

Stat350-Final Project

```
library(dplyr)
```

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

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

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

```
library(stats)  
library(factoextra)
```

```
## Warning: package 'factoextra' was built under R version 3.6.2
```

```
## Loading required package: ggplot2
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa
```

```
library(car)
```

```
## Warning: package 'car' was built under R version 3.6.2
```

```
## Loading required package: carData
```

```
##  
## Attaching package: 'car'
```

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

```
# import data
```

```
df1 <- read.csv('Real estate.csv', header = T)  
names(df1)
```

```
## [1] "No"  
## [2] "X1.transaction.date"  
## [3] "X2.house.age"  
## [4] "X3.distance.to.the.nearest.MRT.station"  
## [5] "X4.number.of.convenience.stores"  
## [6] "X5.latitude"  
## [7] "X6.longitude"  
## [8] "Y.house.price.of.unit.area"
```

```
# create year  
df1$Year <- floor(df1$X1.transaction.date)  
names(df1)
```

```
## [1] "No"
## [2] "X1.transaction.date"
## [3] "X2.house.age"
## [4] "X3.distance.to.the.nearest.MRT.station"
## [5] "X4.number.of.convenience.stores"
## [6] "X5.latitude"
## [7] "X6.longitude"
## [8] "Y.house.price.of.unit.area"
## [9] "Year"
```

```
table(df1$Year)
```

```
##
## 2012 2013
## 126 288
```

```
df1$X1.transaction.date <- df1$Year
names(df1)[2] <- 'Year'
df1 <- df1[, -9]
names(df1)
```

```
## [1] "No"
## [2] "Year"
## [3] "X2.house.age"
## [4] "X3.distance.to.the.nearest.MRT.station"
## [5] "X4.number.of.convenience.stores"
## [6] "X5.latitude"
## [7] "X6.longitude"
## [8] "Y.house.price.of.unit.area"
```

```
# transform of the response
df1$Y2 <- sqrt(df1$Y.house.price.of.unit.area)

df2 <- df1 %>%
  filter(Year == '2012')

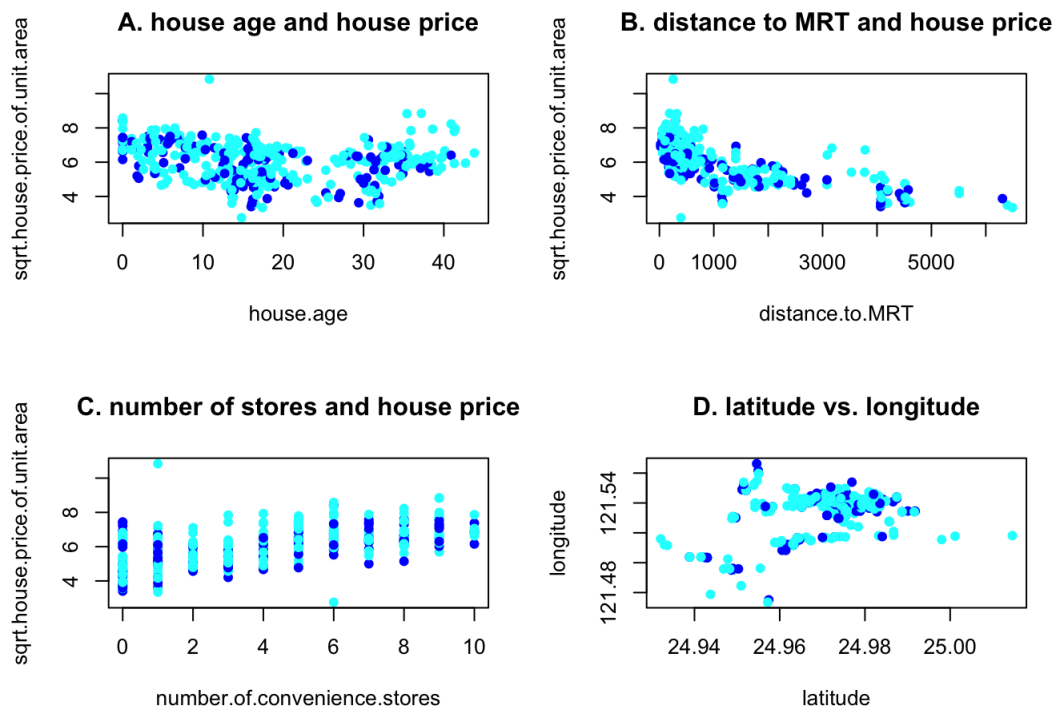
# introduce new data point
df1[415, 1] <- 415
df1[415, 2] <- '2012'
N <- 414
for (i in 3: 8){
  set.seed(i)
  SRS.index <- sample.int(N, 1, replace = FALSE)
  df1[415, i] <- df1[SRS.index, i]
}
df1[415, ]
```

```
##      No Year X2.house.age X3.distance.to.the.nearest.MRT.station
## 415 415 2012          17                379.5575
##      X4.number.of.convenience.stores X5.latitude X6.longitude
## 415                                4    24.98203    121.5458
##      Y.house.price.of.unit.area Y2
## 415                        28.6 NA
```

```
# Potential outlier
df1[which.max(df1$Y2),]
```

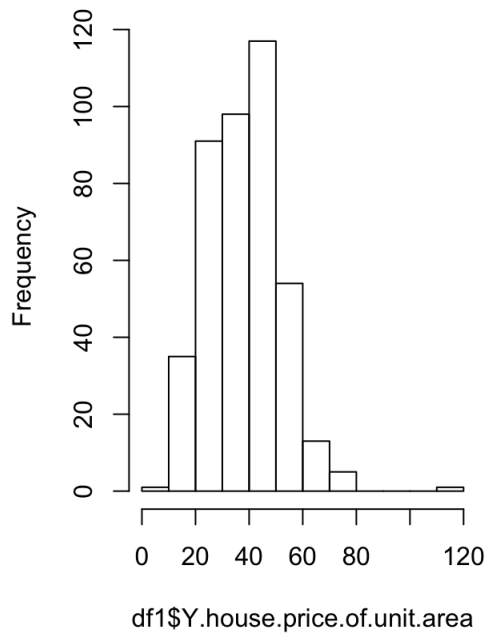
```
##      No Year X2.house.age X3.distance.to.the.nearest.MRT.station
## 271 271 2013          10.8                252.5822
##      X4.number.of.convenience.stores X5.latitude X6.longitude
## 271                                1    24.9746    121.5305
##      Y.house.price.of.unit.area Y2
## 271                        117.5 10.83974
```

```
# Figure 2
par(mfrow=c(2,2))
plot(df1$X2.house.age, df1$Y2, col = df1$Year, pch = 16,
     main = 'A. house age and house price',
     xlab = 'house.age', ylab = 'sqrt.house.price.of.unit.area')
plot(df1$X3.distance.to.the.nearest.MRT.station, df1$Y2, col = df1$Year, pch = 16,
     main = 'B. distance to MRT and house price',
     xlab = 'distance.to.MRT', ylab = 'sqrt.house.price.of.unit.area')
plot(df1$X4.number.of.convenience.stores, df1$Y2, col = df1$Year, pch = 16,
     main = 'C. number of stores and house price',
     xlab = 'number.of.convenience.stores', ylab = 'sqrt.house.price.of.unit.area')
plot(df1$X5.latitude, df1$X6.longitude, col = df1$Year, pch = 16,
     main = 'D. latitude vs. longitude',
     xlab = 'latitude', ylab = 'longitude')
```

