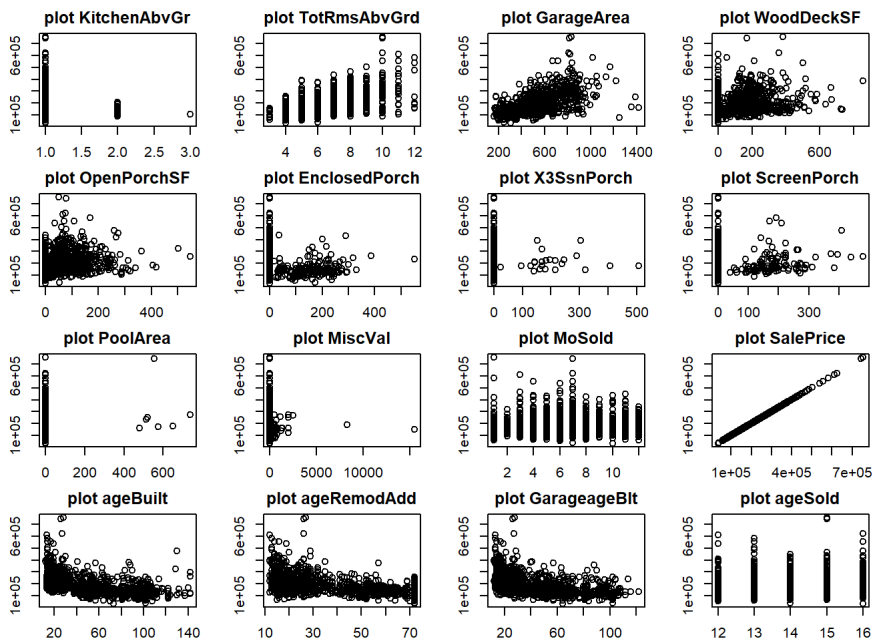
```
par(mfrow = c(1,1))
corrplot(cros_tab)
```



```
par(mar=c(2,2,2,2))
par(mfrow = c(4,4))
for (i in 1:16) {
  plot(cunt1[,i] , cunt2$SalePrice,xlab = "" , main = paste("plot" ,names(cunt1)[i] ))
}
```



```
par(mar=c(2,2,2,2))
par(mfrow = c(4,4))
for (i in 1:16) {
  plot(cunt2[,i] , cunt2$SalePrice,xlab = "" , main = paste("plot" ,names(cunt2)[i] ))
}
```



I put the columns that contain discrete variables in one file

```
categori <- train2[,which( train2 %in% cat)]
colnames(categori)
```

```
## [1] "MSZoning"      "Street"        "LotShape"      "LandContour"
## [5] "Utilities"     "LotConfig"     "LandSlope"     "Neighborhood"
## [9] "Condition1"    "Condition2"    "BldgType"      "HouseStyle"
## [13] "OverallQual"   "OverallCond"   "RoofStyle"     "RoofMatl"
## [17] "Exterior1st"   "Exterior2nd"   "MasVnrType"    "ExterQual"
## [21] "ExterCond"     "Foundation"    "BsmtQual"      "BsmtCond"
## [25] "BsmtExposure" "BsmtFinType1"  "BsmtFinType2"  "Heating"
## [29] "HeatingQC"     "CentralAir"    "Electrical"     "KitchenQual"
## [33] "Functional"    "Fireplaces"    "GarageType"     "GarageQual"
## [37] "GarageCars"    "GarageQual"    "GarageCond"     "PavedDrive"
## [41] "SaleType"      "SaleCondition"
```

```
dim(categori)
```

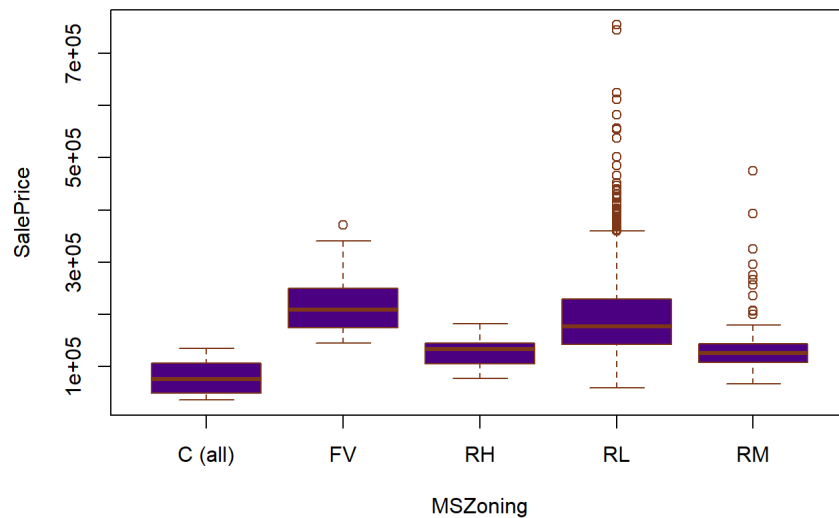
```
## [1] 1338 42
```

Examining each variable with the response variable

```
table(train3$MSZoning)
```

```
##
## C (all)   FV    RH    RL    RM
##         8    62    11   1066  191
```

```
boxplot(SalePrice ~ MSZoning , data = train3 , col = "#4B0082" , border = "#7E3817")
```

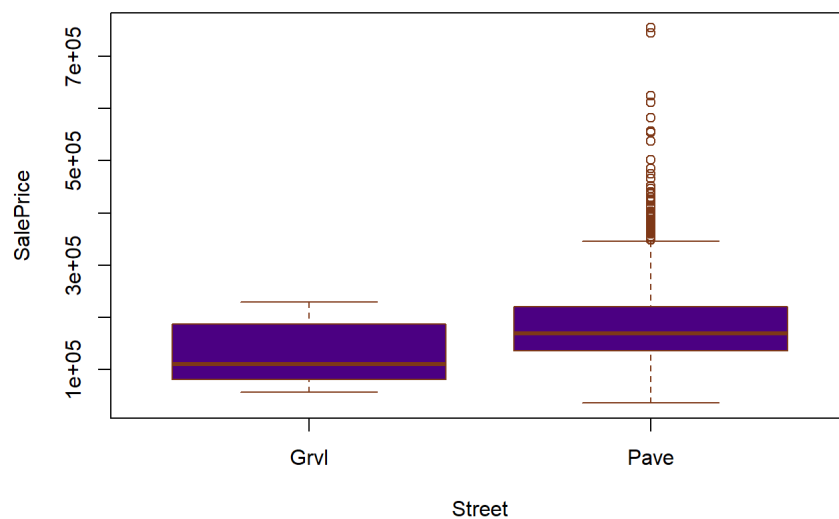



Residential Low Density seems to have the highest price

```
table(train3$Street)
```

```
##
## Grv1 Pave
## 5 1333
```

```
boxplot(SalePrice ~ Street , data = train3 , col = "#4B0082" , border = "#7E3817")
```



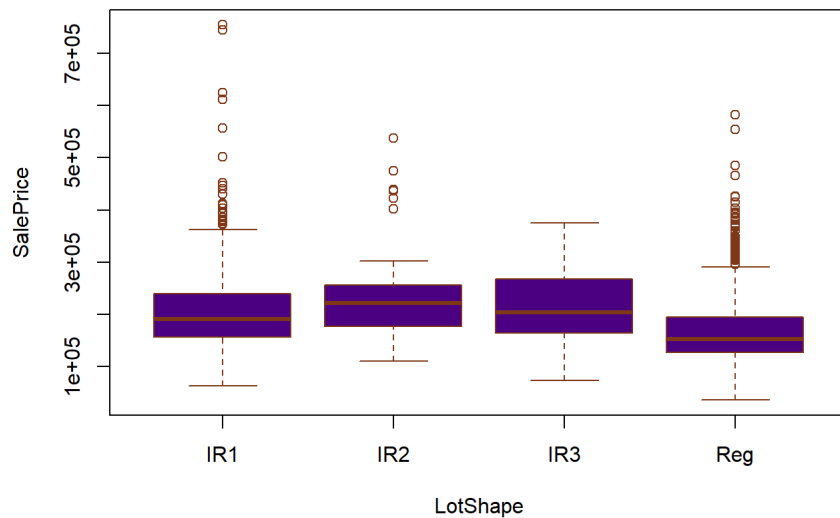
The type of Paved street has a great

effect on the price. It seems that these houses are located in a more expensive area

```
table(train3$LotShape)
```

```
##
## IR1 IR2 IR3 Reg
## 459 40 10 829
```

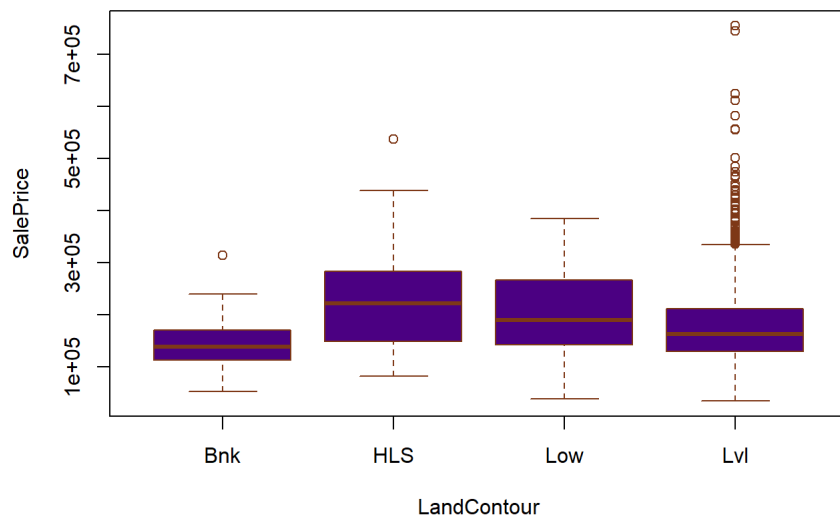
```
boxplot(SalePrice ~ LotShape , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$LandContour)
```

```
##
## Bnk HLS Low Lvl
## 52 48 32 1206
```

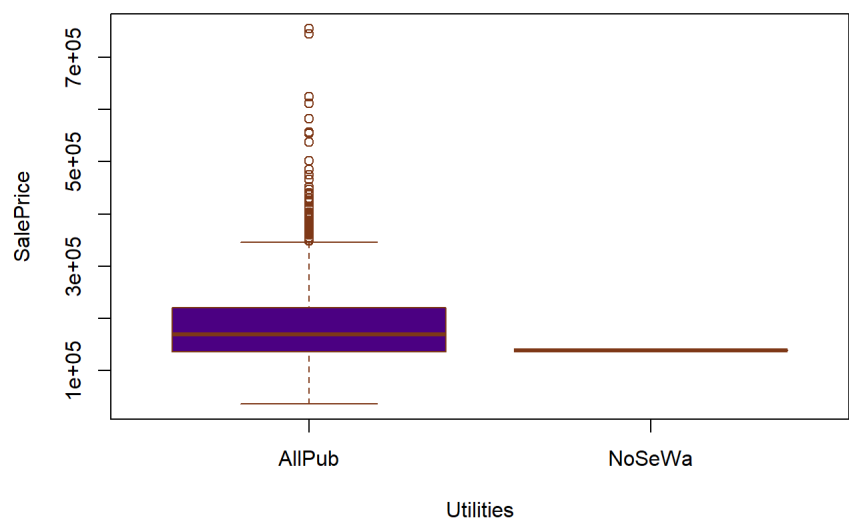
```
boxplot(SalePrice ~ LandContour , data = train , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Utilities)
```

```
##
## AllPub NoSeWa
## 1337 1
```

