Regression

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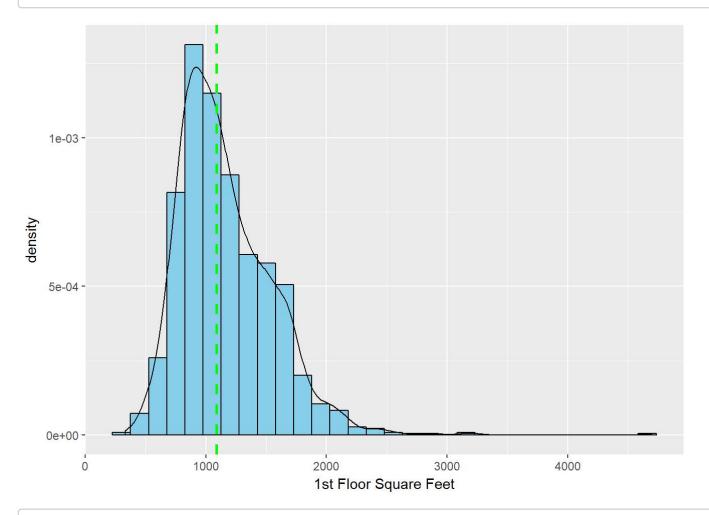
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```
library(MASS)
library(ggplot2)
```

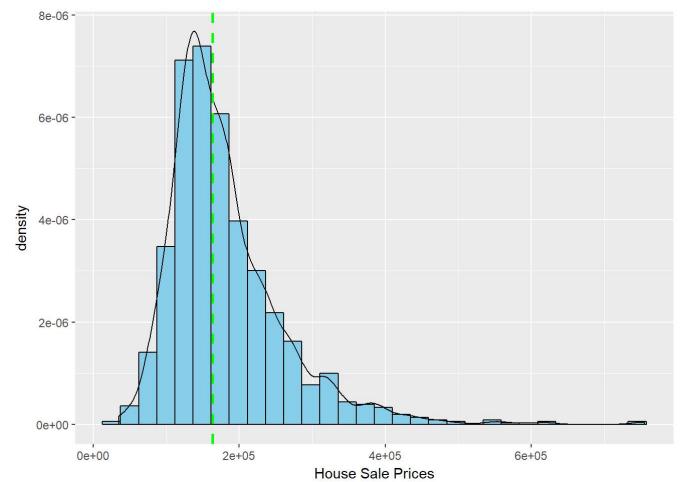
```
mydata <-read.table("https://github.com/angus001/Data605/raw/master/train.csv",header = T, sep=
",")
testdata <- read.table("https://github.com/angus001/Data605/raw/master/test.csv",header = T, sep
= ",")</pre>
```

```
X<-mydata$X1stFlrSF
Y<-mydata$SalePrice
df<-data.frame(X,Y)</pre>
```

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

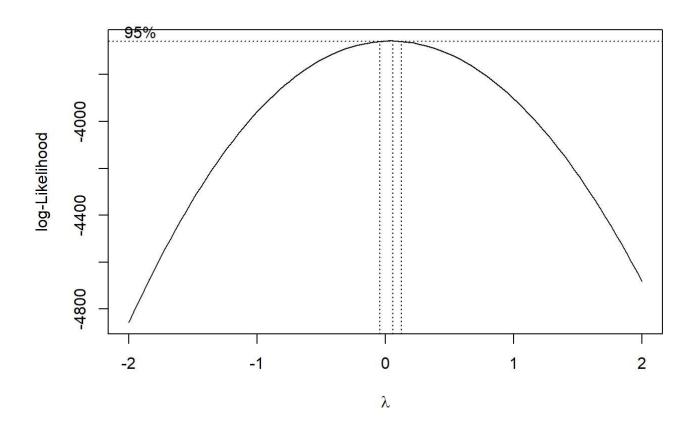


```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



Perform boxcox analysis to find log-Likelihood. Look for Lambda value with max likelihood. THe max likelihood is at 0.06 power.

```
bc = boxcox(Y~X, data = df)
```



```
lamda =bc$x
likelihood = bc$y
bc1=cbind(lamda,likelihood)
head(bc1[order(-likelihood),])
```

```
## lamda likelihood

## [1,] 0.06060606 -3656.618

## [2,] 0.02020202 -3656.622

## [3,] 0.1010101 -3657.490

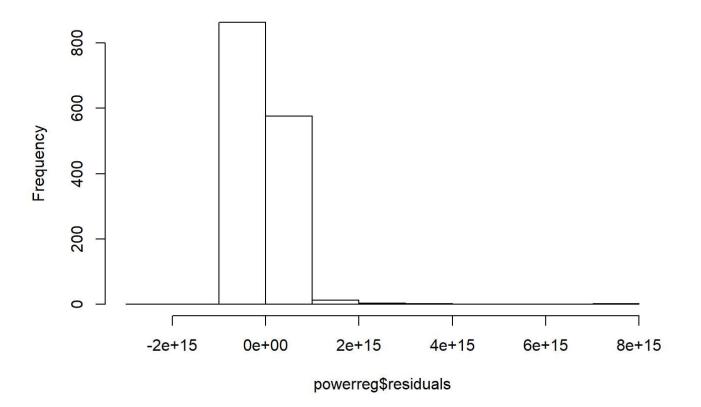
## [4,] -0.02020202 -3657.505

## [5,] 0.14141414 -3659.237

## [6,] -0.06060606 -3659.270
```