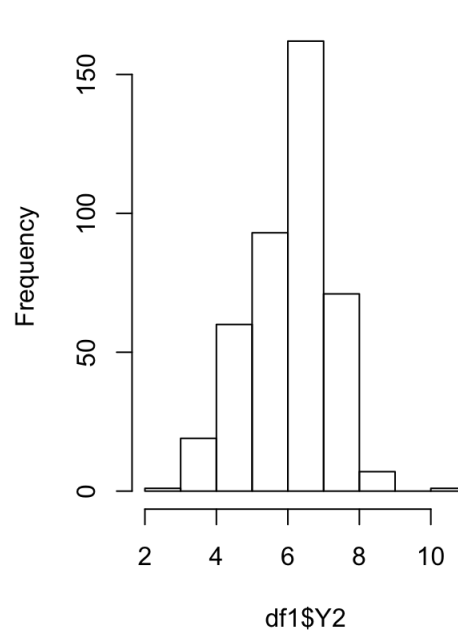


```
# Figure 1
par(mfrow=c(1,2))
hist(df1$Y.house.price.of.unit.area, main = 'A. Original scale')
hist(df1$Y2, main = 'B. Square-root scale')
```

A. Original scale



B. Square-root scale

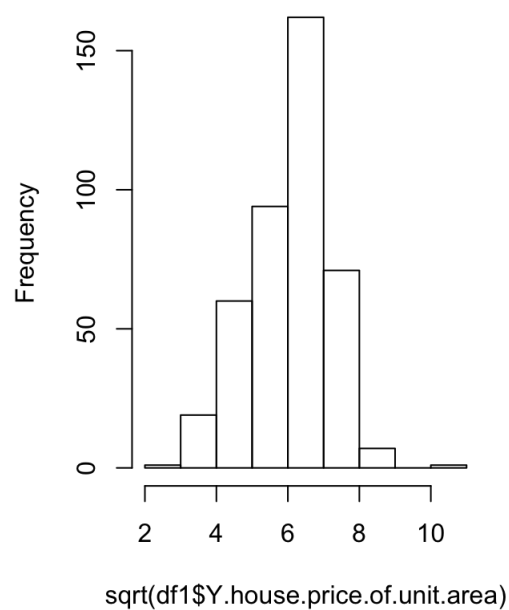
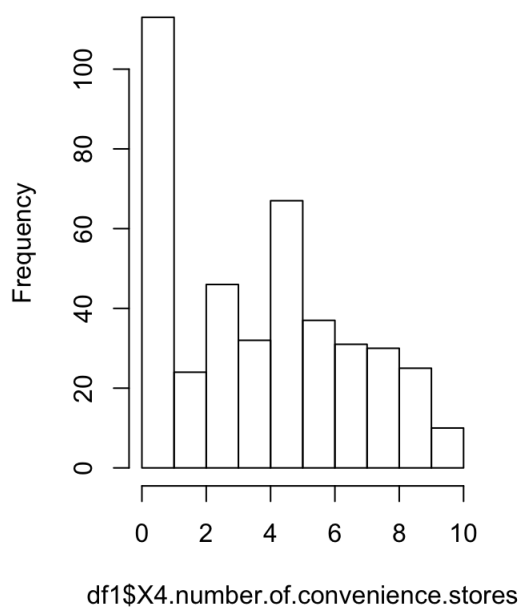


```
hist(df1$X4.number.of.convenience.stores)
fivenum(df1$X4.number.of.convenience.stores)
```

```
## [1] 0 1 4 6 10
```

```
#hist(log(df1$Y.house.price.of.unit.area))
hist(sqrt(df1$Y.house.price.of.unit.area))
```

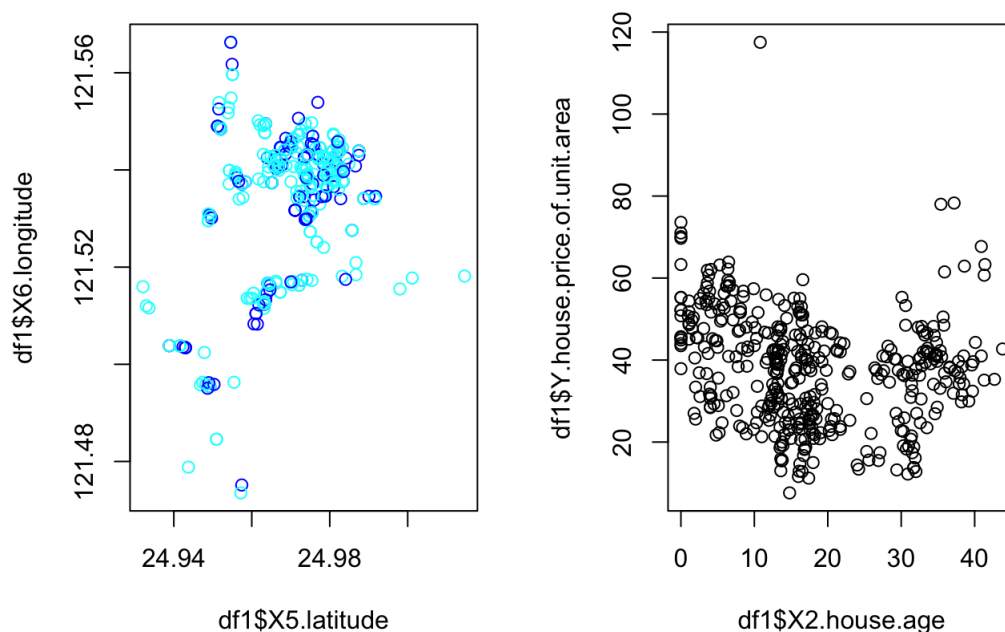
gram of df1\$X4.number.of.convenience stores gram of sqrt(df1\$Y.house.price.of.u



```
# test for normality
shapiro.test(sqrt(df1$Y.house.price.of.unit.area))
```