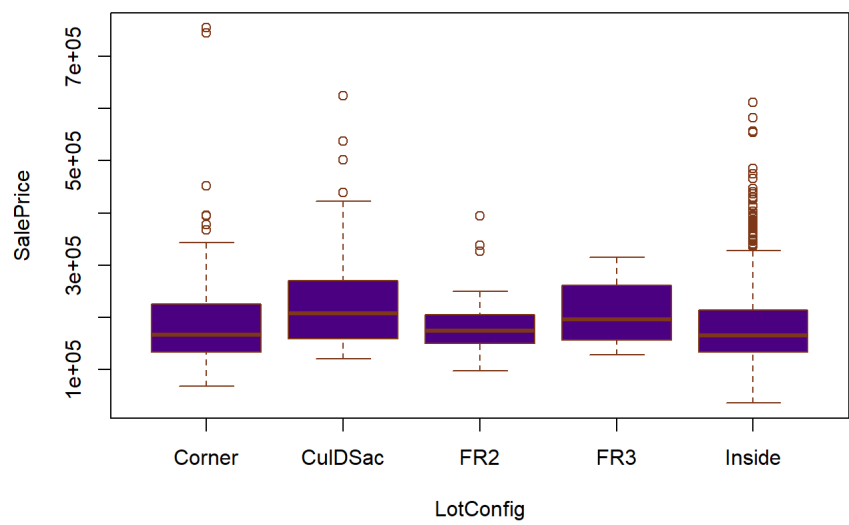
```
boxplot(SalePrice ~ Utilities , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$LotConfig)

##
##  Corner  CulDSac   FR2   FR3  Inside
##    244    90    43    4   957

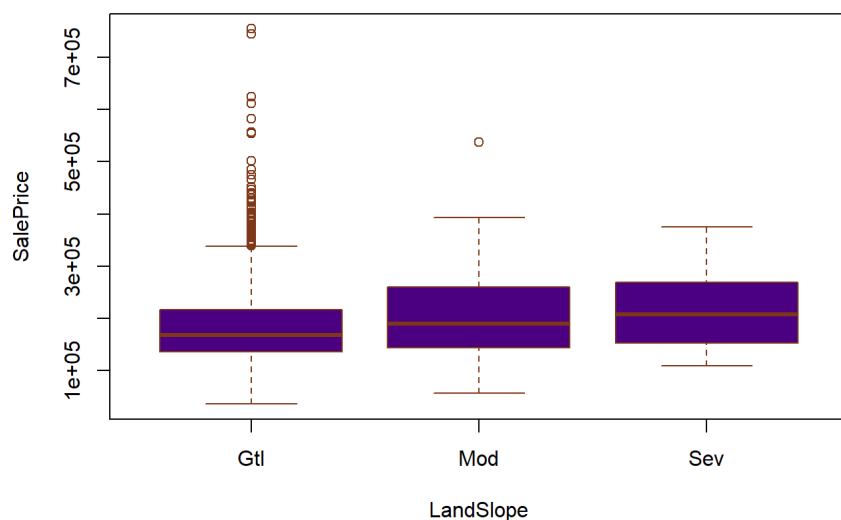
boxplot(SalePrice ~ LotConfig , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$LandSlope)

##
##  Gtl  Mod  Sev
## 1265   61   12

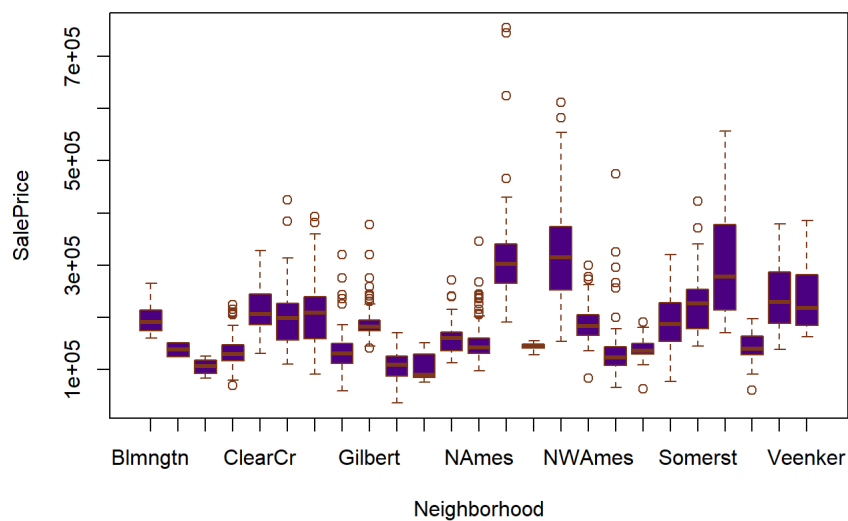
boxplot(SalePrice ~ LandSlope , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Neighborhood)
```

```
##
## Blmngtn Blueste BrDale BrkSide ClearCr CollgCr Crawfor Edwards Gilbert IDOTRR
## 17 2 15 47 26 146 50 70 77 29
## MeadowV Mitchel NAmes NoRidge NPKvill NridgHt NWAmes OldTown Sawyer SawyerW
## 12 42 209 41 9 75 73 100 69 53
## Somerst StoneBr SWISU Timber Veenker
## 83 25 20 37 11
```

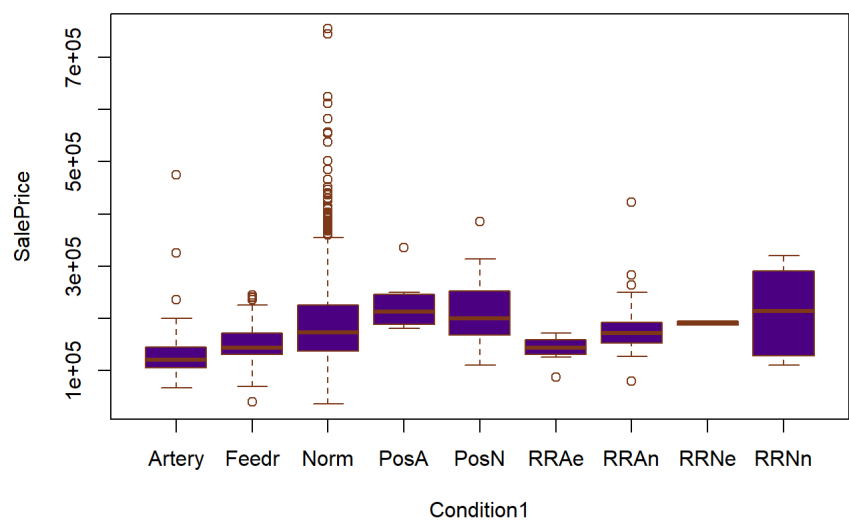
```
boxplot(SalePrice ~ Neighborhood , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Condition1)
```

```
##
## Artery Feedr Norm PosA PosN RRAe RRAn RRNe RRNn
## 43 63 1162 8 19 10 26 2 5
```

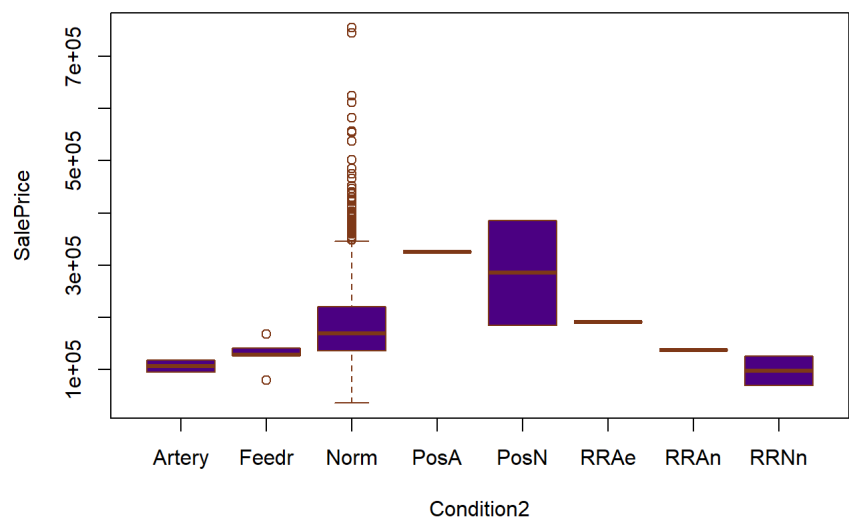
```
boxplot(SalePrice ~ Condition1 , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Condition2)

##
## Artery  Feedr  Norm  PosA  PosN  RRAe  RRAn  RRNn
##      2      5  1324    1     2     1     1     2

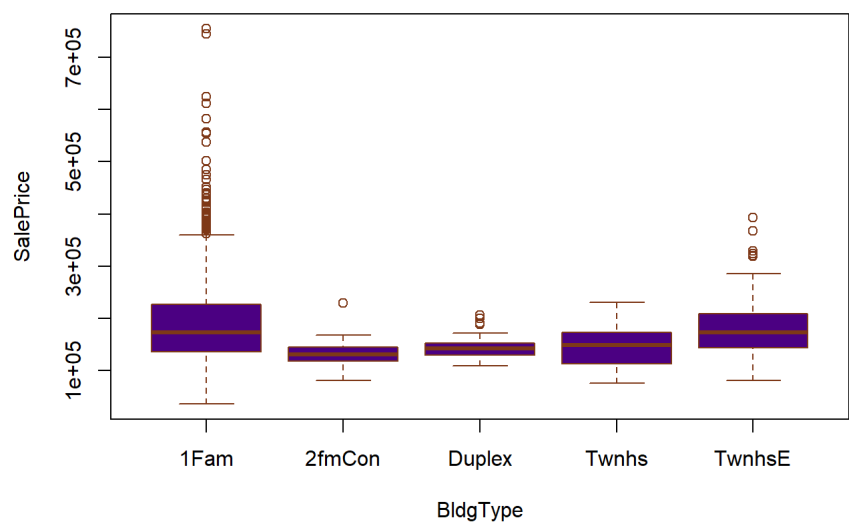
boxplot(SalePrice ~ Condition2 , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BldgType)

##
## 1Fam 2fmCon Duplex  Twnhs  TwnhsE
## 1138   22   28    38   112

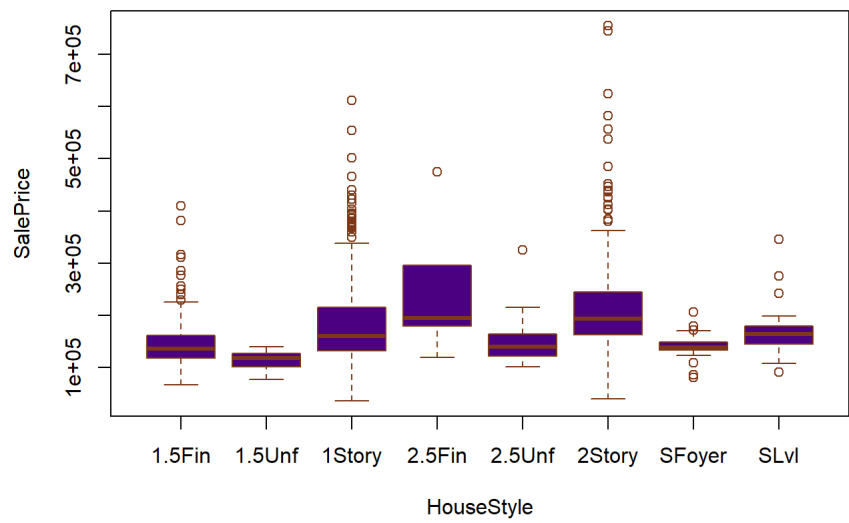
boxplot(SalePrice ~ BldgType , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$HouseStyle)
```

```
##
## 1.5Fin 1.5Unf 1Story 2.5Fin 2.5Unf 2Story SFoyer SLvl
##    134    11   657     6    10   426    30    64
```