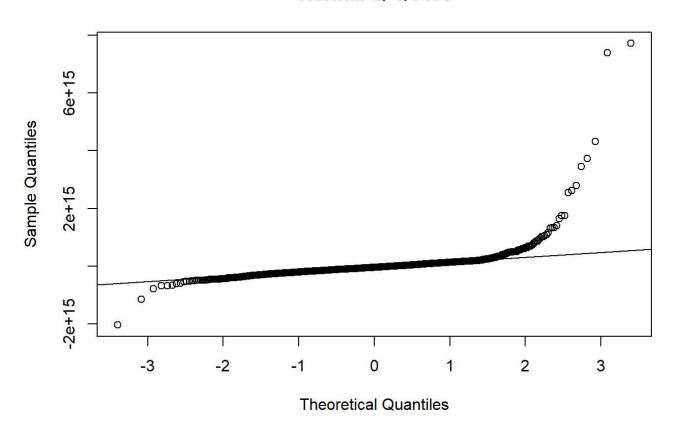
```
df$Ypower = (df$Y)^3/50
powerreg <- lm(Ypower~X, df)
hist(powerreg$residuals)</pre>
```

Histogram of powerreg\$residuals



qqnorm(powerreg\$residuals)
qqline(powerreg\$residuals)

Normal Q-Q Plot



Perform a correlation test between variables. The correlation test shows a correlation of 0.605 without tranforming the variable. The correlation actually become less to 0.441 after transforming the variable.

```
cor(df)
##
                  Х
                             Υ
                                  Ypower
## X
          1.0000000 0.6058522 0.4411002
## Y
          0.6058522 1.0000000 0.8019417
## Ypower 0.4411002 0.8019417 1.0000000
cor.test(df$X,df$Y, conf.level = 0.99)
##
##
    Pearson's product-moment correlation
##
## data: df$X and df$Y
  t = 29.078, df = 1458, p-value < 2.2e-16
  alternative hypothesis: true correlation is not equal to \theta
  99 percent confidence interval:
##
    0.5613896 0.6468270
```

Fitting the data point into different distribution to understand the underlying spread of the data.

```
fit <- fitdistr(df$X, densfun = 'cauchy')
fit

## location scale
## 1059.239655 212.473032
## ( 9.090705) ( 7.210534)</pre>
```

```
lamda2 <- fit$estimate
lamda2
```

##

##

sample estimates:

cor

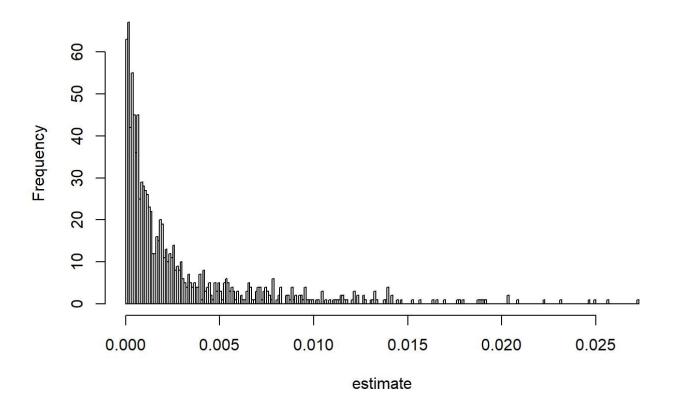
0.6058522

```
## location scale
## 1059.240 212.473
```

Take 1000 samples from the distribution, plot a histogram and compare with the non-transformed original values.

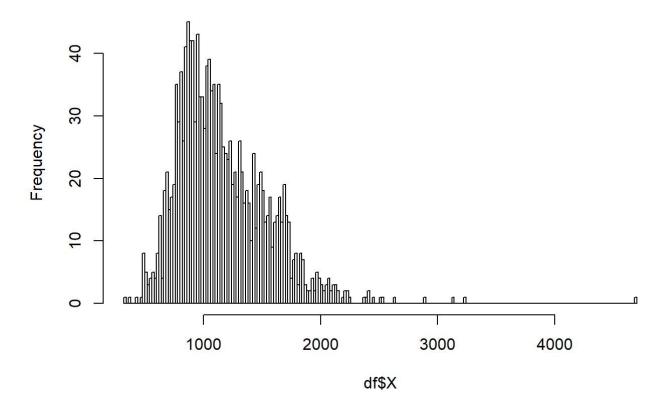
```
estimate <- rexp(1000,lamda2)
hist(estimate, breaks = 200)
```

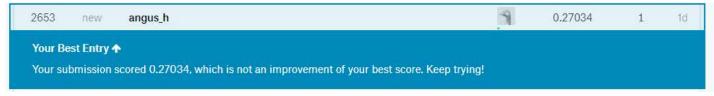
Histogram of estimate



hist(df\$X, breaks = 200)

Histogram of df\$X





Kaggle Result

#![Kaggle Result](https://github.com/angus001/Data605/blob/master/kagglefirsttry.PNG?raw=true)