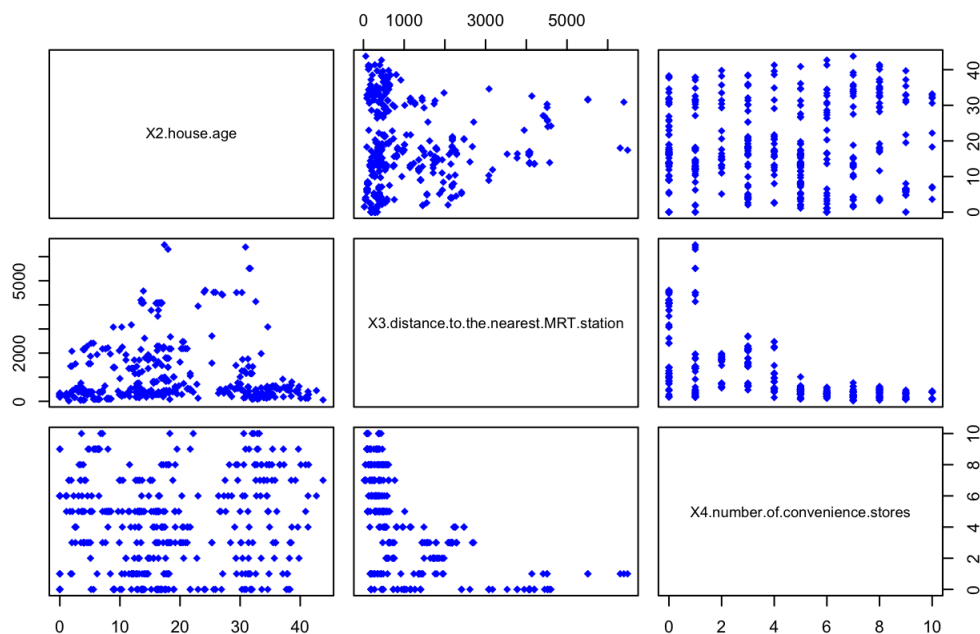
```
##
##  Shapiro-Wilk normality test
##
## data:  sqrt(df1$Y.house.price.of.unit.area)
## W = 0.98833, p-value = 0.002112
```

```
plot(df1$X5.latitude, df1$X6.longitude, col = df1$Year)
plot(df1$X2.house.age, df1$Y.house.price.of.unit.area)
```



```
# Figure 3
# check multi-collinearity
pairs(df1[,3:5], col= "blue", pch=18,
      main= "Relationship between variables")
```

Relationship between variables



```
# variable selection
```

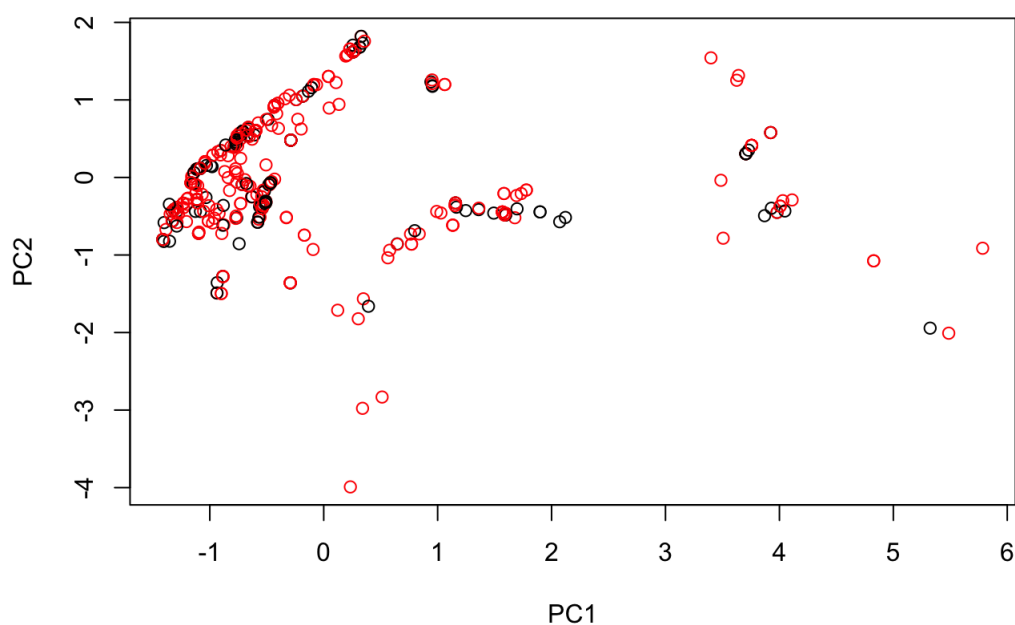
```
# PCA
```

```
df1$Year <- as.factor(df1$Year)
df1.pca <- prcomp(df1[,c(4, 6, 7)], center = TRUE, scale. = TRUE)
summary(df1.pca)
```

```
## Importance of components:
```

```
##              PC1      PC2      PC3
## Standard deviation    1.4908 0.7833 0.40468
## Proportion of Variance 0.7409 0.2045 0.05459
## Cumulative Proportion 0.7409 0.9454 1.00000
```

```
par(mfrow=c(1,1))
plot(df1.pca$x[,1:2], col = df1$Year)
```



```
# M0: Baseline model
```