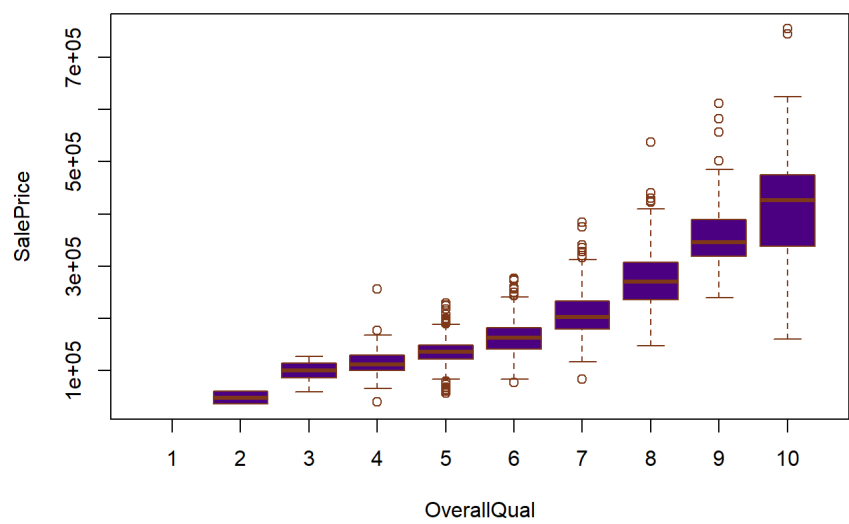
```
boxplot(SalePrice ~ HouseStyle , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$OverallQual)
```

```
##
## 1  2  3  4  5  6  7  8  9 10
## 0  2  8 81 351 359 312 165 43 17
```

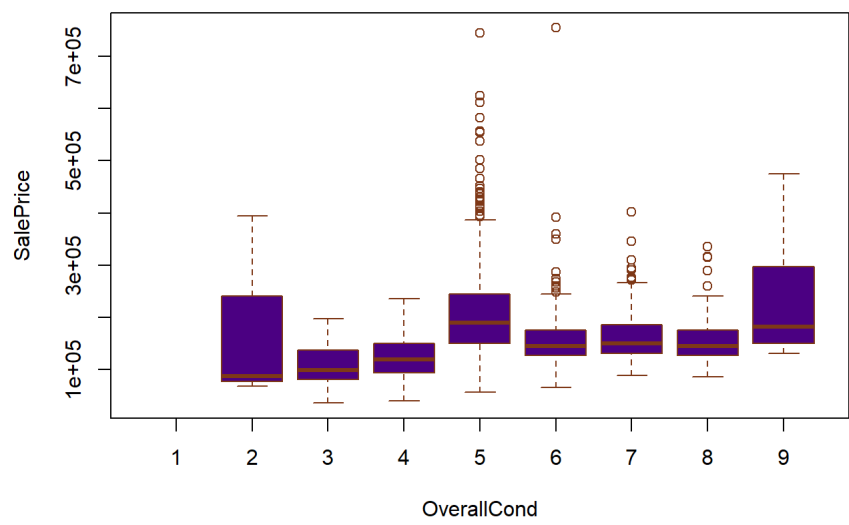
```
boxplot(SalePrice ~ OverallQual , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$OverallCond)

##
##  1  2  3  4  5  6  7  8  9
##  0  3 15 46 770 233 183 68 20

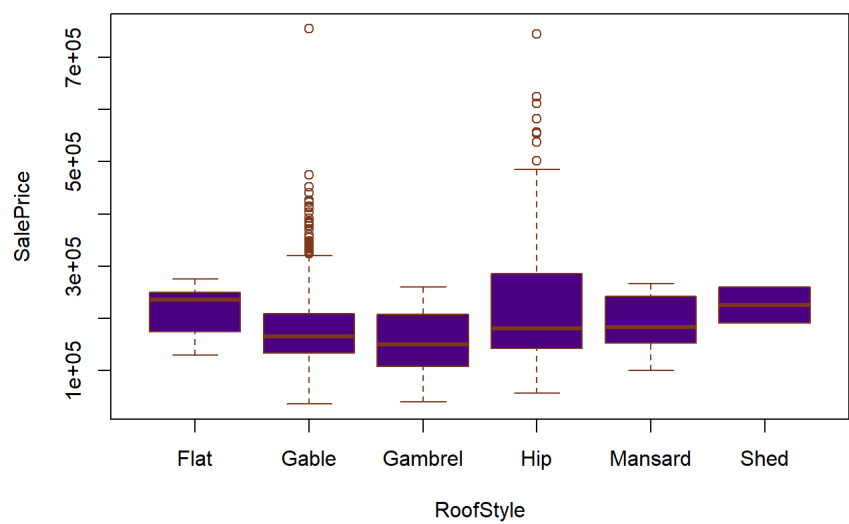
boxplot(SalePrice ~ OverallCond , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$RoofStyle)

##
##  Flat  Gable Gambrel  Hip Mansard  Shed
##    11   1037    10    272     6     2

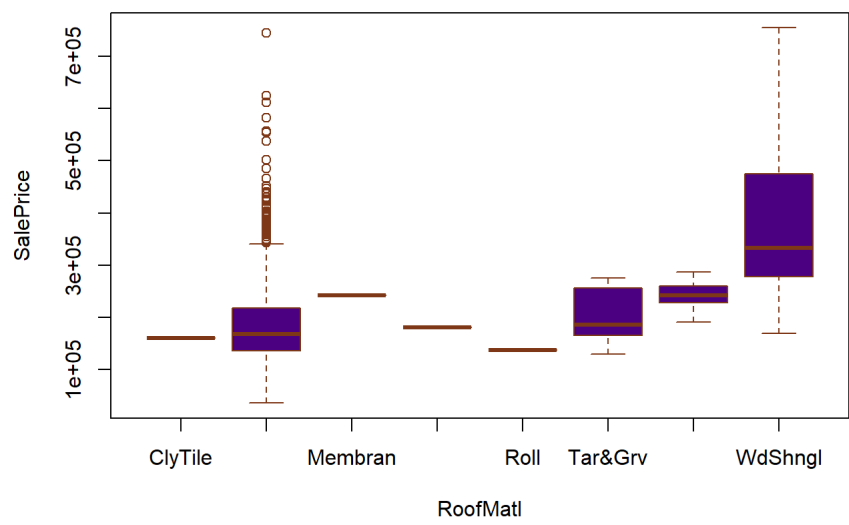
boxplot(SalePrice ~ RoofStyle , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$RoofMat1)

##
## ClyTile CompShg Membran Metal Roll Tar&Grv WdShake WdShngl
## 1 1314 1 1 1 9 5 6

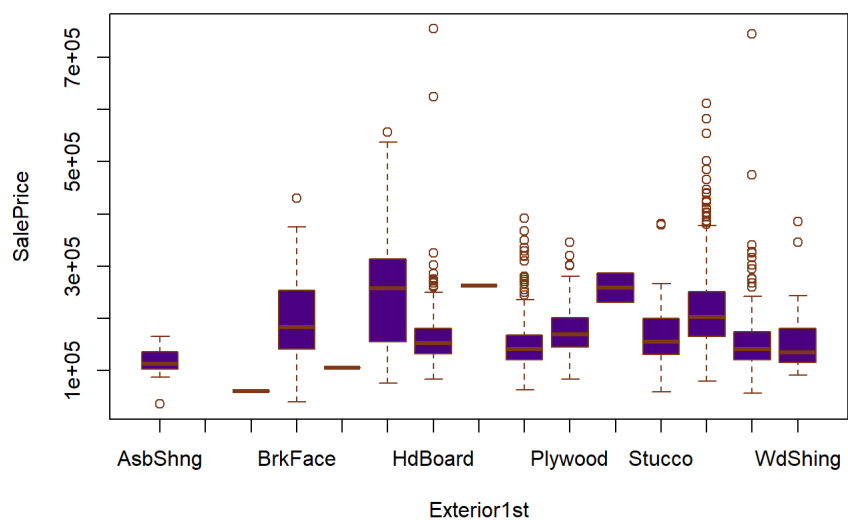
boxplot(SalePrice ~ RoofMat1 , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Exterior1st)

##
## AsbShng AsphShn BrkComm BrkFace CBlock CemntBd HdBoard ImStucc MetalSd Plywood
## 15 0 1 44 1 52 211 1 201 100
## Stone Stucco VinylSd Wd Sdng WdShng
## 2 21 486 183 20

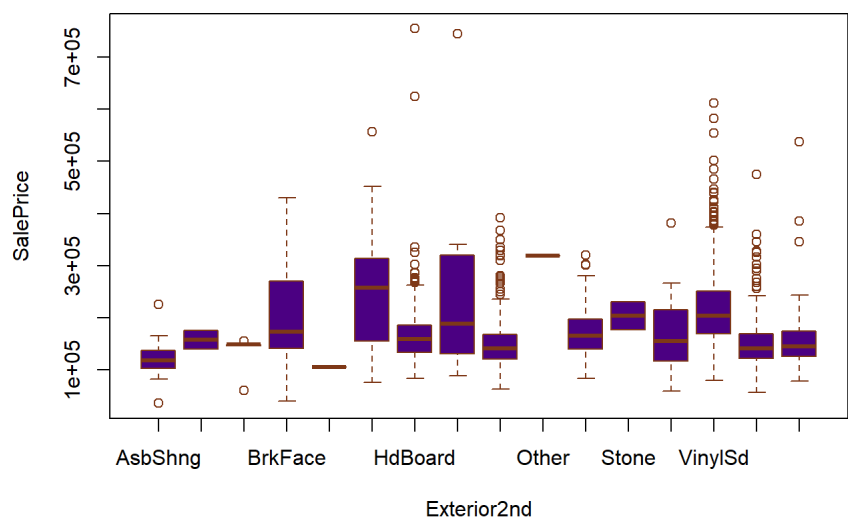
boxplot(SalePrice ~ Exterior1st , data = train3 , col = "#4B0082" , border = "#7E3817")
```

```
table(train3$Exterior2nd)
```

```
##
## AsbShng AsphShn Brk Cmn BrkFace CBlock CmentBd HdBoard ImStucc MetalSd Other
##      16      2      6      22      1      51      197      10      197      1
## Plywood  Stone  Stucco VinylSd Wd Sdng Wd Shng
##      127      2      23      475      176      32
```

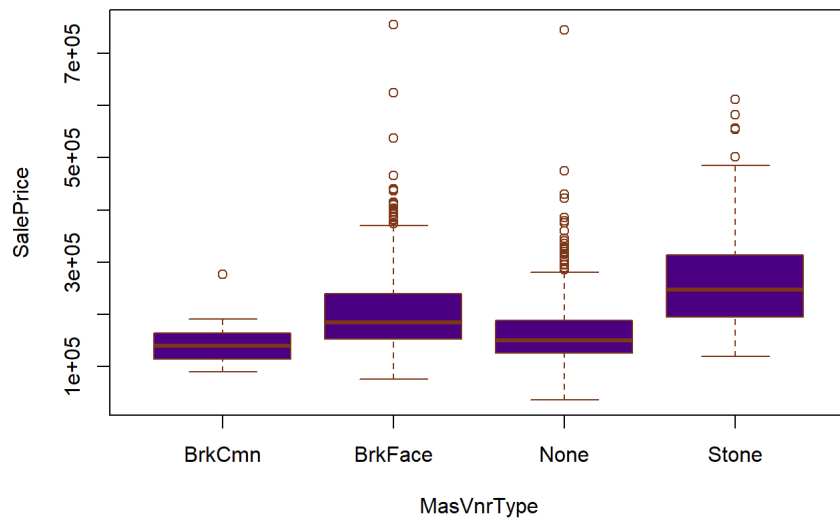
```
boxplot(SalePrice ~ Exterior2nd , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$MasVnrType)
```

```
##
## BrkCmn BrkFace  None  Stone
##      15      432      763      128
```

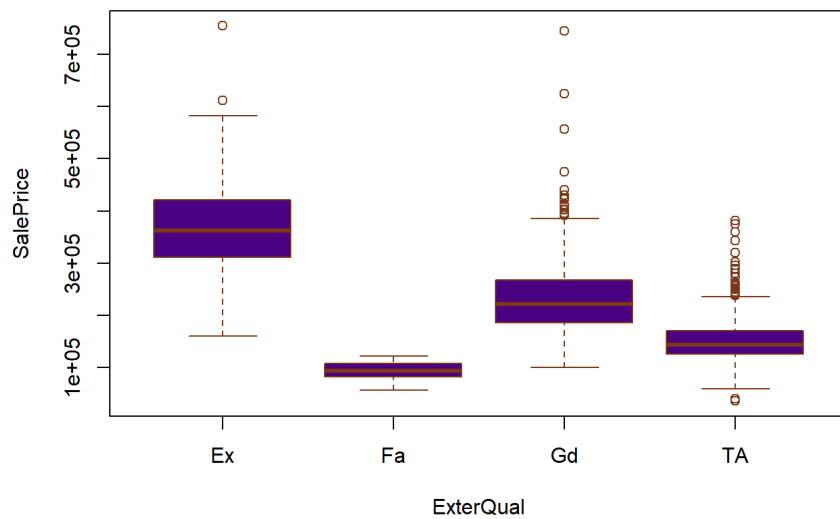
```
boxplot(SalePrice ~ MasVnrType , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$ExterQual)
```

```
##
##  Ex  Fa  Gd  TA
##  51   7 477 803
```