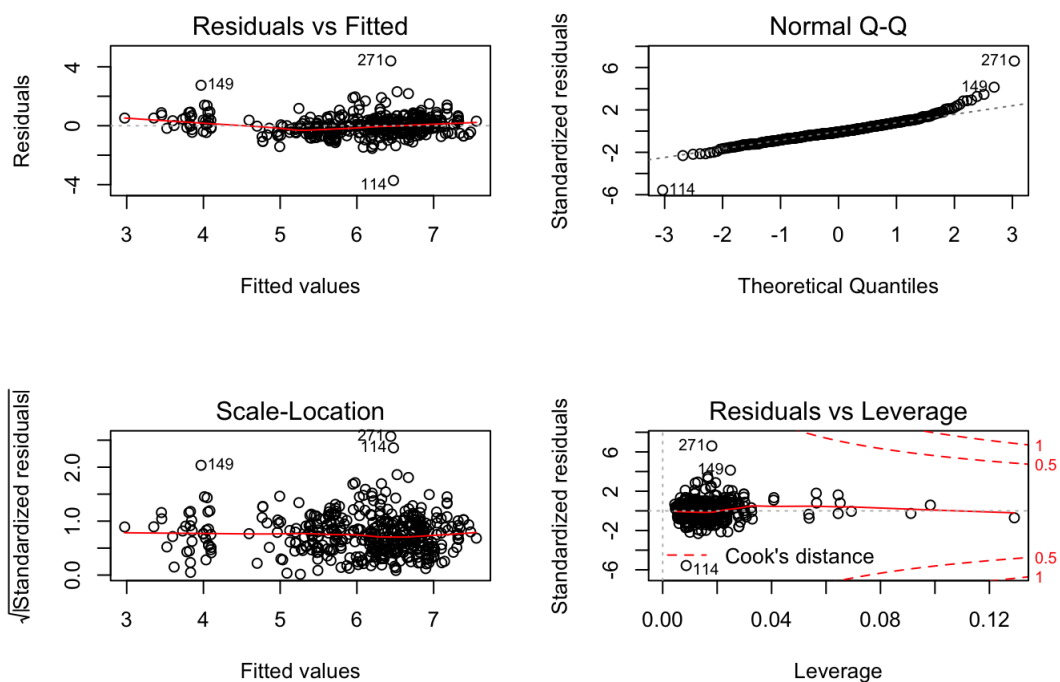
```
#df1$Y2 <- sqrt(df1$Y.house.price.of.unit.area)
my.lm0 <- lm(Y2 ~ ., data = df1[, -c(1, 8)])
vif(my.lm0)
```

```
##              Year              X2.house.age
##              1.005502              1.016652
## X3.distance.to.the.nearest.MRT.station    X4.number.of.convenience.stores
##              4.293775              1.613346
##              X5.latitude              X6.longitude
##              1.599937              2.931216
```

```
summary(my.lm0)
```

```
##
## Call:
## lm(formula = Y2 ~ ., data = df1[, -c(1, 8)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.7196 -0.3947 -0.0600  0.3361  4.3982
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -5.017e+02  4.659e+02  -1.077  0.28223
## Year2013         2.323e-01  7.193e-02   3.230  0.00134
## X2.house.age    -2.178e-02  2.925e-03  -7.447  5.74e-13
## X3.distance.to.the.nearest.MRT.station -3.870e-04  5.426e-05  -7.133  4.53e-12
## X4.number.of.convenience.stores      9.123e-02  1.425e-02   6.402  4.23e-10
## X5.latitude      2.157e+01  3.368e+00   6.403  4.20e-10
## X6.longitude    -2.510e-01  3.687e+00  -0.068  0.94576
##
## (Intercept)
## Year2013          **
## X2.house.age      ***
## X3.distance.to.the.nearest.MRT.station ***
## X4.number.of.convenience.stores      ***
## X5.latitude        ***
## X6.longitude
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6716 on 407 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.6472, Adjusted R-squared:  0.642
## F-statistic: 124.4 on 6 and 407 DF, p-value: < 2.2e-16
```

```
par(mfrow = c(2,2))
plot(my.lm0)
```



```
# M1: full model
df1$Y2 <- sqrt(df1$Y.house.price.of.unit.area)
my.lm1 <- lm(Y2 ~ (.)^2, data = df1[, -c(1, 8)])
#vif(my.lm1)
summary(my.lm1)
```

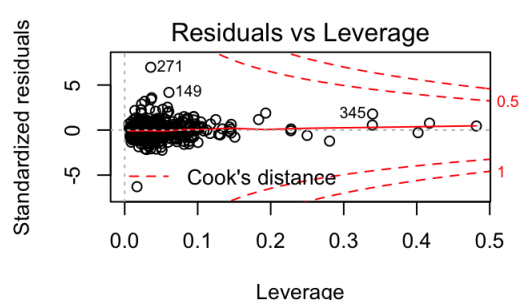
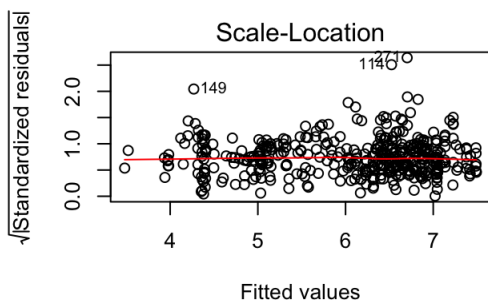
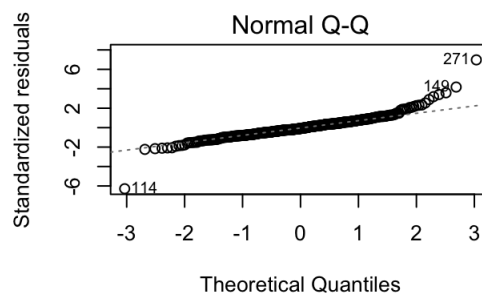
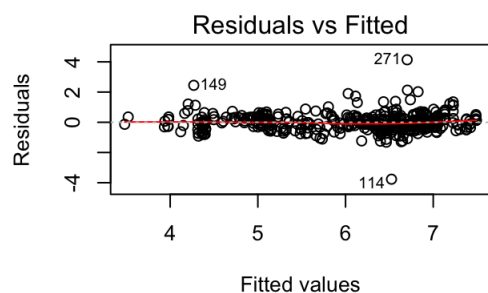
```
##
```

```
## Call:
## lm(formula = Y2 ~ (.)^2, data = df1[, -c(1, 8)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.7674 -0.3246 -0.0498  0.2796  4.1371
##
## Coefficients: (1 not defined because of singularities)
##
##                                     Estimate
## (Intercept)                        -2.295e+03
## Year2013                          -5.142e+02
## X2.house.age                       4.077e+01
## X3.distance.to.the.nearest.MRT.station 3.894e-01
## X4.number.of.convenience.stores     2.694e+02
## X5.latitude                        7.676e+01
## X6.longitude                       3.166e+00
## Year2013:X2.house.age               1.387e-02
## Year2013:X3.distance.to.the.nearest.MRT.station -1.867e-06
## Year2013:X4.number.of.convenience.stores -8.755e-03
## Year2013:X5.latitude                2.181e+00
## Year2013:X6.longitude               3.783e+00
## X2.house.age:X3.distance.to.the.nearest.MRT.station 3.911e-07
## X2.house.age:X4.number.of.convenience.stores 5.525e-04
## X2.house.age:X5.latitude            1.573e-03
## X2.house.age:X6.longitude          -3.360e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -1.141e-04
## X3.distance.to.the.nearest.MRT.station:X5.latitude -1.744e-02
## X3.distance.to.the.nearest.MRT.station:X6.longitude 3.752e-04
## X4.number.of.convenience.stores:X5.latitude -8.764e+00
## X4.number.of.convenience.stores:X6.longitude -4.149e-01
## X5.latitude:X6.longitude           NA
##
##                                     Std. Error
## (Intercept)                        1.314e+03
## Year2013                          8.964e+02
## X2.house.age                       5.419e+01
## X3.distance.to.the.nearest.MRT.station 2.402e-01
## X4.number.of.convenience.stores     2.993e+02
## X5.latitude                        1.319e+01
## X6.longitude                       1.021e+01
## Year2013:X2.house.age               5.801e-03
## Year2013:X3.distance.to.the.nearest.MRT.station 1.061e-04
## Year2013:X4.number.of.convenience.stores 2.783e-02
## Year2013:X5.latitude                7.441e+00
## Year2013:X6.longitude               7.030e+00
## X2.house.age:X3.distance.to.the.nearest.MRT.station 6.008e-06
## X2.house.age:X4.number.of.convenience.stores 1.065e-03
## X2.house.age:X5.latitude            3.350e-01
## X2.house.age:X6.longitude           4.400e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 2.961e-05
## X3.distance.to.the.nearest.MRT.station:X5.latitude 3.370e-03
## X3.distance.to.the.nearest.MRT.station:X6.longitude 2.073e-03
## X4.number.of.convenience.stores:X5.latitude 1.649e+00
## X4.number.of.convenience.stores:X6.longitude 2.380e+00
## X5.latitude:X6.longitude           NA
##
##                                     t value
## (Intercept)                        -1.747
## Year2013                          -0.574
## X2.house.age                       0.752
## X3.distance.to.the.nearest.MRT.station 1.621
## X4.number.of.convenience.stores     0.900
## X5.latitude                        5.820
## X6.longitude                       0.310
## Year2013:X2.house.age               2.391
## Year2013:X3.distance.to.the.nearest.MRT.station -0.018
## Year2013:X4.number.of.convenience.stores -0.315
## Year2013:X5.latitude                0.293
## Year2013:X6.longitude               0.538
## X2.house.age:X3.distance.to.the.nearest.MRT.station 0.065
## X2.house.age:X4.number.of.convenience.stores 0.519
## X2.house.age:X5.latitude            0.005
## X2.house.age:X6.longitude          -0.764
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -3.854
## X3.distance.to.the.nearest.MRT.station:X5.latitude -5.173
```



```
## X3.distance.to.the.nearest.MRT.station:X5.latitude 0.1275
## X3.distance.to.the.nearest.MRT.station:X6.longitude 0.181
## X4.number.of.convenience.stores:X5.latitude -5.315
## X4.number.of.convenience.stores:X6.longitude -0.174
## X5.latitude:X6.longitude NA
## Pr(>|t|)
## (Intercept) 0.081421
## Year2013 0.566525
## X2.house.age 0.452275
## X3.distance.to.the.nearest.MRT.station 0.105765
## X4.number.of.convenience.stores 0.368655
## X5.latitude 1.22e-08
## X6.longitude 0.756710
## Year2013:X2.house.age 0.017248
## Year2013:X3.distance.to.the.nearest.MRT.station 0.985973
## Year2013:X4.number.of.convenience.stores 0.753206
## Year2013:X5.latitude 0.769600
## Year2013:X6.longitude 0.590792
## X2.house.age:X3.distance.to.the.nearest.MRT.station 0.948132
## X2.house.age:X4.number.of.convenience.stores 0.604252
## X2.house.age:X5.latitude 0.996257
## X2.house.age:X6.longitude 0.445451
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 0.000136
## X3.distance.to.the.nearest.MRT.station:X5.latitude 3.68e-07
## X3.distance.to.the.nearest.MRT.station:X6.longitude 0.856448
## X4.number.of.convenience.stores:X5.latitude 1.79e-07
## X4.number.of.convenience.stores:X6.longitude 0.861705
## X5.latitude:X6.longitude NA
##
## (Intercept) .
## Year2013
## X2.house.age
## X3.distance.to.the.nearest.MRT.station
## X4.number.of.convenience.stores
## X5.latitude ***
## X6.longitude *
## Year2013:X2.house.age
## Year2013:X3.distance.to.the.nearest.MRT.station
## Year2013:X4.number.of.convenience.stores
## Year2013:X5.latitude
## Year2013:X6.longitude
## X2.house.age:X3.distance.to.the.nearest.MRT.station
## X2.house.age:X4.number.of.convenience.stores
## X2.house.age:X5.latitude
## X2.house.age:X6.longitude
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores ***
## X3.distance.to.the.nearest.MRT.station:X5.latitude ***
## X3.distance.to.the.nearest.MRT.station:X6.longitude
## X4.number.of.convenience.stores:X5.latitude ***
## X4.number.of.convenience.stores:X6.longitude
## X5.latitude:X6.longitude
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6042 on 394 degrees of freedom
## Multiple R-squared:  0.7238, Adjusted R-squared:  0.7098
## F-statistic: 51.62 on 20 and 394 DF, p-value: < 2.2e-16
```

```
par(mfrow = c(2,2))
plot(my.lm1)
```