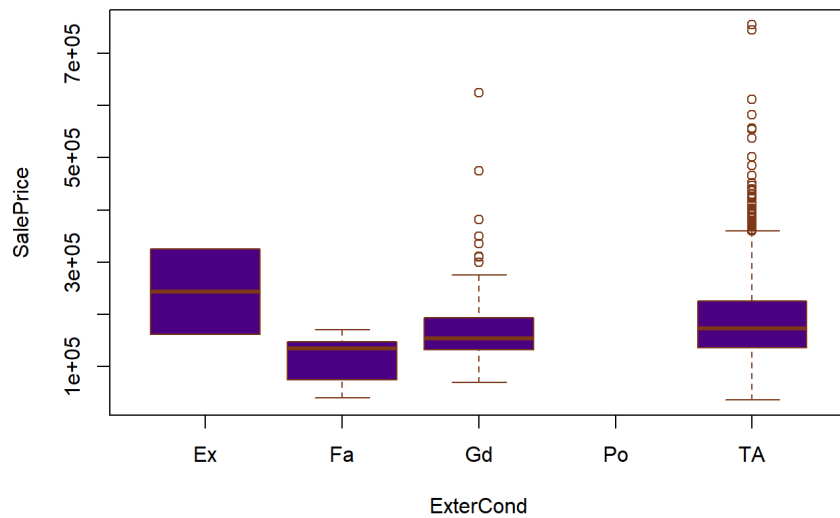
```
boxplot(SalePrice ~ ExterQual ,data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$ExterCond)
```

```
##
##  Ex  Fa  Gd  Po  TA
##   2  16 137   0 1183
```

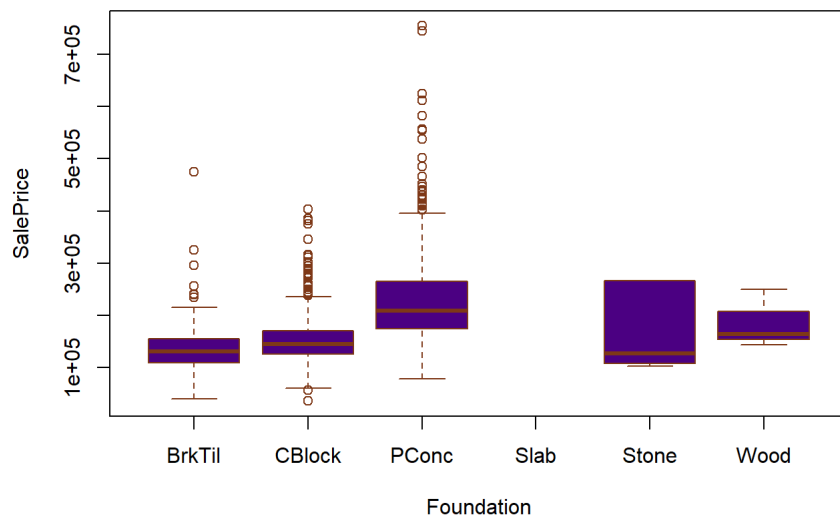
```
boxplot(SalePrice ~ ExterCond , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Foundation)
```

```
##
## BrkTil CBlock PConc Slab Stone Wood
## 129 580 620 0 6 3
```

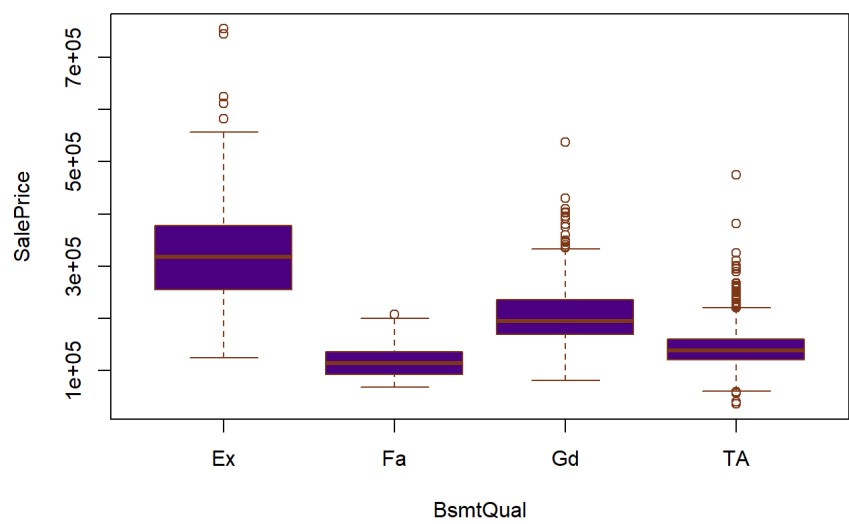
```
boxplot(SalePrice ~ Foundation , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BsmtQual)
```

```
##
## Ex Fa Gd TA
## 120 32 592 594
```

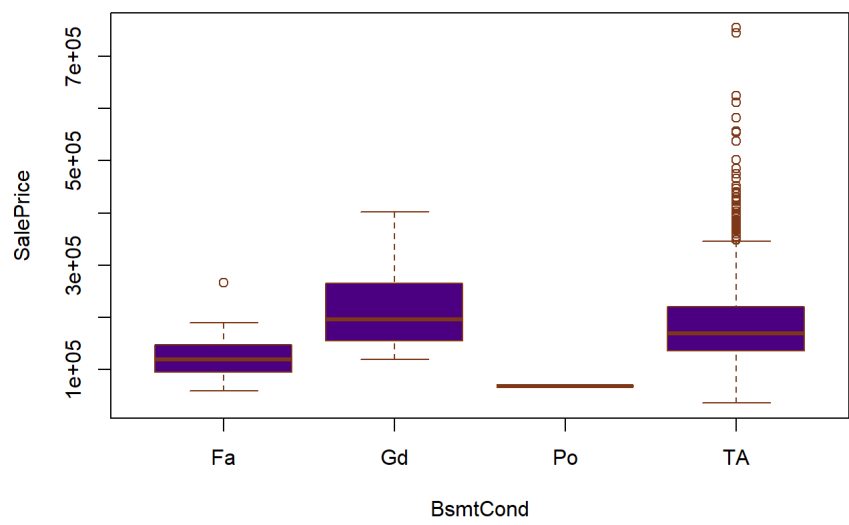
```
boxplot(SalePrice ~ BsmtQual , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BsmtCond)

##
##  Fa  Gd  Po  TA
##  38  62  1 1237

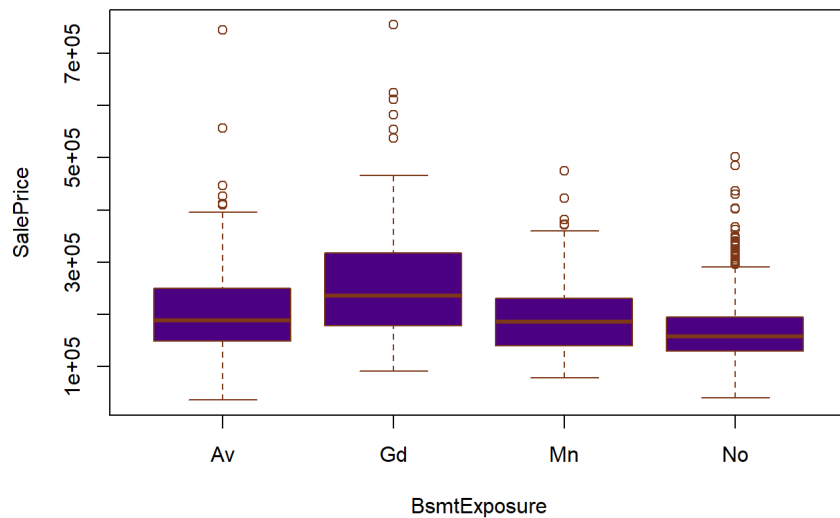
boxplot(SalePrice ~ BsmtCond , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BsmtExposure)

##
##  Av  Gd  Mn  No
## 213 127 111 887

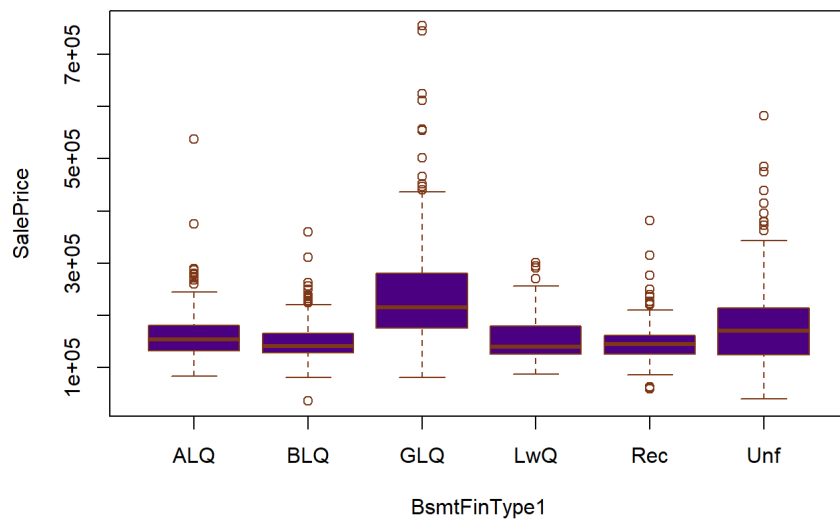
boxplot(SalePrice ~ BsmtExposure , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BsmtFinType1)
```

```
##
## ALQ BLQ GLQ LwQ Rec Unf
## 209 141 402 69 125 392
```

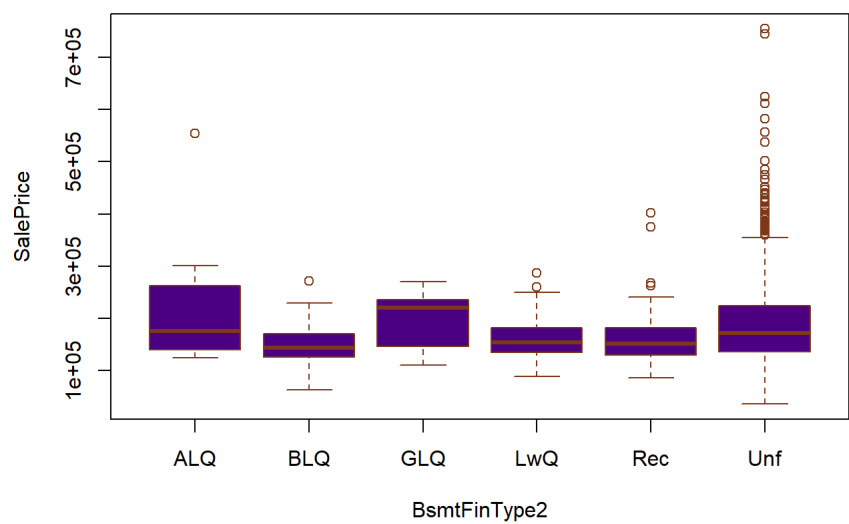
```
boxplot(SalePrice ~ BsmtFinType1 , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$BsmtFinType2)
```

```
##
## ALQ BLQ GLQ LwQ Rec Unf
## 19 32 12 46 53 1176
```

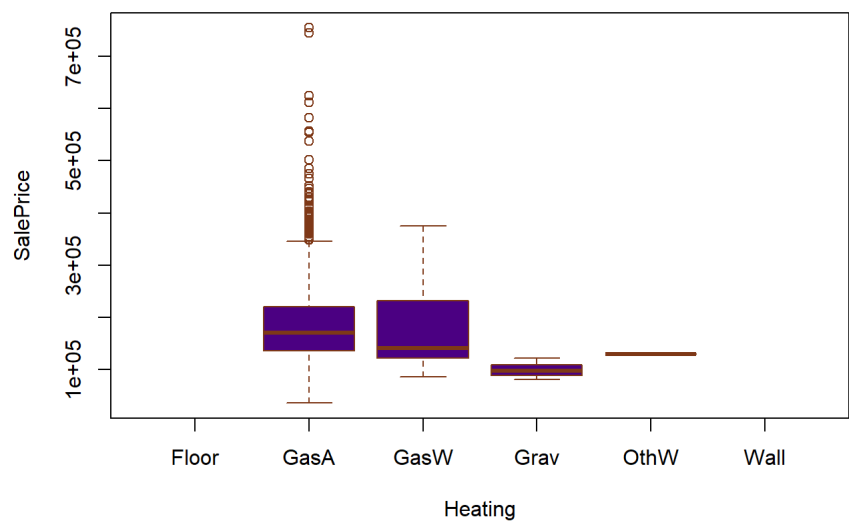
```
boxplot(SalePrice ~ BsmtFinType2 , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Heating)

##
## Floor GasA GasW Grav OthW Wall
##      0 1318  16   3   1   0

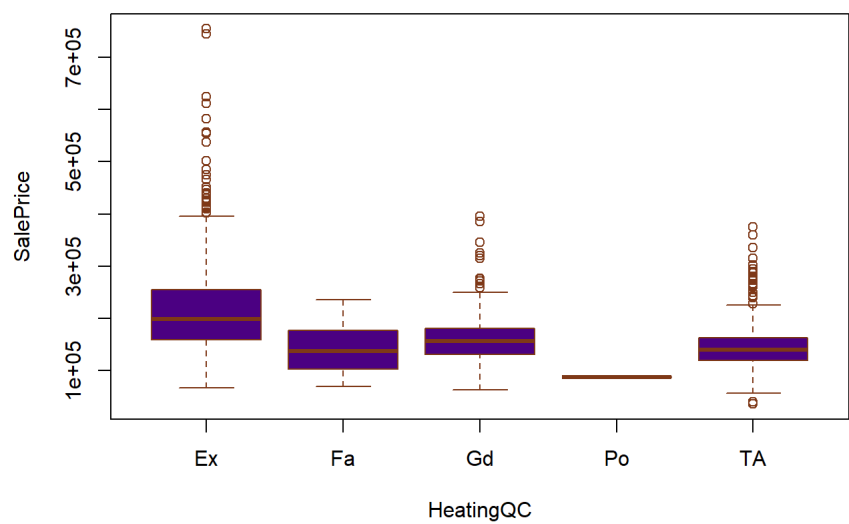
boxplot(SalePrice ~ Heating ,data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$HeatingQC)

##
## Ex Fa Gd Po TA
## 704 36 217  1 380

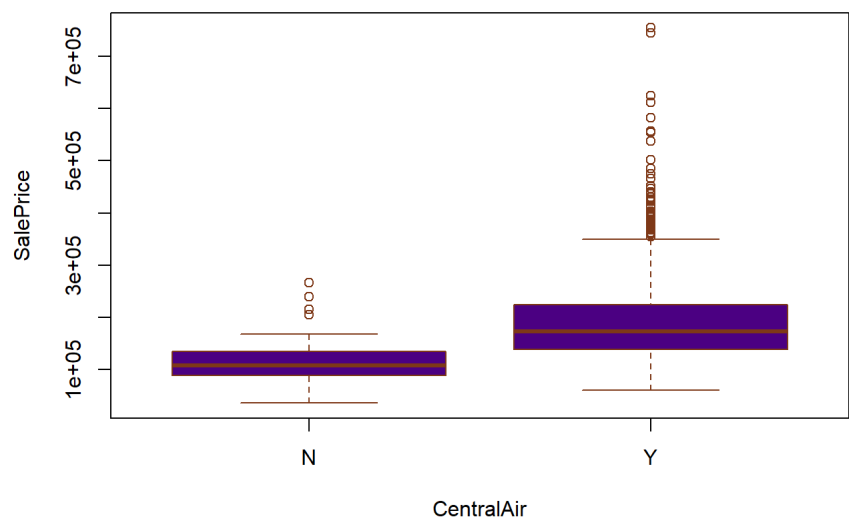
boxplot(SalePrice ~ HeatingQC , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$CentralAir)

##
##   N   Y
##  61 1277

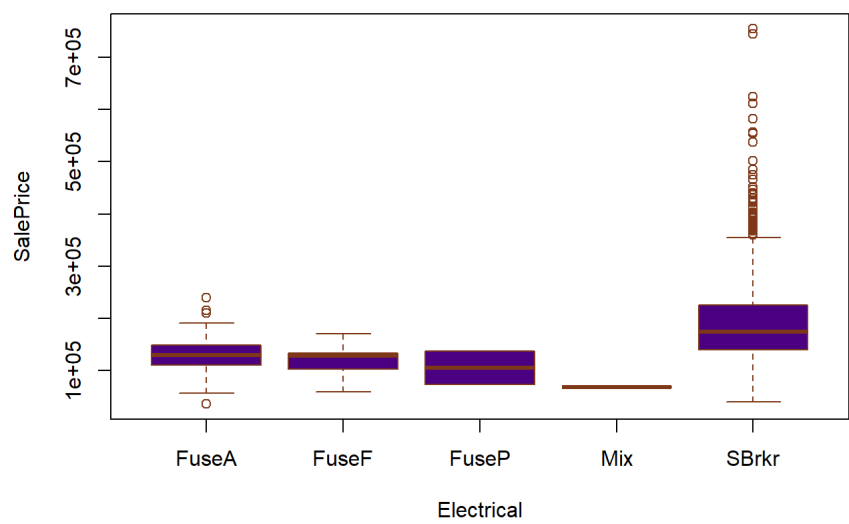
boxplot(SalePrice ~ CentralAir ,data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Electrical)

##
## FuseA FuseF FuseP   Mix SBrkr
##   76   17    2     1  1242

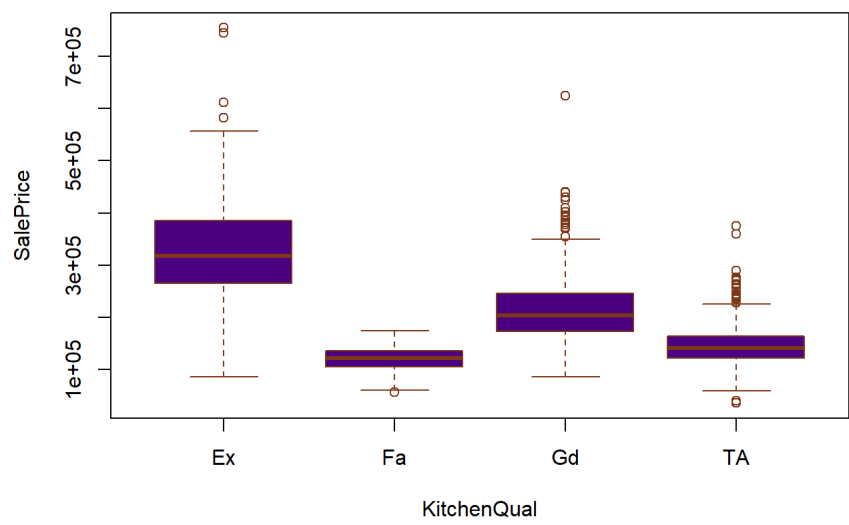
boxplot(SalePrice ~ Electrical , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$KitchenQual)

##
##  Ex  Fa  Gd  TA
##  97  23  568  650

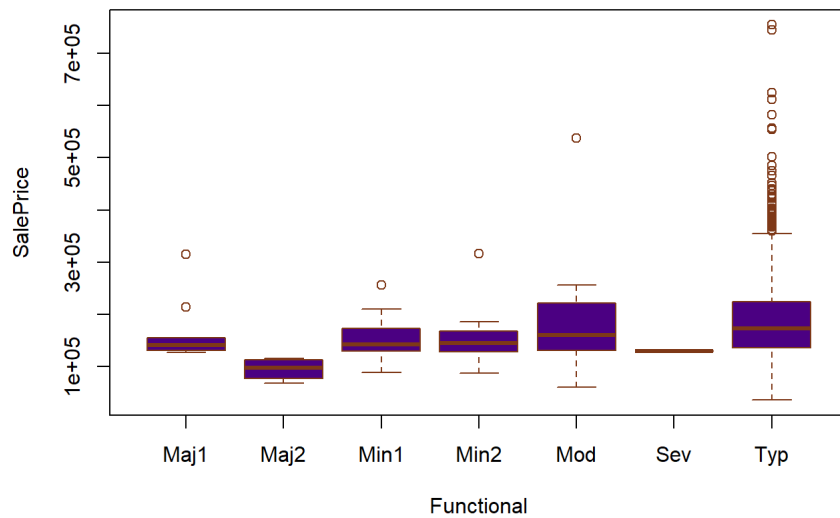
boxplot(SalePrice ~ KitchenQual , data = train3, col = "#4B0082" , border = "#7E3817")
```



```
table(train3$Functional)

##
##  Maj1 Maj2 Min1 Min2  Mod  Sev  Typ
##   10    4   28   30   11    1 1254

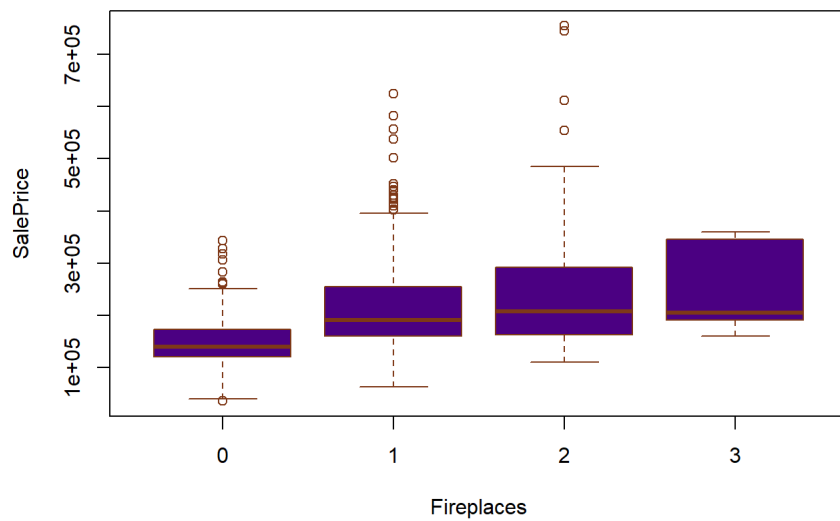
boxplot(SalePrice ~ Functional, data = train3 , col = "#4B0082" , border = "#7E3817")
```

```
table(train3$Fireplaces)
```

```
##
##  0  1  2  3
## 591 631 111  5
```

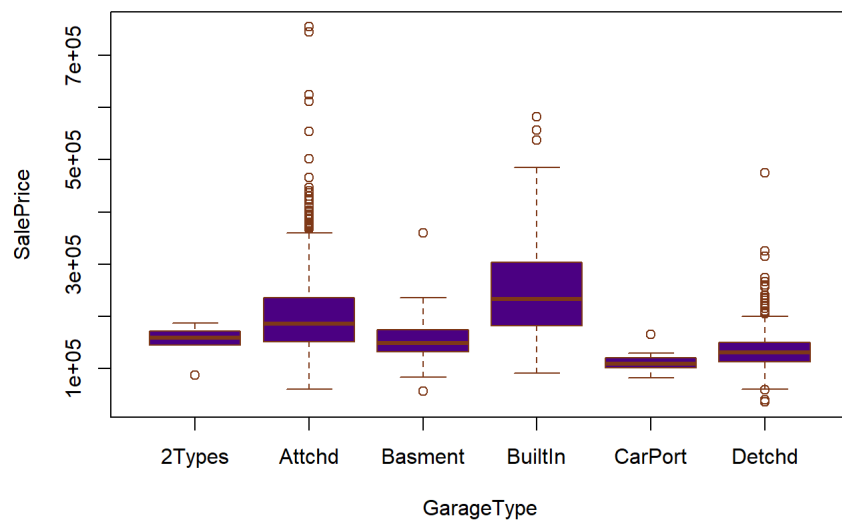
```
boxplot(SalePrice ~ Fireplaces , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$GarageType)
```

```
##
## 2Types Attchd Basment BuiltIn CarPort Detchd
##    6    852    19    85    7    369
```