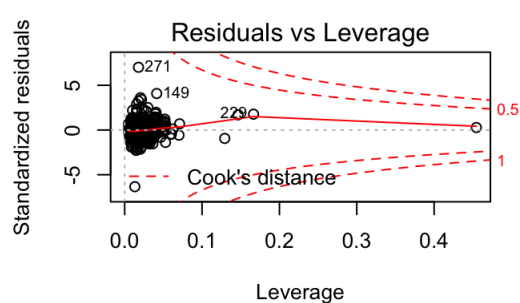
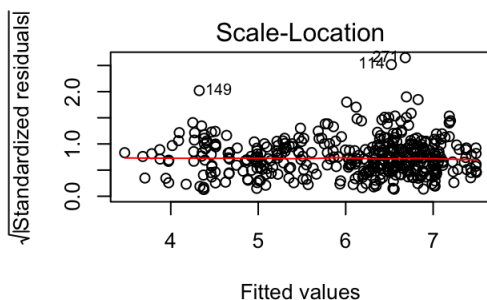
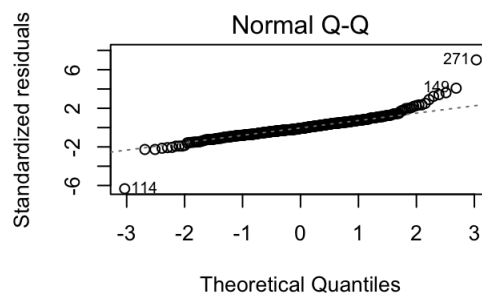
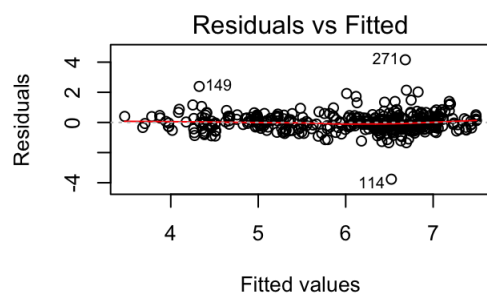


```
# M2: stepwise
fit.start <- lm(Y2 ~ 1, data = df1[, -c(1, 8)])
fit.end <- lm(Y2 ~ (. )^2, data = df1[, -c(1, 8)])
step.BIC = stepAIC(fit.start, list(upper = fit.end),
  K = log(nrow(df1[, -c(1, 8)])), trace = 0)
summary(step.BIC)
```

```
##
## Call:
## lm(formula = Y2 ~ X3.distance.to.the.nearest.MRT.station + X4.number.of.convenience.stores +
##   X2.house.age + X5.latitude + Year + X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.s
##   tores +
##   X2.house.age:Year + X3.distance.to.the.nearest.MRT.station:X5.latitude +
##   X4.number.of.convenience.stores:X5.latitude, data = df1[,
##     -c(1, 8)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.7651 -0.3264 -0.0508  0.2851  4.1582
##
## Coefficients:
##              Estimate
## (Intercept)      -1.926e+03
## X3.distance.to.the.nearest.MRT.station    4.237e-01
## X4.number.of.convenience.stores          2.157e+02
## X2.house.age       -3.068e-02
## X5.latitude         7.740e+01
## Year2013           -1.352e-02
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -1.114e-04
## X2.house.age:Year2013                1.404e-02
## X3.distance.to.the.nearest.MRT.station:X5.latitude -1.698e-02
## X4.number.of.convenience.stores:X5.latitude -8.633e+00
##              Std. Error
## (Intercept)      1.965e+02
## X3.distance.to.the.nearest.MRT.station    6.587e-02
## X4.number.of.convenience.stores          3.423e+01
## X2.house.age       4.702e-03
## X5.latitude         7.870e+00
## Year2013           1.160e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 1.823e-05
## X2.house.age:Year2013                5.626e-03
## X3.distance.to.the.nearest.MRT.station:X5.latitude 2.639e-03
## X4.number.of.convenience.stores:X5.latitude 1.371e+00
##              t value
```

```
## (Intercept) -9.801
## X3.distance.to.the.nearest.MRT.station 6.432
## X4.number.of.convenience.stores 6.301
## X2.house.age -6.526
## X5.latitude 9.834
## Year2013 -0.117
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -6.111
## X2.house.age:Year2013 2.496
## X3.distance.to.the.nearest.MRT.station:X5.latitude -6.435
## X4.number.of.convenience.stores:X5.latitude -6.298
## Pr(>|t|)
## (Intercept) < 2e-16
## X3.distance.to.the.nearest.MRT.station 3.56e-10
## X4.number.of.convenience.stores 7.72e-10
## X2.house.age 2.02e-10
## X5.latitude < 2e-16
## Year2013 0.907
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 2.32e-09
## X2.house.age:Year2013 0.013
## X3.distance.to.the.nearest.MRT.station:X5.latitude 3.49e-10
## X4.number.of.convenience.stores:X5.latitude 7.87e-10
##
## (Intercept) ***
## X3.distance.to.the.nearest.MRT.station ***
## X4.number.of.convenience.stores ***
## X2.house.age ***
## X5.latitude ***
## Year2013
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores ***
## X2.house.age:Year2013 *
## X3.distance.to.the.nearest.MRT.station:X5.latitude ***
## X4.number.of.convenience.stores:X5.latitude ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5981 on 405 degrees of freedom
## Multiple R-squared:  0.7218, Adjusted R-squared:  0.7157
## F-statistic: 116.8 on 9 and 405 DF,  p-value: < 2.2e-16
```

```
plot(step.BIC)
```



```
df1$Y2 <- sqrt(df1$Y.house.price.of.unit.area)
my.lm1 <- lm(Y2 ~., data = df1[, -c(1, 8)])
vif(my.lm1)
```

```
##
##              Year              X2.house.age
##              1.005202              1.016661
## X3.distance.to.the.nearest.MRT.station      X4.number.of.convenience.stores
##              4.294917              1.612241
##              X5.latitude              X6.longitude
##              1.599704              2.931653
```

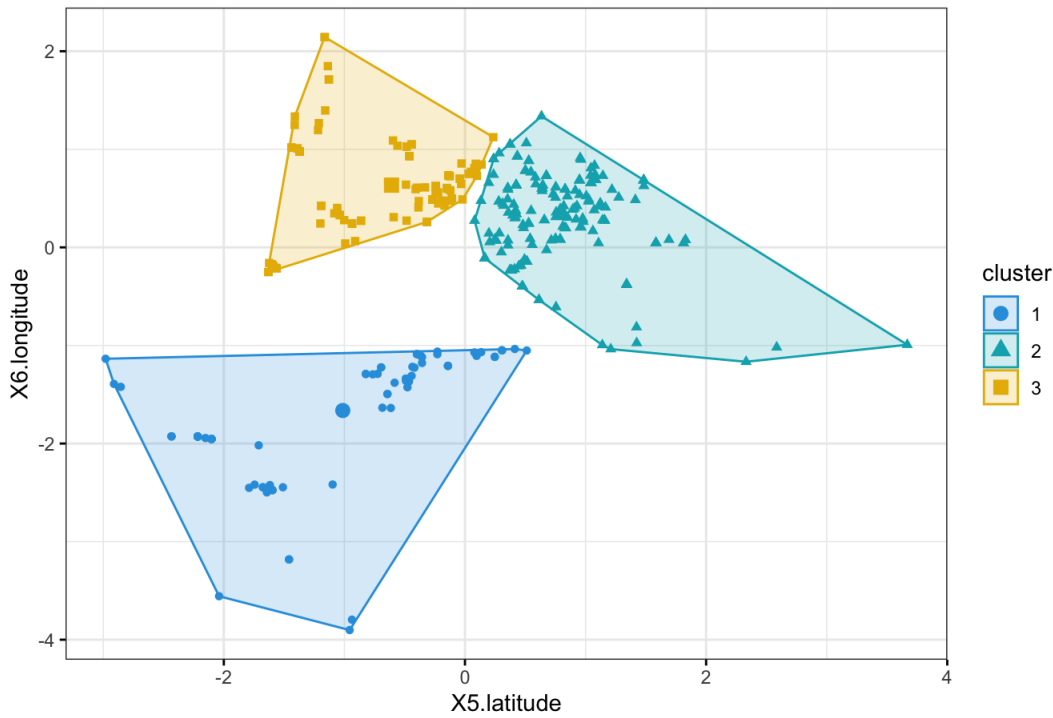
```
summary(my.lm1)
```

```
##
## Call:
## lm(formula = Y2 ~ ., data = df1[, -c(1, 8)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.7233 -0.4009 -0.0593  0.3415  4.3993
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      -4.666e+02  4.664e+02  -1.000  0.317680
## Year2013           2.413e-01  7.187e-02   3.357  0.000862
## X2.house.age      -2.179e-02  2.931e-03  -7.433  6.29e-13
## X3.distance.to.the.nearest.MRT.station -3.890e-04  5.435e-05  -7.156  3.89e-12
## X4.number.of.convenience.stores       9.184e-02  1.427e-02   6.434  3.49e-10
## X5.latitude        2.128e+01  3.370e+00   6.313  7.16e-10
## X6.longitude      -4.798e-01  3.691e+00  -0.130  0.896656
##
## (Intercept)
## Year2013          ***
## X2.house.age      ***
## X3.distance.to.the.nearest.MRT.station ***
## X4.number.of.convenience.stores      ***
## X5.latitude       ***
## X6.longitude
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.673 on 408 degrees of freedom
## Multiple R-squared:  0.6452, Adjusted R-squared:  0.64
## F-statistic: 123.7 on 6 and 408 DF,  p-value: < 2.2e-16
```

```
# K-means clustering
set.seed(2020)
km.res <- kmeans(df1[,c(6, 7)], 3, nstart = 25)

fviz_cluster(km.res, data = df1[,c(6, 7)],
              palette = c("#2E9FDF", "#00AFBB", "#E7B800"),
              geom = "point",
              ellipse.type = "convex",
              ggtheme = theme_bw()
)
```

Cluster plot



```
df3 <- cbind(df1, cluster = km.res$cluster)
head(df3)
```

```
##      No Year X2.house.age X3.distance.to.the.nearest.MRT.station
## 1  1 2012          32.0                                84.87882
## 2  2 2012          19.5                               306.59470
## 3  3 2013          13.3                               561.98450
## 4  4 2013          13.3                               561.98450
## 5  5 2012           5.0                               390.56840
## 6  6 2012           7.1                               2175.03000
##      X4.number.of.convenience.stores X5.latitude X6.longitude
## 1                                10    24.98298    121.5402
## 2                                 9    24.98034    121.5395
## 3                                 5    24.98746    121.5439
## 4                                 5    24.98746    121.5439
## 5                                 5    24.97937    121.5425
## 6                                 3    24.96305    121.5125
##      Y.house.price.of.unit.area      Y2 cluster
## 1              37.9 6.156298          2
## 2              42.2 6.496153          2
## 3              47.3 6.877500          2
## 4              54.8 7.402702          2
## 5              43.1 6.565059          2
## 6              32.1 5.665686          1
```

```
names(df3)
```

```
## [1] "No"
## [2] "Year"
## [3] "X2.house.age"
## [4] "X3.distance.to.the.nearest.MRT.station"
## [5] "X4.number.of.convenience.stores"
## [6] "X5.latitude"
## [7] "X6.longitude"
## [8] "Y.house.price.of.unit.area"
## [9] "Y2"
## [10] "cluster"
```

```
df3$cluster <- as.factor(df3$cluster)
my.lm2 <- lm(Y2 ~., data = df3[, c(2:5, 9:10)])
vif(my.lm2)
```

	GVIF	Df	GVIF^(1/(2*Df))
## Year	1.003309	1	1.001653
## X2.house.age	1.026710	1	1.013267
## X3.distance.to.the.nearest.MRT.station	3.557034	1	1.886010
## X4.number.of.convenience.stores	1.620469	1	1.272976
## cluster	2.716769	2	1.283847

```
summary(my.lm2)
```