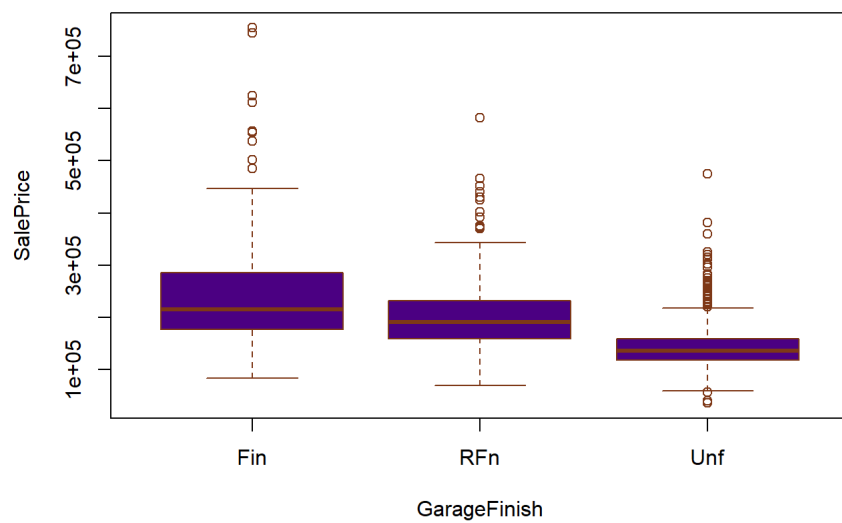
```
boxplot(SalePrice ~ GarageType , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$GarageFinish)
```

```
##
## Fin RFn Unf
## 345 413 580
```

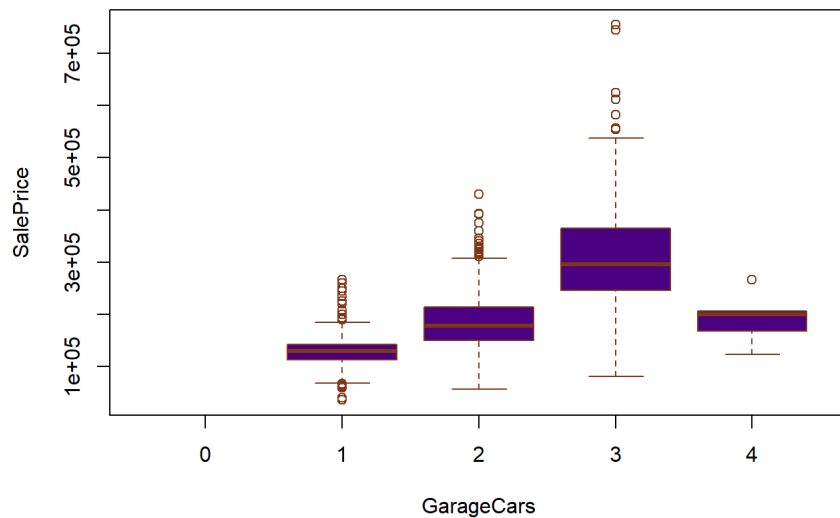
```
boxplot(SalePrice ~ GarageFinish , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$GarageCars)
```

```
##
## 0 1 2 3 4
## 0 361 793 179 5
```

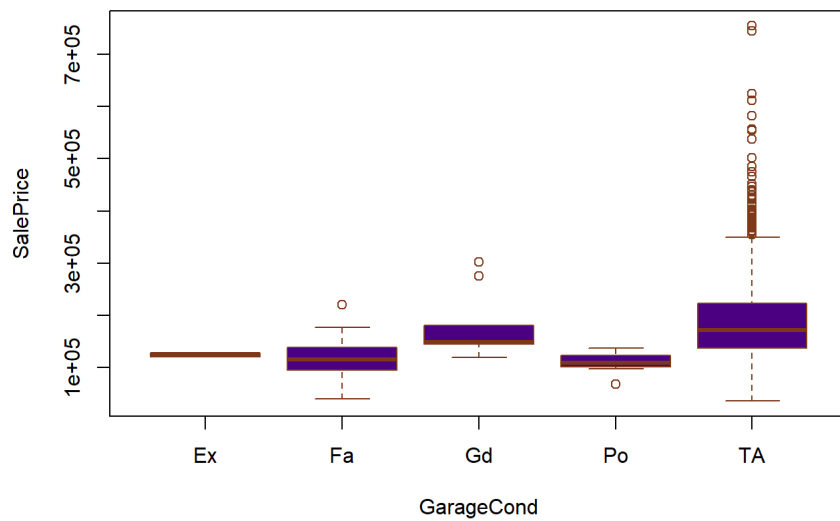
```
boxplot(SalePrice ~ GarageCars , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$GarageCond)
```

```
##
##  Ex  Fa  Gd  Po  TA
##    2  33  9   7 1287
```

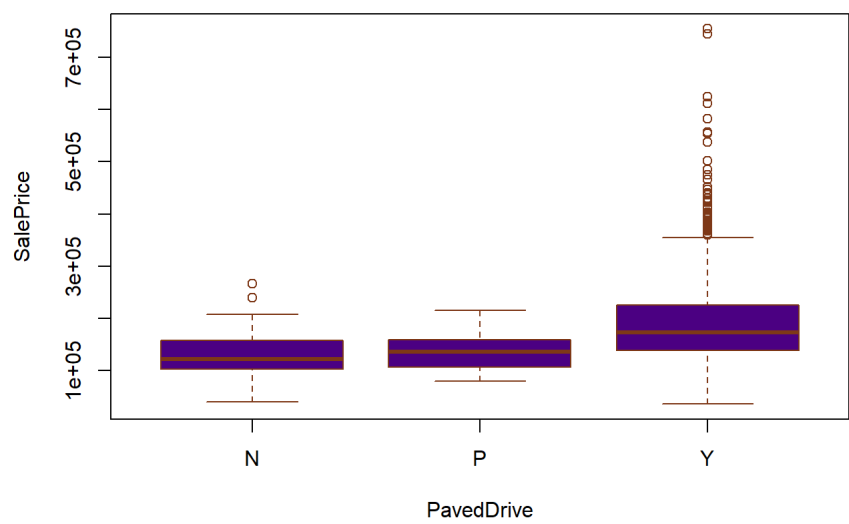
```
boxplot(SalePrice ~ GarageCond , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$PavedDrive)
```

```
##
##  N  P  Y
## 54 27 1257
```

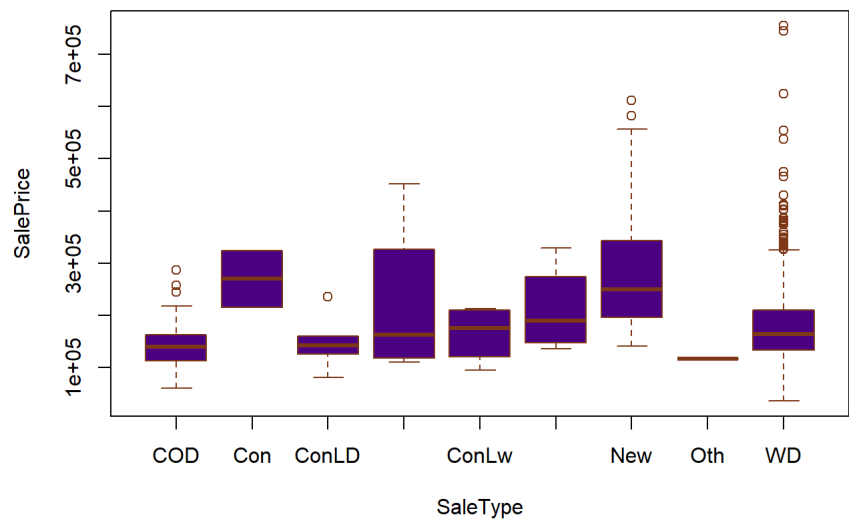
```
boxplot(SalePrice ~ PavedDrive , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$SaleType)
```

```
##
## COD Con ConLD ConLI ConLw CWD New Oth WD
## 42 2 6 4 4 4 117 1 1158
```

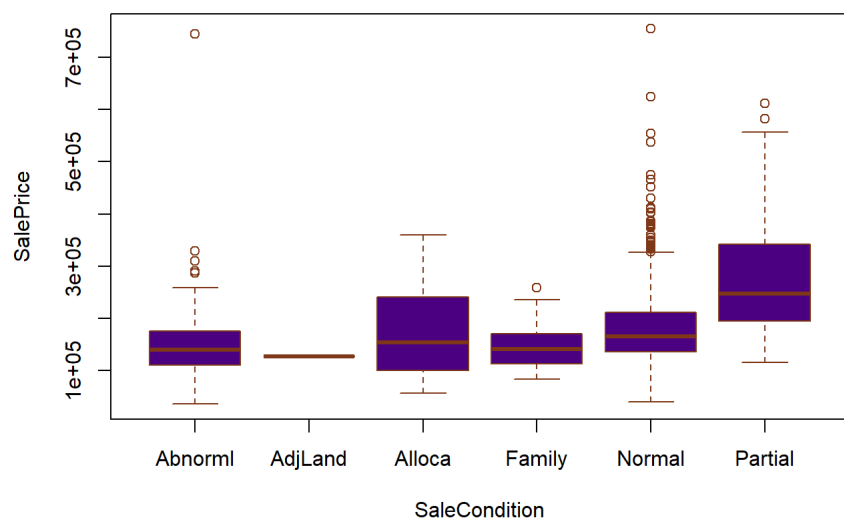
```
boxplot(SalePrice ~ SaleType , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
table(train3$SaleCondition)
```

```
##
## Abnorml AdjLand Alloca Family Normal Partial
## 86 1 7 20 1104 120
```

```
boxplot(SalePrice ~ SaleCondition , data = train3 , col = "#4B0082" , border = "#7E3817")
```



```
m1 <- lm(SalePrice ~ . ,data = train3 )  
summary(m1)
```

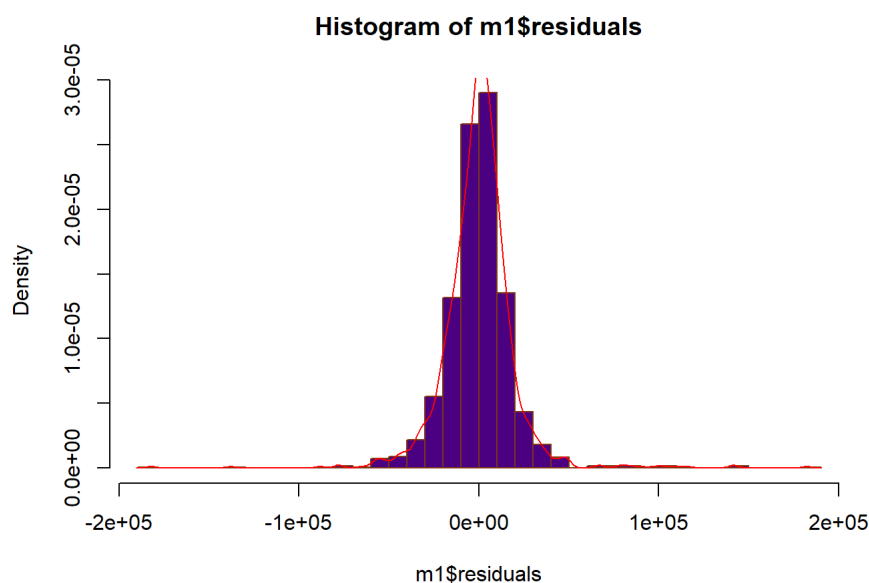
```
##
## Call:
## lm(formula = SalePrice ~ ., data = train3)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -182117   -8946     118     8735   182117
##
## Coefficients: (4 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -7.146e+05  6.781e+04 -10.539 < 2e-16 ***
## Id             4.377e-01  1.600e+00  0.274  0.784447
## MSSubClass    -3.399e+01  8.907e+01 -0.382  0.702841
## MSZoningFV     4.453e+04  1.327e+04  3.356  0.000817 ***
## MSZoningRH     3.768e+04  1.398e+04  2.694  0.007157 **
## MSZoningRL     3.652e+04  1.170e+04  3.121  0.001849 **
## MSZoningRM     3.257e+04  1.103e+04  2.954  0.003201 **
## LotArea       7.732e-01  1.119e-01  6.912  8.12e-12 ***
## StreetPave     4.897e+04  1.583e+04  3.094  0.002024 **
## LotShapeIR2    4.958e+03  4.273e+03  1.160  0.246167
## LotShapeIR3    4.846e+03  8.954e+03  0.541  0.588502
## LotShapeReg     9.651e+02  1.657e+03  0.582  0.560435
## LandContourHLS  6.227e+03  5.546e+03  1.123  0.261765
## LandContourLow -1.113e+04  6.989e+03 -1.593  0.111415
## LandContourLvl  4.992e+03  4.122e+03  1.211  0.226142
## UtilitiesNoSewa -3.559e+04  2.637e+04 -1.350  0.177359
## LotConfigCulDSac  6.750e+03  3.271e+03  2.063  0.039302 *
## LotConfigFR2    -9.238e+03  4.130e+03 -2.237  0.025489 *
## LotConfigFR3    -1.607e+04  1.255e+04 -1.281  0.200615
## LotConfigInside -1.598e+03  1.832e+03 -0.872  0.383205
## LandSlopeMod     8.688e+03  4.210e+03  2.063  0.039309 *
## LandSlopeSev    -4.411e+04  1.149e+04 -3.838  0.000131 ***
## NeighborhoodBlueste  2.020e+03  1.913e+04  0.106  0.915926
## NeighborhoodBrDale  5.752e+02  1.124e+04  0.051  0.959195
## NeighborhoodBrkSide  7.465e+02  1.006e+04  0.074  0.940842
## NeighborhoodClearCr -8.848e+03  9.591e+03 -0.922  0.356475
## NeighborhoodCollgCr -8.182e+03  7.386e+03 -1.108  0.268201
## NeighborhoodCrawfor  1.434e+04  8.786e+03  1.632  0.102914
## NeighborhoodEdwards -2.073e+04  8.274e+03 -2.506  0.012370 *
## NeighborhoodGilbert -7.041e+03  7.824e+03 -0.900  0.368375
## NeighborhoodIDOTRR -4.021e+03  1.145e+04 -0.351  0.725485
## NeighborhoodMeadowV -1.496e+04  1.187e+04 -1.261  0.207744
## NeighborhoodMitchel -1.540e+04  8.484e+03 -1.815  0.069726 .
## NeighborhoodNAMES -1.568e+04  8.016e+03 -1.956  0.050694 .
## NeighborhoodNoRidge  1.975e+04  8.534e+03  2.315  0.020804 *
## NeighborhoodNPKVill  1.066e+04  1.401e+04  0.761  0.446934
## NeighborhoodNridgHt  1.434e+04  7.580e+03  1.891  0.058869 .
## NeighborhoodNAMES -1.401e+04  8.232e+03 -1.701  0.089139 .
## NeighborhoodOldTown -9.599e+03  1.011e+04 -0.950  0.342556
## NeighborhoodSawyer -9.397e+03  8.281e+03 -1.135  0.256698
## NeighborhoodSawyerW -2.951e+03  7.998e+03 -0.369  0.712243
## NeighborhoodSomerst -2.675e+03  9.085e+03 -0.294  0.768506
## NeighborhoodStoneBr  3.286e+04  8.465e+03  3.881  0.000110 ***
## NeighborhoodSWISU   -9.256e+03  1.036e+04 -0.894  0.371645
## NeighborhoodTimber  -8.042e+03  8.244e+03 -0.975  0.329532
## NeighborhoodVeenker  2.837e+03  1.064e+04  0.267  0.789865
## Condition1Feedr    3.471e+03  5.545e+03  0.626  0.531493
## Condition1Norm     1.350e+04  4.497e+03  3.002  0.002745 **
## Condition1PosA     8.959e+03  1.008e+04  0.889  0.374090
## Condition1PosN     9.994e+03  7.649e+03  1.307  0.191634
## Condition1RR Ae    -1.325e+04  9.656e+03 -1.372  0.170287
## Condition1RRAn     7.722e+03  7.087e+03  1.089  0.276184
## Condition1RRNe     -5.488e+03  1.748e+04 -0.314  0.753591
## Condition1RRNn     1.249e+04  1.307e+04  0.956  0.339379
## Condition2Feedr    1.475e+02  2.538e+04  0.006  0.995363
## Condition2Norm     4.263e+03  2.211e+04  0.193  0.847109
## Condition2PosA    -1.476e+04  4.351e+04 -0.339  0.734568
## Condition2PosN    -2.580e+05  2.934e+04 -8.794 < 2e-16 ***
## Condition2RR Ae    -1.131e+05  4.715e+04 -2.398  0.016661 *
## Condition2RRAn    -2.184e+03  3.275e+04 -0.067  0.946837
## Condition2RRNn     1.268e+04  2.862e+04  0.443  0.657704
## BldgType2fmCon     -6.813e+03  1.418e+04 -0.480  0.630974
## BldgTypeDuplex     -1.145e+04  8.827e+03 -1.297  0.195007
## BldgTypeTwnhs     -1.765e+04  1.053e+04 -1.676  0.094055 .
## BldgTypeTwnhsE    -1.275e+04  9.571e+03 -1.332  0.183149
## HouseStyle1.5Unf    1.221e+04  9.420e+03  1.296  0.195210
## HouseStyle1Story    7.816e+03  4.772e+03  1.638  0.101713
## HouseStyle2.5Fin   -2.223e+04  1.457e+04 -1.526  0.127417
```

## HouseStyle2.5Unf	-3.381e+02	1.057e+04	-0.032	0.974490
## HouseStyle2Story	-3.285e+03	3.783e+03	-0.868	0.385426
## HouseStyleSFoyer	4.858e+03	7.026e+03	0.691	0.489474
## HouseStyleSLv1	5.512e+03	5.922e+03	0.931	0.352186
## OverallQual3	-1.571e+04	2.116e+04	-0.743	0.457865
## OverallQual4	-1.809e+04	1.931e+04	-0.937	0.348990
## OverallQual5	-1.560e+04	1.944e+04	-0.802	0.422453
## OverallQual6	-1.346e+04	1.949e+04	-0.691	0.490023
## OverallQual7	-8.562e+03	1.966e+04	-0.436	0.663259
## OverallQual8	4.710e+03	1.987e+04	0.237	0.812616
## OverallQual9	3.648e+04	2.051e+04	1.778	0.075611
## OverallQual10	7.783e+04	2.165e+04	3.595	0.000339 ***
## OverallCond3	-6.999e+03	2.101e+04	-0.333	0.739104
## OverallCond4	2.500e+02	2.006e+04	0.012	0.990057
## OverallCond5	9.413e+03	1.974e+04	0.477	0.633515
## OverallCond6	1.612e+04	1.985e+04	0.812	0.416901
## OverallCond7	2.265e+04	1.988e+04	1.140	0.254710
## OverallCond8	2.638e+04	2.014e+04	1.310	0.190502
## OverallCond9	3.772e+04	2.105e+04	1.792	0.073444
## RoofStyleGable	-3.535e+01	1.852e+04	-0.002	0.998477
## RoofStyleGambrel	1.990e+03	2.051e+04	0.097	0.922727
## RoofStyleHip	1.368e+02	1.856e+04	0.007	0.994123
## RoofStyleMansard	9.716e+03	2.241e+04	0.434	0.664712
## RoofStyleShed	9.652e+04	3.446e+04	2.801	0.005184 **
## RoofMatlCompShg	6.745e+05	3.537e+04	19.070	< 2e-16 ***
## RoofMatlMembran	7.630e+05	4.969e+04	15.353	< 2e-16 ***
## RoofMatlMetal	7.345e+05	4.847e+04	15.154	< 2e-16 ***
## RoofMatlRoll	6.840e+05	4.353e+04	15.713	< 2e-16 ***
## RoofMatlTar&Grv	6.747e+05	4.023e+04	16.771	< 2e-16 ***
## RoofMatlWdShake	6.649e+05	3.891e+04	17.088	< 2e-16 ***
## RoofMatlWdShngl	7.188e+05	3.648e+04	19.706	< 2e-16 ***
## Exterior1stBrkComm	-3.581e+04	3.403e+04	-1.052	0.292873
## Exterior1stBrkFace	6.028e+03	1.441e+04	0.418	0.675687
## Exterior1stCBlock	-2.787e+04	2.955e+04	-0.943	0.345827
## Exterior1stCemntBd	-5.551e+03	2.038e+04	-0.272	0.785415
## Exterior1stHdBoard	-1.454e+04	1.456e+04	-0.998	0.318264
## Exterior1stImStucc	-4.500e+04	2.804e+04	-1.605	0.108775
## Exterior1stMetalSd	-6.398e+03	1.671e+04	-0.383	0.701945
## Exterior1stPlywood	-1.521e+04	1.443e+04	-1.054	0.291899
## Exterior1stStone	-1.001e+04	2.700e+04	-0.371	0.710803
## Exterior1stStucco	-9.197e+03	1.604e+04	-0.573	0.566566
## Exterior1stVinylSd	-1.356e+04	1.479e+04	-0.917	0.359416
## Exterior1stWd Sdng	-1.218e+04	1.414e+04	-0.861	0.389360
## Exterior1stWdShing	-1.157e+04	1.516e+04	-0.763	0.445747
## Exterior2ndAsphShn	1.822e+04	2.354e+04	0.774	0.438981
## Exterior2ndBrk Cmn	1.829e+04	2.081e+04	0.879	0.379683
## Exterior2ndBrkFace	1.209e+04	1.465e+04	0.826	0.409170
## Exterior2ndCBlock	NA	NA	NA	NA
## Exterior2ndCmentBd	1.359e+04	1.974e+04	0.688	0.491309
## Exterior2ndHdBoard	1.656e+04	1.378e+04	1.202	0.229761
## Exterior2ndImStucc	2.890e+04	1.546e+04	1.870	0.061811
## Exterior2ndMetalSd	1.368e+04	1.602e+04	0.854	0.393231
## Exterior2ndOther	-1.186e+04	2.750e+04	-0.431	0.666412
## Exterior2ndPlywood	1.531e+04	1.344e+04	1.139	0.254915
## Exterior2ndStone	5.036e+03	2.435e+04	0.207	0.836219
## Exterior2ndStucco	1.496e+04	1.522e+04	0.983	0.325786
## Exterior2ndVinylSd	1.903e+04	1.402e+04	1.357	0.174932
## Exterior2ndWd Sdng	1.809e+04	1.341e+04	1.349	0.177538
## Exterior2ndWd Shng	1.440e+04	1.396e+04	1.031	0.302735
## MasVnrTypeBrkFace	8.737e+03	6.824e+03	1.280	0.200729
## MasVnrTypeNone	1.023e+04	6.884e+03	1.487	0.137380
## MasVnrTypeStone	1.165e+04	7.189e+03	1.621	0.105341
## MasVnrArea	1.505e+01	5.790e+00	2.600	0.009454 **
## ExterQualFa	1.734e+04	1.571e+04	1.104	0.269956
## ExterQualGd	-3.422e+03	5.224e+03	-0.655	0.512606
## ExterQualTA	-4.531e+03	5.675e+03	-0.798	0.424832
## ExterCondFa	-5.850e+02	2.656e+04	-0.022	0.982430
## ExterCondGd	-8.986e+03	2.551e+04	-0.352	0.724678
## ExterCondTA	-4.038e+03	2.553e+04	-0.158	0.874341
## FoundationCBlock	2.935e+03	3.537e+03	0.830	0.406862
## FoundationPConc	5.431e+03	3.757e+03	1.446	0.148508
## FoundationStone	1.008e+04	1.142e+04	0.883	0.377345
## FoundationWood	-3.276e+04	1.480e+04	-2.214	0.027019 *
## BsmtQualFa	-6.454e+03	6.638e+03	-0.972	0.331101
## BsmtQualGd	-1.033e+04	3.400e+03	-3.039	0.002429 **
## BsmtQualTA	-8.927e+03	4.212e+03	-2.119	0.034289 *
## BsmtCondGd	-5.016e+02	5.592e+03	-0.090	0.928541
## BsmtCondPo	1.195e+04	4.272e+04	0.280	0.779791
## BsmtCondTA	2.892e+03	4.661e+03	0.621	0.535026