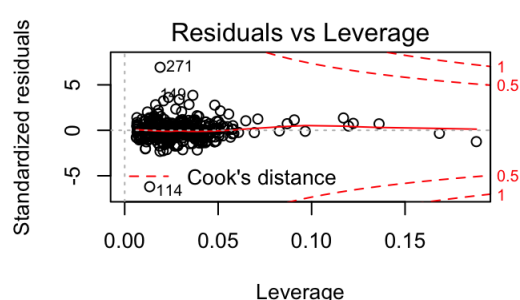
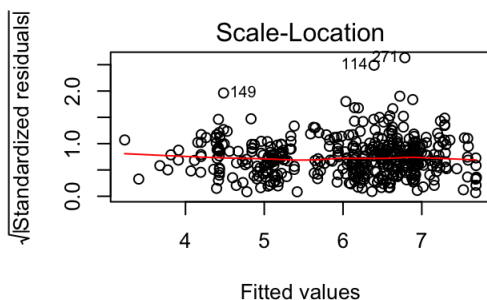
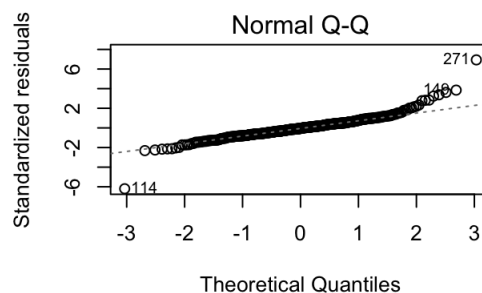
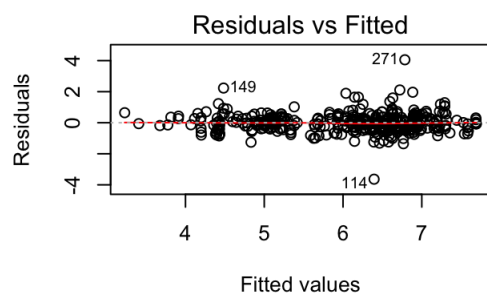
```
##
## Call:
## lm(formula = Y2 ~ ., data = df3[, c(2:5, 9:10)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5898 -0.3796 -0.0341  0.2872  4.2499
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5.781e+00  1.742e-01  33.179 < 2e-16
## Year2013         2.352e-01  6.813e-02   3.453 0.000613
## X2.house.age     -2.325e-02  2.795e-03  -8.320 1.33e-15
## X3.distance.to.the.nearest.MRT.station -3.304e-04  4.693e-05  -7.039 8.25e-12
## X4.number.of.convenience.stores      1.072e-01  1.358e-02   7.894 2.73e-14
## cluster2         8.013e-01  1.305e-01   6.138 1.98e-09
## cluster3         1.616e-01  1.309e-01   1.235 0.217524
##
## (Intercept)      ***
## Year2013          ***
## X2.house.age      ***
## X3.distance.to.the.nearest.MRT.station ***
## X4.number.of.convenience.stores      ***
## cluster2          ***
## cluster3
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6385 on 408 degrees of freedom
## Multiple R-squared:  0.6806, Adjusted R-squared:  0.6759
## F-statistic: 144.9 on 6 and 408 DF, p-value: < 2.2e-16
```

```
fit.start <- lm(Y2 ~., data = df3[, c(2:5, 9:10)])
fit.end <- lm(Y2 ~ (.^2, data = df3[, c(2:5, 9:10)])
step.BIC2 = step(fit.start, list(upper = fit.end),
                 K = log(nrow(df1[, -c(1, 8)])), trace = 0)
summary(step.BIC2)
```

```
##
## Call:
## lm(formula = Y2 ~ Year + X2.house.age + X3.distance.to.the.nearest.MRT.station +
##      X4.number.of.convenience.stores + cluster + X4.number.of.convenience.stores:cluster +
##      X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores +
##      X2.house.age:X3.distance.to.the.nearest.MRT.station + Year:X2.house.age,
##      data = df3[, c(2:5, 9:10)])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.6374 -0.3397 -0.0325  0.2727  4.0549
##
## Coefficients:
##              Estimate
## (Intercept)      5.489e+00
## Year2013         5.327e-02
## X2.house.age     -3.289e-02
## X3.distance.to.the.nearest.MRT.station -2.863e-04
## X4.number.of.convenience.stores      4.560e-01
## cluster2         1.467e+00
## cluster3         9.378e-02
## X4.number.of.convenience.stores:cluster2 -3.479e-01
## X4.number.of.convenience.stores:cluster3 -1.992e-01
```

```
## A4.number.of.convenience.stores:cluster3 -1.772e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -1.618e-04
## X2.house.age:X3.distance.to.the.nearest.MRT.station 6.173e-06
## Year2013:X2.house.age 1.097e-02
## Std. Error
## (Intercept) 3.179e-01
## Year2013 1.151e-01
## X2.house.age 4.940e-03
## X3.distance.to.the.nearest.MRT.station 8.856e-05
## X4.number.of.convenience.stores 1.086e-01
## cluster2 2.743e-01
## cluster3 2.403e-01
## X4.number.of.convenience.stores:cluster2 1.010e-01
## X4.number.of.convenience.stores:cluster3 9.612e-02
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 3.209e-05
## X2.house.age:X3.distance.to.the.nearest.MRT.station 2.838e-06
## Year2013:X2.house.age 5.573e-03
## t value
## (Intercept) 17.267
## Year2013 0.463
## X2.house.age -6.658
## X3.distance.to.the.nearest.MRT.station -3.234
## X4.number.of.convenience.stores 4.197
## cluster2 5.349
## cluster3 0.390
## X4.number.of.convenience.stores:cluster2 -3.443
## X4.number.of.convenience.stores:cluster3 -2.072
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -5.041
## X2.house.age:X3.distance.to.the.nearest.MRT.station 2.175
## Year2013:X2.house.age 1.968
## Pr(>|t|)
## (Intercept) < 2e-16
## Year2013 0.643659
## X2.house.age 9.09e-11
## X3.distance.to.the.nearest.MRT.station 0.001324
## X4.number.of.convenience.stores 3.33e-05
## cluster2 1.48e-07
## cluster3 0.696528
## X4.number.of.convenience.stores:cluster2 0.000636
## X4.number.of.convenience.stores:cluster3 0.038886
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores 7.02e-07
## X2.house.age:X3.distance.to.the.nearest.MRT.station 0.030185
## Year2013:X2.house.age 0.049802
## ***
## (Intercept) ***
## Year2013 ***
## X2.house.age ***
## X3.distance.to.the.nearest.MRT.station **
## X4.number.of.convenience.stores ***
## cluster2 ***
## cluster3 ***
## X4.number.of.convenience.stores:cluster2 ***
## X4.number.of.convenience.stores:cluster3 *
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores ***
## X2.house.age:X3.distance.to.the.nearest.MRT.station *
## Year2013:X2.house.age *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5905 on 403 degrees of freedom
## Multiple R-squared: 0.7301, Adjusted R-squared: 0.7228
## F-statistic: 99.13 on 11 and 403 DF, p-value: < 2.2e-16
```

```
plot(step.BIC2)
```



```
names(step.BIC2)
```

```
## [1] "coefficients" "residuals"      "effects"      "rank"
## [5] "fitted.values" "assign"        "qr"           "df.residual"
## [9] "contrasts"     "xlevels"       "call"         "terms"
## [13] "model"         "anova"
```

```
step.BIC2$coefficients
```

```
## (Intercept)
## 5.489256e+00
## Year2013
## 5.327209e-02
## X2.house.age
## -3.288955e-02
## X3.distance.to.the.nearest.MRT.station
## -2.863489e-04
## X4.number.of.convenience.stores
## 4.559547e-01
## cluster2
## 1.467409e+00
## cluster3
## 9.377991e-02
## X4.number.of.convenience.stores:cluster2
## -3.478953e-01
## X4.number.of.convenience.stores:cluster3
## -1.991656e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores
## -1.617496e-04
## X2.house.age:X3.distance.to.the.nearest.MRT.station
## 6.172852e-06
## Year2013:X2.house.age
## 1.096547e-02
```

```
confint(step.BIC2)
```



```

##                                     2.5 %
## (Intercept)                        4.864288e+00
## Year2013                          -1.729483e-01
## X2.house.age                      -4.260016e-02
## X3.distance.to.the.nearest.MRT.station -4.604399e-04
## X4.number.of.convenience.stores    2.423712e-01
## cluster2                          9.281539e-01
## cluster3                         -3.785827e-01
## X4.number.of.convenience.stores:cluster2 -5.465367e-01
## X4.number.of.convenience.stores:cluster3 -3.881154e-01
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -2.248320e-04
## X2.house.age:X3.distance.to.the.nearest.MRT.station 5.944025e-07
## Year2013:X2.house.age              9.540238e-06
##                                     97.5 %
## (Intercept)                        6.114225e+00
## Year2013                          2.794925e-01
## X2.house.age                      -2.317895e-02
## X3.distance.to.the.nearest.MRT.station -1.122579e-04
## X4.number.of.convenience.stores    6.695381e-01
## cluster2                          2.006664e+00
## cluster3                          5.661425e-01
## X4.number.of.convenience.stores:cluster2 -1.492538e-01
## X4.number.of.convenience.stores:cluster3 -1.021575e-02
## X3.distance.to.the.nearest.MRT.station:X4.number.of.convenience.stores -9.866711e-05
## X2.house.age:X3.distance.to.the.nearest.MRT.station 1.175130e-05
## Year2013:X2.house.age              2.192140e-02

```

```

my.ci <- cbind(step.BIC2$coefficients, confint(step.BIC2))
write.csv(my.ci, 'ci.csv')

```