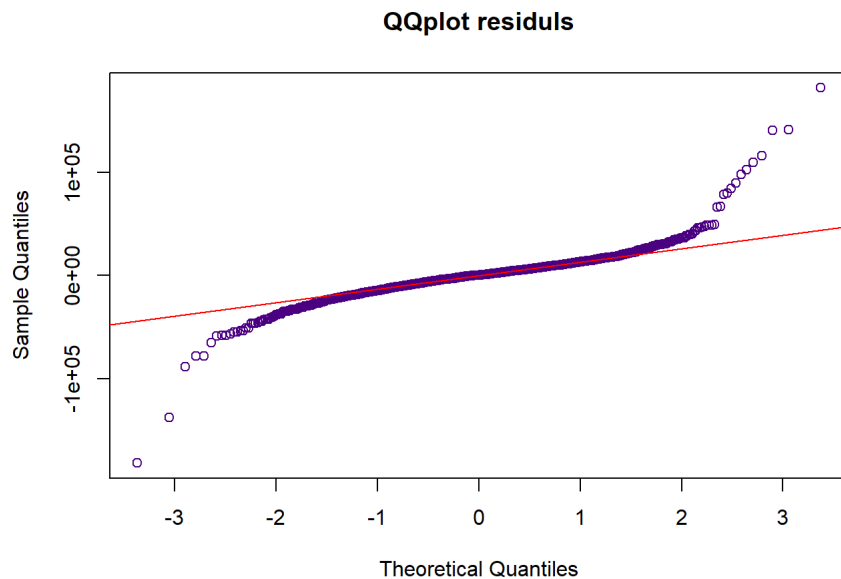
## BsmtExposureGd	1.168e+04	3.070e+03	3.806	0.000149	***
## BsmtExposureMn	-3.189e+03	3.061e+03	-1.042	0.297700	
## BsmtExposureNo	-4.374e+03	2.215e+03	-1.975	0.048518	*
## BsmtFinType1BLQ	1.608e+03	2.862e+03	0.562	0.574275	
## BsmtFinType1GLQ	5.912e+03	2.569e+03	2.301	0.021555	*
## BsmtFinType1LwQ	-2.654e+03	3.874e+03	-0.685	0.493428	
## BsmtFinType1Rec	9.310e+02	3.095e+03	0.301	0.763581	
## BsmtFinType1Unf	3.569e+03	3.031e+03	1.177	0.239341	
## BsmtFinSF1	3.456e+01	5.713e+00	6.049	1.99e-09	***
## BsmtFinType2BLQ	-8.625e+03	7.622e+03	-1.132	0.258074	
## BsmtFinType2GLQ	-2.648e+03	9.853e+03	-0.269	0.788210	
## BsmtFinType2LwQ	-1.280e+04	7.415e+03	-1.726	0.084611	.
## BsmtFinType2Rec	-9.464e+03	7.104e+03	-1.332	0.183073	
## BsmtFinType2Unf	-7.594e+03	7.580e+03	-1.002	0.316655	
## BsmtFinSF2	2.644e+01	9.290e+00	2.846	0.004503	**
## BsmtUnfSF	1.579e+01	5.337e+00	2.960	0.003147	**
## TotalBsmtSF	NA	NA	NA	NA	
## HeatingGasW	6.569e+02	7.653e+03	0.086	0.931611	
## HeatingGrav	1.023e+04	1.922e+04	0.533	0.594464	
## HeatingOthW	-1.006e+04	2.748e+04	-0.366	0.714315	
## HeatingQCFa	4.284e+02	5.072e+03	0.084	0.932704	
## HeatingQCGd	-2.771e+03	2.139e+03	-1.296	0.195387	
## HeatingQCPo	1.174e+04	2.746e+04	0.427	0.669163	
## HeatingQCTA	-2.935e+03	2.178e+03	-1.348	0.178027	
## CentralAirY	5.440e+03	4.685e+03	1.161	0.245828	
## ElectricalFuseF	-6.540e+03	7.315e+03	-0.894	0.371496	
## ElectricalFuseP	1.033e+04	2.302e+04	0.449	0.653669	
## ElectricalMix	NA	NA	NA	NA	
## ElectricalSBrkr	-1.362e+03	3.222e+03	-0.423	0.672612	
## X1stFlrSF	5.195e+01	6.133e+00	8.471	< 2e-16	***
## X2ndFlrSF	6.549e+01	5.842e+00	11.212	< 2e-16	***
## LowQualFinSF	1.764e+01	2.253e+01	0.783	0.433797	
## GrLivArea	NA	NA	NA	NA	
## BsmtFullBath	9.382e+02	2.044e+03	0.459	0.646374	
## BsmtHalfBath	1.560e+03	3.093e+03	0.504	0.614234	
## FullBath	3.839e+03	2.388e+03	1.607	0.108282	
## HalfBath	7.892e+02	2.206e+03	0.358	0.720640	
## BedroomAbvGr	-2.660e+03	1.473e+03	-1.805	0.071305	.
## KitchenAbvGr	-1.656e+04	7.406e+03	-2.237	0.025519	*
## KitchenQualFa	-1.535e+04	7.017e+03	-2.188	0.028883	*
## KitchenQualGd	-1.726e+04	3.646e+03	-4.732	2.51e-06	***
## KitchenQualTA	-1.619e+04	4.096e+03	-3.952	8.26e-05	***
## TotRmsAbvGrd	9.444e+02	9.845e+02	0.959	0.337620	
## FunctionalMaj2	-4.360e+03	1.728e+04	-0.252	0.800823	
## FunctionalMin1	6.682e+02	9.331e+03	0.072	0.942921	
## FunctionalMin2	-1.372e+02	9.490e+03	-0.014	0.988469	
## FunctionalMod	-6.256e+03	1.188e+04	-0.526	0.598654	
## FunctionalSev	-4.449e+04	2.994e+04	-1.486	0.137616	
## FunctionalTyp	1.209e+04	8.231e+03	1.469	0.142240	
## Fireplaces1	9.276e+02	1.781e+03	0.521	0.602602	
## Fireplaces2	8.036e+03	3.107e+03	2.586	0.009827	**
## Fireplaces3	2.341e+03	1.229e+04	0.191	0.848940	
## GarageTypeAttchd	2.150e+04	1.156e+04	1.859	0.063307	.
## GarageTypeBasement	2.408e+04	1.339e+04	1.798	0.072392	.
## GarageTypeBuiltIn	2.049e+04	1.200e+04	1.707	0.088108	.
## GarageTypeCarPort	2.575e+04	1.652e+04	1.559	0.119317	
## GarageTypeDetchd	2.332e+04	1.157e+04	2.015	0.044192	*
## GarageFinishRfn	-9.552e+02	1.984e+03	-0.481	0.630345	
## GarageFinishUnf	-1.784e+02	2.440e+03	-0.073	0.941744	
## GarageCars2	-8.922e+02	2.698e+03	-0.331	0.740942	
## GarageCars3	8.816e+03	4.947e+03	1.782	0.074993	.
## GarageCars4	1.849e+04	1.260e+04	1.467	0.142575	
## GarageArea	1.824e+01	8.091e+00	2.254	0.024364	*
## GarageQualFa	-4.916e+04	3.189e+04	-1.541	0.123544	
## GarageQualGd	-3.905e+04	3.270e+04	-1.194	0.232591	
## GarageQualPo	-7.294e+04	4.066e+04	-1.794	0.073098	.
## GarageQualTA	-4.490e+04	3.154e+04	-1.423	0.154919	
## GarageCondFa	3.953e+04	3.622e+04	1.091	0.275320	
## GarageCondGd	4.025e+04	3.738e+04	1.077	0.281822	
## GarageCondPo	4.563e+04	3.886e+04	1.174	0.240512	
## GarageCondTA	4.177e+04	3.583e+04	1.166	0.243897	
## PavedDriveP	-3.260e+03	6.270e+03	-0.520	0.603167	
## PavedDriveY	1.486e+03	4.237e+03	0.351	0.725811	
## WoodDeckSF	9.165e+00	5.956e+00	1.539	0.124127	
## OpenPorchSF	8.870e+00	1.230e+01	0.721	0.470957	
## EnclosedPorch	6.918e+00	1.309e+01	0.528	0.597335	
## X3SsnPorch	4.719e+01	2.254e+01	2.094	0.036496	*
## ScreenPorch	3.755e+01	1.234e+01	3.042	0.002402	**
## PoolArea	6.694e+01	1.846e+01	3.625	0.000302	***


```
## MiscVal          4.415e-01  1.428e+00  0.309 0.757275
## MoSold           -4.235e+02  2.549e+02 -1.662 0.096877 .
## SaleTypeCon      2.692e+04  1.746e+04  1.542 0.123370
## SaleTypeConLD    2.016e+04  1.195e+04  1.687 0.091839 .
## SaleTypeConLI    -4.808e+03  1.292e+04 -0.372 0.709919
## SaleTypeConLw     1.025e+03  1.305e+04  0.079 0.937417
## SaleTypeCWD       8.743e+03  1.293e+04  0.676 0.499006
## SaleTypeNew       2.961e+04  1.569e+04  1.887 0.059424 .
## SaleTypeOth       2.012e+04  2.331e+04  0.863 0.388310
## SaleTypeWD        -4.980e+02  4.321e+03 -0.115 0.908262
## SaleConditionAdjLand 3.759e+04  2.545e+04  1.477 0.140077
## SaleConditionAlloca 1.100e+04  1.116e+04  0.985 0.324823
## SaleConditionFamily -1.491e+03  6.164e+03 -0.242 0.808968
## SaleConditionNormal 5.279e+03  3.098e+03  1.704 0.088696 .
## SaleConditionPartial -7.685e+03  1.507e+04 -0.510 0.610221
## ageBuilt          -3.402e+02  8.867e+01 -3.837 0.000132 ***
## ageRemodAdd        -6.951e+01  6.188e+01 -1.123 0.261535
## GarageageBlt       -1.718e+00  6.478e+01 -0.027 0.978851
## ageSold            2.071e+01  5.323e+02  0.039 0.968975
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 22360 on 1097 degrees of freedom
## Multiple R-squared:  0.9341, Adjusted R-squared:  0.9197
## F-statistic: 64.81 on 240 and 1097 DF,  p-value: < 2.2e-16
```

```
hist(m1$residuals , probability = TRUE , breaks = 35 ,col = "#4B0082" , border = "#7E3817" )
lines(density(m1$residuals) , col = "red")
```



```
qqnorm(m1$residuals , main = "QQplot residuls" ,col = "#4B0082" )
qqline(m1$residuals , col = "red")
```



```
jarque.test(m1$residuals)
```

```
##  
## Jarque-Bera Normality Test  
##  
## data: m1$residuals  
## JB = 17222, p-value < 2.2e-16  
## alternative hypothesis: greater
```

```
anscombe.test(m1$residuals)
```

```
##  
## Anscombe-Glynn kurtosis test  
##  
## data: m1$residuals  
## kurt = 20.532, z = 18.004, p-value < 2.2e-16  
## alternative hypothesis: kurtosis is not equal to 3
```